Nighut.Project.Rmd

Sheetal Gulab Nighut

04/07/2023

Breast cancer sub-type classification using proteomic data

Problem statement: Luminal A, Luminal B, HER2, Basal-like are the molecular subtypes of breast cancer that are typically recognized. Each one is related to various prognoses, therapies, and treatments.

Objective: The main objective of this project is to perform classification on breast cancer dataset based on the proteomic expression dataset to identify the subtype of breast cancer. A collection of proteins produced by cancer cells is known as the breast cancer proteome.

I am examining if the cancer subtype can be correctly identified using only the proteome data for each patient. The target variable in the study is PAM50 mRNA which is used in breast cancer intrinsic subtyping based on gene expression. The variable is categorical presenting four subtypes (basal-like, HER2-enriched, Luminal A and Luminal B). PAMA50 mRNA proteins are my predictor variables. Using supervised ML classification, I am examining if the cancer subtype can be correctly identified using only the proteome data for each patient.

• Question 1 : where does the data come from?

This data collection includes 77 breast cancer samples from the Clinical Proteomic Tumor Analysis Consortium (NCI/NIH) that have had their iTRAQ proteome profiles published. For each sample, it comprises expression levels for around 12,000 proteins, with missing values present when a particular protein could not be measured. Kaggle URL: https://www.kaggle.com/datasets/piotrgrabo/breastcancerproteomes?resource=download

It has three different files: 77cancerproteomesCPTACitraq.csv, clinicaldatabreast_cancer.csv, PAM50_proteins.csv

- 1. File: 77cancerproteomesCPTACitraq.csv -RefSeqaccessionnumber: RefSeq protein ID (each protein has a unique ID in a RefSeq database) -gene_symbol: a symbol unique to each gene (every protein is encoded by some gene) -gene_name: a full name of that gene -Remaining columns: log2 iTRAQ ratios for each sample (protein expression data, most important), three last columns are from healthy individuals
- 2. File: clinicaldatabreast_cancer.csv -First column "Complete TCGA ID" is used to match the sample IDs in the main cancer proteomes file (see example script). -All other columns have self-explanatory names, contain data about the cancer classification of a given sample using different methods.
- 3. File: PAM50_proteins.csv -Contains the list of genes and proteins used by the PAM50 classification system. -The column RefSeqProteinID contains the protein IDs that can be matched with the IDs in the main protein expression data set.
- Provided installation statements for all packages
- Importing all required libraries

```
# install.packages("stabs")
# install.packages("factoextra")
# install.packages("NbClust")
# install.packages("qqfortify")
# install.packages("qlmnet")
# install.packages("foreign")
# install.packages("ggplot2")
# install.packages("MASS")
# install.packages("Hmisc")
# install.packages("reshape2")
# install.packages("randomForest")
# install.packages("data.table")
# install.packages("mlr")
# install.packages("caret")
# install.packages("dplyr") # data manipulation
# install.packages("readr") # data input and manipulation
# install.packages("MASS") # contains the data
# install.packages("DataExplorer") #data set visualization
# install.packages("nnet") # used for Multinomial Classification
# install.packages("readr") #assist with text manipulation
# install.packages("kernlab") #assist with SVM feature selection
# install.packages("class") # used for an object-oriented style of programmin
# install.packages("KernelKnn") # used for K- Nearest-Neighbors method
# install.packages("nnet") # Used for Neural Net
# install.packages("e1071") #supports vector machine algorithm
# install.packages("forecast") # for model prediction
# install.pacakges("rpart") #construct recursive partitions for classification
# install.packages("neuralnet")
# install.packages("NbClust")
# install.packages("psych")
# install.packages("pheatmap")
# install.packages("OneR")
# install.packages("naivebayes")
# install.packages("dplyr")
# install.packages("qqplot2")
# install.packages("psych")
# install.packages("tidyverse")
# install.packages("assertr")
# install.packages("knitr")
# install.packages("psych")
# install.packages("caret")
# install.packages("e1071")
# install.packages("C50")
# install.packages("qmodels")
# install.packages("pROC")
# install.packages("caTools")
#install.packages("tinytex")
#library(tinytex)
```

• Importing all required libraries

```
#loading required libraries
library(stabs)
library(factoextra)
library(NbClust)
library(ggfortify)
library(glmnet)
library(foreign)
library(ggplot2)
library(MASS)
library(Hmisc)
library(reshape2)
library(randomForest)
library(data.table)
library(mlr)
library(caret)
library(dplyr) # data manipulation
library(readr) # data input and manipulation
library(caret) #select tuning parameters
library(MASS) # contains the data
library(DataExplorer) #data set visualization
library(nnet) # used for Multinomial Classification
library(readr) #assist with text manipulation
library(kernlab) #assist with SVM feature selection
library(class) # used for an object-oriented style of programmin
library(KernelKnn) # used for K- Nearest-Neighbors method
library(nnet) # Used for Neural Net
library(e1071) #supports vector machine algorithm
library(forecast) # for model prediction
library(rpart) #construct recursive partitions for classification
library(neuralnet)
library(NbClust)
library(psych)
library(pheatmap)
library(OneR)
library(naivebayes)
library(dplyr)
library(ggplot2)
library(psych)
library(pROC)
library(tidyverse)
library(assertr)
library(knitr)
library(psych)
library(caret)
library(e1071)
library(C50)
library(gmodels)
library(pROC)
library(caTools)
library(ggplot2)
library(GGally)
library(ipred)
```

1.Data Acquisition and manipulation

- for importing data, I used the read.csv() function
- Later, I combined all the datasets into one
- Converted the patients id to one format
- used the head(), str(), dim(), glimpse(), summary() to explore the dataset

Data import - PAM50_proteins.csv

First steps: importing the data and getting the data set into a workable format. list of genes and proteins

```
#Importing data which has list of genes and proteins
gene_proteins <- read.csv("/Users/sheetalnighut/Desktop/cancer-project/PAM50_proteins.csv")
# 100 rows and 4 columns
head(gene_proteins)</pre>
```

```
##
     GeneSymbol RefSeqProteinID
                                      Species
## 1
            MIA
                      NP_006524 Homo sapiens
## 2
          FGFR4
                      NP 002002 Homo sapiens
## 3
          FGFR4
                      NP_998812 Homo sapiens
## 4
          FGFR4
                      NP_075252 Homo sapiens
                      NP_055188 Homo sapiens
## 5
         GPR160
## 6
         ACTR3B
                      NP_065178 Homo sapiens
##
                                           Gene.Name
                       melanoma inhibitory activity
## 1
## 2
                fibroblast growth factor receptor 4
## 3
                fibroblast growth factor receptor 4
## 4
                fibroblast growth factor receptor 4
                     G protein-coupled receptor 160
## 6 ARP3 actin-related protein 3 homolog B (yeast)
```

tail(gene_proteins)

```
##
       GeneSymbol RefSegProteinID
                                        Species
## 95
             MDM2
                        NP 002383 Homo sapiens
## 96
            FOXC1
                        NP_001444 Homo sapiens
## 97
             GRB7
                     NP_001025173 Homo sapiens
## 98
             GRB7
                        NP_005301 Homo sapiens
## 99
             MELK
                        NP_055606 Homo sapiens
## 100
            UBE2T
                        NP_054895 Homo sapiens
##
                                          Gene.Name
## 95
          Mdm2 p53 binding protein homolog (mouse)
## 96
                                    forkhead box C1
## 97
            growth factor receptor-bound protein 7
## 98
            growth factor receptor-bound protein 7
          maternal embryonic leucine zipper kinase
## 100 ubiquitin-conjugating enzyme E2T (putative)
```

```
str(gene_proteins)
```

```
## 'data.frame': 100 obs. of 4 variables:
```

```
## $ GeneSymbol
                     : chr
                            "MIA" "FGFR4" "FGFR4" "FGFR4" ...
## $ RefSeqProteinID: chr
                            "NP_006524" "NP_002002" "NP_998812" "NP_075252" ...
## $ Species
                            "Homo sapiens" "Homo sapiens" "Homo sapiens" "Homo sapiens" ...
                     : chr
## $ Gene.Name
                            "melanoma inhibitory activity" "fibroblast growth factor receptor 4" "fibro
                     : chr
dim(gene_proteins)
## [1] 100
glimpse(gene_proteins)
## Rows: 100
## Columns: 4
## $ GeneSymbol
                     <chr> "MIA", "FGFR4", "FGFR4", "FGFR4", "GPR160", "ACTR3B", ~
## $ RefSeqProteinID <chr> "NP_006524", "NP_002002", "NP_998812", "NP_075252", "N~
## $ Species
                     <chr> "Homo sapiens", "Homo sapiens", "Homo sapiens", "Homo ~
## $ Gene.Name
                     <chr> "melanoma inhibitory activity", "fibroblast growth fac~
summary(gene_proteins)
    GeneSymbol
                       RefSeqProteinID
                                                              Gene.Name
                                            Species
## Length:100
                       Length:100
                                          Length:100
                                                             Length: 100
                       Class :character
                                          Class :character
                                                             Class : character
## Class :character
## Mode :character
                       Mode :character
                                          Mode :character
                                                             Mode :character
Data import - clinicaldatabreast cancer.csv
```

main cancer dataset (has information about patients suffering from each sub-type classified based on PAM50.mRNA)

```
clinical <- read.csv("/Users/sheetalnighut/Desktop/cancer-project/clinical_data_breast_cancer.csv")
# 105 obseravtions and 30 columns
head(clinical)</pre>
```

```
##
     Complete.TCGA.ID Gender Age.at.Initial.Pathologic.Diagnosis ER.Status
## 1
        TCGA-A2-A0T2 FEMALE
                                                              66 Negative
        TCGA-A2-AOCM FEMALE
## 2
                                                              40 Negative
## 3
        TCGA-BH-A18V FEMALE
                                                              48 Negative
                                                              56 Negative
## 4
        TCGA-BH-A18Q FEMALE
## 5
        TCGA-BH-AOEO FEMALE
                                                              38 Negative
## 6
        TCGA-A7-A0CE FEMALE
                                                              57 Negative
    PR.Status HER2.Final.Status Tumor Tumor..T1.Coded Node Node.Coded Metastasis
##
## 1 Negative
                        Negative
                                    TЗ
                                               T Other
                                                              Positive
                                               T_Other
## 2 Negative
                        Negative
                                    T2
                                                              Negative
                                                                               MO
                                                         NO
     Negative
                        Negative
                                    T2
                                               T_Other
                                                         N1
                                                              Positive
                                                                               MO
     Negative
                                    T2
                                               T_Other N1
## 4
                        Negative
                                                              Positive
                                                                               MO
## 5
     Negative
                        Negative
                                    Т3
                                               T_{Other}
                                                        NЗ
                                                              Positive
                                                                               MO
     Negative
                        Negative
                                    T2
                                               T_Other NO
                                                              Negative
## 6
                                                                               MO
    Metastasis.Coded AJCC.Stage Converted.Stage Survival.Data.Form Vital.Status
                        Stage IV
                                   No_Conversion
## 1
            Positive
                                                           followup
                                                                        DECEASED
```

```
## 2
             Negative Stage IIA
                                         Stage IIA
                                                              followup
                                                                            DECEASED
## 3
             Negative Stage IIB
                                    No_Conversion
                                                            enrollment
                                                                            DECEASED
## 4
             Negative Stage IIB
                                                            enrollment
                                     No Conversion
                                                                            DECEASED
## 5
             Negative Stage IIIC
                                     No_Conversion
                                                                              LIVING
                                                              followup
             Negative Stage IIA
                                         Stage IIA
                                                              followup
                                                                              LIVING
     Days.to.Date.of.Last.Contact Days.to.date.of.Death OS.event OS.Time
                                                       240
                                                                  1
## 2
                                                                         754
                               754
                                                       754
                                                                   1
## 3
                              1555
                                                      1555
                                                                  1
                                                                        1555
## 4
                              1692
                                                      1692
                                                                  1
                                                                        1692
## 5
                               133
                                                        NA
                                                                         133
## 6
                               309
                                                        NA
                                                                  0
                                                                         309
     PAM50.mRNA SigClust.Unsupervised.mRNA SigClust.Intrinsic.mRNA miRNA.Clusters
## 1 Basal-like
                                           0
                                                                  -13
## 2 Basal-like
                                         -12
                                                                  -13
                                                                                     4
## 3 Basal-like
                                         -12
                                                                  -13
                                                                                     5
## 4 Basal-like
                                         -12
                                                                  -13
                                                                                     5
## 5 Basal-like
                                           0
                                                                  -13
                                                                                     5
## 6 Basal-like
                                           0
                                                                                    5
                                                                  -13
     methylation.Clusters RPPA.Clusters CN.Clusters
## 1
                         5
                                   Basal
## 2
                         4
                                   Basal
## 3
                                   Basal
                                                     1
                         5
## 4
                         5
                                   Basal
                                                     1
## 5
                         5
                                   Basal
                         5
                                   Basal
                                                     1
##
     Integrated.Clusters..with.PAM50. Integrated.Clusters..no.exp.
## 1
                                      2
                                                                     2
                                      2
## 2
                                                                     1
                                      2
                                                                     2
## 3
                                      2
                                                                     2
## 4
## 5
                                      2
                                                                     2
                                                                     2
## 6
##
     Integrated.Clusters..unsup.exp.
## 1
## 2
                                     1
## 3
                                     2
## 4
                                     2
## 5
                                     2
## 6
```

tail(clinical)

```
##
       Complete.TCGA.ID Gender Age.at.Initial.Pathologic.Diagnosis ER.Status
## 100
           TCGA-BH-AOBZ FEMALE
                                                                 59 Positive
## 101
           TCGA-BH-AOC7 FEMALE
                                                                 48 Positive
## 102
           TCGA-BH-AODD
                          MALE
                                                                 58
                                                                     Positive
## 103
           TCGA-C8-A12U FEMALE
                                                                 46 Positive
## 104
           TCGA-C8-A12W FEMALE
                                                                 49 Positive
## 105
           TCGA-E2-A15A FEMALE
                                                                 45 Positive
##
       PR.Status HER2.Final.Status Tumor Tumor..T1.Coded Node Node.Coded
## 100 Positive
                          Negative
                                      Т3
                                                  T_{Other}
                                                            N1
                                                                 Positive
## 101 Negative
                                      T2
                                                  T Other
                          Positive
                                                            N1
                                                                 Positive
## 102 Positive
                          Positive
                                      T2
                                                  T_Other
                                                            N1
                                                                 Positive
```

```
## 103 Positive
                            Negative
                                        T2
                                                    T Other
                                                               N1
                                                                    Positive
## 104 Positive
                                        Т4
                                                    T_Other
                           Negative
                                                               N1
                                                                    Positive
## 105
                           Negative
       Positive
                                        T2
                                                    T Other
                                                               NЗ
                                                                    Positive
##
       Metastasis Metastasis.Coded AJCC.Stage Converted.Stage Survival.Data.Form
## 100
               MO
                            Negative Stage IIIA
                                                      Stage IIIA
                                                                           enrollment
## 101
               MO
                                                                           enrollment
                            Negative Stage IIB
                                                       Stage IIB
## 102
                                                                           enrollment
               MO
                            Negative
                                      Stage IIB
                                                       Stage IIB
## 103
                MO
                            Negative
                                       Stage IB
                                                       Stage IIB
                                                                           enrollment
## 104
                MO
                           Negative Stage IIIB
                                                      Stage IIIB
                                                                           enrollment
## 105
                МО
                            Negative Stage IIIC
                                                      Stage IIIC
                                                                           enrollment
       Vital.Status Days.to.Date.of.Last.Contact Days.to.date.of.Death OS.event
## 100
                                               1492
             LIVING
                                                                        NA
## 101
                                               1305
                                                                                   0
             LIVING
                                                                        NA
## 102
             LIVING
                                               1393
                                                                        NA
                                                                                   0
## 103
             LIVING
                                                  0
                                                                        NA
                                                                                   0
## 104
             LIVING
                                                  0
                                                                        NA
                                                                                   0
## 105
                                                502
             LIVING
                                                                        NA
       OS.Time PAM50.mRNA SigClust.Unsupervised.mRNA SigClust.Intrinsic.mRNA
          1492 Luminal B
## 100
                                                     -5
                                                                               -2
          1305 Luminal B
                                                                                0
## 101
                                                     -3
## 102
          1393 Luminal B
                                                     -3
                                                                               -6
## 103
             0 Luminal B
                                                     -5
                                                                               -2
             0 Luminal B
                                                     -5
## 104
                                                                               -2
## 105
           502 Luminal B
                                                     -5
                                                                               -2
##
       miRNA.Clusters methylation.Clusters RPPA.Clusters CN.Clusters
## 100
                     6
                                            4
                                                          Х
## 101
                     4
                                            4
                                                     LumA/B
                                                                        5
## 102
                     4
                                            4
                                                     LumA/B
                                                                        3
                     5
                                                                        5
## 103
                                                      Basal
## 104
                                                                        3
                     4
                                                     ReacII
## 105
                     4
                                                       Her2
                                                                        4
##
       Integrated.Clusters..with.PAM50. Integrated.Clusters..no.exp.
## 100
                                        4
## 101
                                        4
                                                                        1
## 102
                                        4
                                                                        1
## 103
                                        4
                                                                        1
## 104
                                                                        1
## 105
                                                                        1
       Integrated.Clusters..unsup.exp.
## 100
                                       1
## 101
                                       3
                                       3
## 102
## 103
                                       1
## 104
                                       1
                                       1
## 105
```

summary(clinical)

```
Complete.TCGA.ID
                           Gender
                                            Age.at.Initial.Pathologic.Diagnosis
##
    Length: 105
                        Length: 105
                                            Min.
                                                   :30.00
    Class : character
                        Class : character
                                            1st Qu.:49.00
##
    Mode :character
                        Mode : character
                                            Median :58.00
##
                                            Mean
                                                   :58.69
##
                                            3rd Qu.:67.00
```

```
##
                                           Max.
                                                  :88.00
##
                                                                 Tumor
##
     ER.Status
                        PR.Status
                                           HER2.Final.Status
                                                              Length: 105
##
   Length: 105
                       Length: 105
                                           Length:105
##
   Class : character
                       Class : character
                                           Class : character
                                                              Class : character
##
   Mode :character
                       Mode : character
                                           Mode :character
                                                              Mode : character
##
##
##
##
   Tumor..T1.Coded
                           Node
                                            Node.Coded
                                                               Metastasis
                       Length: 105
##
   Length: 105
                                           Length: 105
                                                              Length: 105
##
   Class : character
                       Class :character
                                           Class : character
                                                              Class : character
##
   Mode :character
                       Mode :character
                                           Mode :character
                                                              Mode :character
##
##
##
##
##
   Metastasis.Coded
                        AJCC.Stage
                                           Converted.Stage
                                                              Survival.Data.Form
##
   Length: 105
                       Length: 105
                                           Length: 105
                                                              Length: 105
##
   Class : character
                       Class : character
                                           Class : character
                                                              Class : character
   Mode :character
                       Mode :character
                                           Mode :character
                                                              Mode :character
##
##
##
##
##
   Vital.Status
                       Days.to.Date.of.Last.Contact Days.to.date.of.Death
   Length:105
                       Min.
                             : 0.0
                                                     Min.
                                                            : 160.0
##
                       1st Qu.: 240.0
                                                     1st Qu.: 947.5
##
   Class :character
                                                     Median :1364.0
   Mode :character
                       Median: 643.0
                             : 788.4
##
                       Mean
                                                     Mean
                                                            :1254.5
##
                       3rd Qu.:1288.0
                                                     3rd Qu.:1627.5
##
                       Max.
                              :2850.0
                                                     Max.
                                                            :2483.0
##
                                                     NA's
                                                            :94
                                        PAM50.mRNA
##
       OS.event
                        OS.Time
##
   Min.
           :0.0000
                           :
                                0.0
                                      Length: 105
                     Min.
   1st Qu.:0.0000
                     1st Qu.: 240.0
                                       Class : character
##
   Median :0.0000
                     Median : 665.0
                                      Mode :character
                     Mean : 817.6
##
   Mean
         :0.1048
   3rd Qu.:0.0000
##
                     3rd Qu.:1305.0
##
   Max. :1.0000
                     Max.
                            :2850.0
##
## SigClust.Unsupervised.mRNA SigClust.Intrinsic.mRNA miRNA.Clusters
## Min.
           :-12.000
                               Min.
                                       :-13.000
                                                        Min.
                                                               :1
  1st Qu.: -6.000
                                1st Qu.:-12.000
                                                        1st Qu.:3
## Median : -5.000
                               Median : -6.000
                                                        Median:4
   Mean : -4.886
                               Mean : -7.181
                                                        Mean
##
   3rd Qu.: -3.000
                                3rd Qu.: -2.000
                                                        3rd Qu.:5
##
  Max.
          : 0.000
                               Max.
                                     : 0.000
                                                        Max.
                                                               :7
##
   methylation.Clusters RPPA.Clusters
                                              CN.Clusters
## Min.
          :1.000
                         Length: 105
                                             Min.
                                                    :1.00
## 1st Qu.:2.000
                         Class : character
                                             1st Qu.:1.00
## Median: 4.000
                         Mode :character
                                             Median:3.00
```

```
##
    Mean
           :3.343
                                              Mean
                                                      :2.59
    3rd Qu.:4.000
                                              3rd Qu.:3.00
##
##
    Max.
           :5.000
                                              Max.
                                                      :5.00
##
##
    Integrated.Clusters..with.PAM50. Integrated.Clusters..no.exp.
##
   Min.
           :1.000
                                       Min.
                                              :1.000
    1st Qu.:2.000
                                       1st Qu.:1.000
   Median :3.000
                                       Median :2.000
##
##
    Mean
           :2.743
                                       Mean
                                              :1.981
##
    3rd Qu.:4.000
                                       3rd Qu.:3.000
##
    Max.
           :4.000
                                       Max.
                                              :4.000
##
##
    Integrated.Clusters..unsup.exp.
##
    Min.
           :1.000
##
    1st Qu.:1.000
##
    Median :2.000
##
    Mean
           :2.352
##
    3rd Qu.:3.000
   Max.
##
           :5.000
##
```

dim(clinical)

[1] 105 30

clinical\$PAM50.mRNA

```
"Basal-like"
##
     [1] "Basal-like"
                                           "Basal-like"
                                                            "Basal-like"
##
     [5] "Basal-like"
                          "Basal-like"
                                           "Basal-like"
                                                            "Basal-like"
##
     [9] "Basal-like"
                          "Basal-like"
                                           "Basal-like"
                                                            "Basal-like"
##
    [13] "Basal-like"
                          "Basal-like"
                                           "Basal-like"
                                                            "Basal-like"
##
    [17] "Basal-like"
                          "Basal-like"
                                           "Basal-like"
                                                            "Basal-like"
##
    [21] "Basal-like"
                          "Basal-like"
                                           "Basal-like"
                                                            "Basal-like"
##
   [25] "Basal-like"
                          "HER2-enriched" "HER2-enriched" "HER2-enriched"
##
   [29] "HER2-enriched"
                          "HER2-enriched"
                                                            "HER2-enriched"
                                           "HER2-enriched"
    [33] "HER2-enriched"
                          "HER2-enriched"
                                           "HER2-enriched" "HER2-enriched"
##
    [37] "HER2-enriched"
                         "HER2-enriched" "HER2-enriched" "HER2-enriched"
    [41] "HER2-enriched"
                          "HER2-enriched" "HER2-enriched" "Luminal A"
##
   [45] "Luminal A"
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
    [49] "Luminal A"
##
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
##
   [53] "Luminal A"
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
   [57] "Luminal A"
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
##
    [61] "Luminal A"
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
##
    [65] "Luminal A"
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
##
    [69] "Luminal A"
                          "Luminal A"
                                           "Luminal A"
                                                            "Luminal A"
##
   [73] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
##
    [77] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
   [81] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
##
##
   [85] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
   [89] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
##
##
    [93] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
   [97] "Luminal B"
##
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
## [101] "Luminal B"
                          "Luminal B"
                                           "Luminal B"
                                                            "Luminal B"
## [105] "Luminal B"
```

str(clinical)

```
## 'data.frame':
                   105 obs. of 30 variables:
  $ Complete.TCGA.ID
                                              "TCGA-A2-A0T2" "TCGA-A2-A0CM" "TCGA-BH-A18V" "TCGA-BH-A
                                       : chr
                                              "FEMALE" "FEMALE" "FEMALE" ...
## $ Gender
                                       : chr
   $ Age.at.Initial.Pathologic.Diagnosis: int
                                              66 40 48 56 38 57 74 60 61 67 ...
                                              "Negative" "Negative" "Negative" "Negative" ...
## $ ER.Status
                                       : chr
## $ PR.Status
                                              "Negative" "Negative" "Negative" "Negative" ...
                                       : chr
                                              "Negative" "Negative" "Negative" ...
## $ HER2.Final.Status
                                       : chr
                                              "T3" "T2" "T2" "T2" ...
##
   $ Tumor
                                       : chr
## $ Tumor..T1.Coded
                                              "T_Other" "T_Other" "T_Other" "T_Other" ...
                                       : chr
                                              "N3" "NO" "N1" "N1" ...
## $ Node
                                       : chr
## $ Node.Coded
                                       : chr
                                              "Positive" "Negative" "Positive" "Positive" ...
## $ Metastasis
                                              "M1" "MO" "MO" "MO" ...
                                       : chr
## $ Metastasis.Coded
                                      : chr
                                              "Positive" "Negative" "Negative" "Negative" ...
## $ AJCC.Stage
                                              "Stage IV" "Stage IIA" "Stage IIB" "Stage IIB" ...
                                       : chr
                                              "No_Conversion" "Stage IIA" "No_Conversion" "No_Convers
   $ Converted.Stage
                                       : chr
##
                                              "followup" "followup" "enrollment" "enrollment" ...
## $ Survival.Data.Form
                                       : chr
## $ Vital.Status
                                              "DECEASED" "DECEASED" "DECEASED" ...
                                       : chr
                                              240 754 1555 1692 133 309 425 643 775 964 ...
## $ Days.to.Date.of.Last.Contact
                                       : int
## $ Days.to.date.of.Death
                                              240 754 1555 1692 NA NA NA NA NA NA ...
                                       : int
## $ OS.event
                                       : int 1 1 1 1 0 0 0 0 0 0 ...
## $ OS.Time
                                              240 754 1555 1692 133 309 425 643 775 964 ...
                                       : int
                                              "Basal-like" "Basal-like" "Basal-like" ...
##
   $ PAM50.mRNA
                                       : chr
   $ SigClust.Unsupervised.mRNA
                                       : int 0 -12 -12 -12 0 0 0 -12 -12 -12 ...
##
## $ SigClust.Intrinsic.mRNA
                                             -13 -13 -13 -13 -13 -13 -13 -13 -13 ...
                                       : int
## $ miRNA.Clusters
                                       : int 3 4 5 5 5 5 3 5 2 5 ...
                                       : int 5 4 5 5 5 5 5 5 5 5 ...
## $ methylation.Clusters
## $ RPPA.Clusters
                                       : chr "Basal" "Basal" "Basal" ...
## $ CN.Clusters
                                      : int 3 4 1 1 1 1 1 1 3 ...
## $ Integrated.Clusters..with.PAM50. : int 2 2 2 2 2 2 2 2 2 2 ...
                                       : int 2 1 2 2 2 2 2 2 2 2 ...
   $ Integrated.Clusters..no.exp.
   $ Integrated.Clusters..unsup.exp.
                                       : int 2 1 2 2 2 2 2 2 2 2 ...
```

summary(clinical)

```
Age.at.Initial.Pathologic.Diagnosis
## Complete.TCGA.ID
                        Gender
## Length:105
                                        Min. :30.00
                      Length: 105
                                        1st Qu.:49.00
## Class :character Class :character
## Mode :character Mode :character
                                        Median :58.00
##
                                        Mean :58.69
##
                                        3rd Qu.:67.00
##
                                        Max.
                                               :88.00
##
##
    ER.Status
                      PR.Status
                                        HER2.Final.Status
                                                             Tumor
## Length:105
                      Length: 105
                                        Length: 105
                                                          Length: 105
## Class :character
                      Class :character
                                        Class :character
                                                          Class : character
  Mode :character Mode :character
                                        Mode :character
                                                          Mode :character
##
##
##
##
##
   Tumor..T1.Coded
                         Node
                                         Node.Coded
                                                           Metastasis
```

```
Length: 105
                       Length: 105
                                          Length: 105
                                                             Length: 105
   Class : character
                       Class : character
                                          Class : character
                                                             Class : character
                                          Mode :character
##
   Mode :character
                       Mode :character
                                                             Mode :character
##
##
##
##
##
  Metastasis.Coded
                        AJCC.Stage
                                          Converted.Stage
                                                             Survival.Data.Form
##
   Length:105
                       Length: 105
                                          Length:105
                                                             Length: 105
##
                       Class : character
                                          Class : character
                                                             Class : character
   Class : character
   Mode :character
                       Mode :character
                                          Mode : character
                                                             Mode : character
##
##
##
##
##
   Vital.Status
                       Days.to.Date.of.Last.Contact Days.to.date.of.Death
##
   Length: 105
                       Min. : 0.0
                                                    Min.
                                                           : 160.0
   Class :character
                       1st Qu.: 240.0
                                                    1st Qu.: 947.5
##
   Mode :character
                       Median : 643.0
                                                    Median: 1364.0
                       Mean : 788.4
##
                                                    Mean
                                                           :1254.5
##
                       3rd Qu.:1288.0
                                                    3rd Qu.:1627.5
##
                       Max.
                              :2850.0
                                                    Max.
                                                           :2483.0
                                                    NA's
##
                                                           :94
##
       OS.event
                        OS.Time
                                       PAM50.mRNA
##
           :0.0000
                           :
                                      Length: 105
   Min.
                     Min.
                                0.0
   1st Qu.:0.0000
                     1st Qu.: 240.0
                                      Class : character
##
   Median :0.0000
                     Median : 665.0
                                      Mode :character
   Mean
           :0.1048
                            : 817.6
                     Mean
##
   3rd Qu.:0.0000
                     3rd Qu.:1305.0
##
  Max.
           :1.0000
                     Max.
                            :2850.0
##
   SigClust.Unsupervised.mRNA SigClust.Intrinsic.mRNA miRNA.Clusters
## Min.
          :-12.000
                               Min.
                                     :-13.000
                                                       Min.
##
  1st Qu.: -6.000
                               1st Qu.:-12.000
                                                       1st Qu.:3
   Median : -5.000
                               Median : -6.000
##
                                                       Median:4
   Mean
          : -4.886
                               Mean
                                     : -7.181
                                                       Mean
##
   3rd Qu.: -3.000
                               3rd Qu.: -2.000
                                                       3rd Qu.:5
##
  Max. : 0.000
                               Max.
                                     : 0.000
                                                       Max.
                                                              :7
##
   methylation.Clusters RPPA.Clusters
##
                                             CN.Clusters
   Min. :1.000
                         Length: 105
                                            Min. :1.00
##
   1st Qu.:2.000
                         Class :character
                                            1st Qu.:1.00
  Median :4.000
                                            Median:3.00
                         Mode :character
## Mean
          :3.343
                                            Mean
                                                   :2.59
   3rd Qu.:4.000
                                            3rd Qu.:3.00
                                                   :5.00
## Max.
          :5.000
                                            Max.
##
##
  Integrated.Clusters..with.PAM50. Integrated.Clusters..no.exp.
## Min.
          :1.000
                                     Min.
                                            :1.000
## 1st Qu.:2.000
                                     1st Qu.:1.000
## Median :3.000
                                     Median :2.000
## Mean
         :2.743
                                     Mean
                                           :1.981
## 3rd Qu.:4.000
                                     3rd Qu.:3.000
## Max.
         :4.000
                                     Max.
                                            :4.000
```

```
##
##
    Integrated.Clusters..unsup.exp.
##
           :1.000
   1st Qu.:1.000
##
##
   Median :2.000
##
  Mean
           :2.352
   3rd Qu.:3.000
  {\tt Max.}
##
            :5.000
##
```

Data import - 77_cancer_proteomes_CPTAC_itraq.csv

The following datset has information about the protein expression data

proteomes <- read.csv("/Users/sheetalnighut/Desktop/cancer-project/77_cancer_proteomes_CPTAC_itraq.csv"
12553 observations(proteins) and 86 columns
head(proteomes)</pre>

```
RefSeq_accession_number gene_symbol
                                                    gene name AO.A12D.O1TCGA
## 1
                    NP 958782
                                     PLEC plectin isoform 1
                                                                     1.096131
## 2
                    NP 958785
                                     <NA> plectin isoform 1g
                                                                     1.111370
## 3
                    NP_958786
                                     PLEC plectin isoform 1a
                                                                     1.111370
## 4
                    NP_000436
                                      <NA> plectin isoform 1c
                                                                     1.107561
## 5
                    NP_958781
                                      <NA> plectin isoform 1e
                                                                     1.115180
## 6
                    NP_958780
                                     PLEC plectin isoform 1f
                                                                     1.107561
##
     C8.A131.01TCGA AO.A12B.01TCGA BH.A18Q.02TCGA C8.A130.02TCGA C8.A138.03TCGA
## 1
           2.609943
                         -0.6598280
                                          0.1953407
                                                        -0.4940596
                                                                          2.765081
## 2
           2.650422
                         -0.6487422
                                                        -0.5038992
                                         0.2154129
                                                                          2.779709
## 3
           2.650422
                         -0.6542851
                                         0.2154129
                                                        -0.5006193
                                                                          2.779709
## 4
           2.646374
                         -0.6321133
                                         0.2053768
                                                        -0.5104589
                                                                          2.797995
## 5
           2.646374
                         -0.6404277
                                         0.2154129
                                                        -0.5038992
                                                                          2.787023
## 6
                         -0.6542851
                                                        -0.5038992
           2.646374
                                         0.2154129
                                                                          2.779709
     E2.A154.O3TCGA C8.A12L.O4TCGA A2.A0EX.O4TCGA AO.A12D.O5TCGA AN.A04A.O5TCGA
##
                                                           1.100688
## 1
          0.8626593
                           1.407570
                                           1.185108
                                                                         0.3845877
## 2
          0.8701860
                           1.407570
                                           1.192612
                                                          1.100688
                                                                         0.3713928
## 3
          0.8701860
                           1.410312
                                           1.188860
                                                          1.100688
                                                                         0.3713928
          0.8664226
                           1.407570
                                           1.185108
                                                           1.100688
                                                                         0.3779903
## 5
          0.8701860
                           1.413053
                                           1.200116
                                                           1.093358
                                                                         0.3746916
## 6
                                                           1.097023
          0.8701860
                           1.407570
                                           1.188860
                                                                         0.3779903
##
     BH.AOAV.O5TCGA C8.A12T.O6TCGA A8.AO6Z.O7TCGA A2.AOCM.O7TCGA BH.A18U.O8TCGA
## 1
          0.3505357
                         -0.2049179
                                        -0.4964091
                                                                        -0.2650304
                                                          0.6834035
## 2
          0.3674053
                         -0.1624185
                                         -0.4985089
                                                          0.6944241
                                                                        -0.2516423
## 3
          0.3674053
                         -0.1666684
                                        -0.4964091
                                                          0.6980976
                                                                        -0.2516423
## 4
          0.3606575
                         -0.1836682
                                         -0.4922095
                                                          0.6870771
                                                                        -0.2516423
## 5
          0.3707793
                         -0.1666684
                                         -0.4880099
                                                          0.6870771
                                                                        -0.2516423
## 6
          0.3674053
                         -0.1666684
                                         -0.4964091
                                                          0.6980976
                                                                        -0.2516423
##
     A2.A0EQ.08TCGA AR.A0U4.09TCGA AD.A0J9.10TCGA AR.A1AP.11TCGA AN.A0FK.11TCGA
                        -0.03322133
## 1
         -0.9126703
                                        0.020007050
                                                          0.4610875
                                                                         0.9735642
         -0.9279787
                        -0.03021642
## 2
                                        0.011955318
                                                          0.4610875
                                                                         0.9774761
## 3
         -0.9279787
                        -0.02721152
                                        0.011955318
                                                          0.4610875
                                                                         0.9774761
## 4
         -0.9318057
                        -0.03021642
                                        0.003903587
                                                         0.4610875
                                                                         0.9696523
## 5
         -0.9279787
                        -0.03021642
                                        0.011955318
                                                         0.4610875
                                                                         0.9852998
## 6
         -0.9279787
                        -0.03021642
                                       0.011955318
                                                         0.4610875
                                                                         0.9774761
```

```
AO.AOJ6.11TCGA A7.A13F.12TCGA BH.AOE1.12TCGA A7.AOCE.13TCGA A2.AOYC.13TCGA
## 1
          0.8311317
                           1.279185
                                          0.7620444
                                                          -1.123173
                                                                          0.8188241
## 2
          0.8565398
                           1.275167
                                          0.7620444
                                                          -1.123173
                                                                          0.8148772
## 3
                                          0.7663844
          0.8565398
                           1.275167
                                                          -1.116861
                                                                          0.8148772
## 4
          0.8367780
                           1.279185
                                          0.7577045
                                                          -1.129486
                                                                          0.7990900
                           1.279185
## 5
          0.8650092
                                          0.7663844
                                                          -1.129486
                                                                          0.8188241
## 6
          0.8565398
                           1.279185
                                          0.7620444
                                                          -1.120017
                                                                          0.8148772
##
     AO.AOJC.14TCGA A8.AO8Z.14TCGA AR.AOTX.14TCGA A8.AO76.15TCGA AO.A126.15TCGA
## 1
         -0.3072668
                          0.5688946
                                         -0.5834286
                                                           1.873982
                                                                          0.1958767
## 2
         -0.3072668
                          0.5688946
                                         -0.5725489
                                                           1.870383
                                                                          0.1958767
## 3
         -0.3072668
                          0.5688946
                                         -0.5671090
                                                           1.870383
                                                                          0.1958767
## 4
         -0.3072668
                          0.5688946
                                         -0.5834286
                                                           1.859587
                                                                          0.2189346
## 5
         -0.3010327
                          0.5688946
                                         -0.5725489
                                                           1.870383
                                                                          0.1997197
                                                                          0.1997197
## 6
         -0.3072668
                          0.5688946
                                         -0.5779888
                                                           1.870383
##
     BH.AOC1.16TCGA A2.AOEY.16TCGA AR.A1AW.17TCGA AR.A1AV.17TCGA C8.A135.17TCGA
## 1
         -0.5183665
                           1.174881
                                          0.5783087
                                                         -0.7598231
                                                                           1.120502
## 2
         -0.5100020
                                          0.5822129
                                                         -0.7598231
                           1.183209
                                                                           1.137618
## 3
                                          0.5783087
                                                         -0.7491137
         -0.5072138
                           1.183209
                                                                           1.137618
## 4
                                          0.5900212
                                                         -0.7357270
                                                                           1.137618
         -0.5183665
                           1.174881
## 5
         -0.5127902
                           1.179045
                                          0.5861170
                                                         -0.7491137
                                                                           1.120502
## 6
         -0.5072138
                           1.183209
                                          0.5783087
                                                         -0.7437590
                                                                           1.127348
     A2.A0EV.18TCGA AN.A0AM.18TCGA D8.A142.18TCGA AN.A0FL.19TCGA BH.A0DG.19TCGA
##
          0.4529859
## 1
                           1.501967
                                          0.5385958
                                                                         -0.2056375
                                                           2.455138
## 2
          0.4725901
                           1.510348
                                          0.5422105
                                                           2.480137
                                                                         -0.2056375
## 3
          0.4725901
                           1.501967
                                          0.5422105
                                                           2.480137
                                                                         -0.2056375
## 4
          0.4585871
                           1.501967
                                          0.5349810
                                                           2.461956
                                                                         -0.2150062
## 5
                                                                         -0.2056375
          0.4725901
                           1.501967
                                          0.5422105
                                                           2.477864
## 6
          0.4725901
                           1.510348
                                          0.5422105
                                                           2.471046
                                                                         -0.2103218
     AR.AOTV.2OTCGA C8.A12Z.2OTCGA AO.AOJJ.2OTCGA AO.AOJE.21TCGA AN.AOAJ.21TCGA
##
## 1
                         -0.7871950
                                          0.7571881
                                                                         -0.4281815
          -1.514278
                                                          0.5597770
## 2
          -1.528285
                         -0.7559406
                                          0.7808707
                                                          0.5634069
                                                                         -0.4063780
## 3
          -1.528285
                         -0.7559406
                                          0.7741042
                                                          0.5597770
                                                                         -0.4063780
## 4
          -1.531087
                         -0.7746932
                                          0.7639546
                                                          0.5416274
                                                                         -0.4063780
## 5
          -1.514278
                         -0.7715678
                                          0.7707210
                                                          0.5597770
                                                                         -0.4063780
## 6
                                          0.7774874
                                                          0.5597770
                                                                         -0.4063780
          -1.525484
                         -0.7715678
     A7.A0CJ.22TCGA A0.A12F.22TCGA A8.A079.23TCGA A2.A0T3.24TCGA A2.A0YD.24TCGA
##
## 1
         -1.0012398
                          -1.947792
                                           1.048959
                                                          0.5837133
                                                                         0.06377853
## 2
         -1.0046198
                          -1.952718
                                           1.052257
                                                          0.5806231
                                                                         0.09333637
## 3
         -1.0046198
                          -1.955180
                                           1.052257
                                                          0.5806231
                                                                         0.08446902
                          -1.947792
                                           1.058852
## 4
         -0.9978599
                                                          0.5868034
                                                                         0.06673431
## 5
         -1.0012398
                          -1.957643
                                           1.052257
                                                          0.5868034
                                                                         0.08446902
## 6
         -1.0012398
                          -1.955180
                                           1.052257
                                                          0.5868034
                                                                         0.09333637
##
     AR.AOTR.25TCGA AO.AO30.25TCGA AO.A12E.26TCGA A8.AO6N.26TCGA A2.AOYG.27TCGA
## 1
          -1.101675
                           1.053225
                                          0.2648591
                                                          0.2385471
                                                                        -0.07820182
## 2
          -1.108783
                           1.055948
                                          0.2757113
                                                          0.2498182
                                                                        -0.06805814
## 3
          -1.108783
                           1.055948
                                          0.2757113
                                                          0.2441826
                                                                        -0.07143937
                           1.058671
## 4
          -1.096937
                                          0.2784244
                                                          0.2498182
                                                                        -0.05791445
## 5
          -1.111152
                           1.058671
                                          0.2784244
                                                          0.2498182
                                                                        -0.06467691
## 6
          -1.106413
                           1.055948
                                          0.2729983
                                                          0.2498182
                                                                        -0.06805814
##
     BH.A18N.27TCGA AN.AOAL.28TCGA A2.AOT6.29TCGA E2.A158.29TCGA E2.A15A.29TCGA
## 1
                                                          -1.086529
                          0.3236627
                                          0.7939756
                                                                           2.180123
           1.101261
## 2
           1.101261
                          0.3269726
                                          0.8181815
                                                          -1.095492
                                                                           2.180123
## 3
           1.097767
                          0.3269726
                                          0.8147235
                                                          -1.095492
                                                                           2.180123
## 4
           1.090779
                          0.3302826
                                          0.8008915
                                                          -1.095492
                                                                           2.180123
```

```
## 5
           1.108248
                          0.3269726
                                           0.8181815
                                                           -1.095492
                                                                            2.180123
## 6
                                                           -1.093252
            1.101261
                          0.3269726
                                           0.8112655
                                                                            2.180123
##
     AO.AOJM.3OTCGA C8.A12V.3OTCGA A2.AOD2.31TCGA C8.A12U.31TCGA AR.A1AS.31TCGA
## 1
            1.395247
                                                          -0.4815502
                                                                            1.222507
                          0.6739047
                                         0.10749090
##
           1.408922
                          0.6887176
                                         0.10416449
                                                          -0.4778898
                                                                            1.218974
## 3
           1.412341
                          0.6887176
                                         0.10749090
                                                         -0.4815502
                                                                            1.222507
## 4
           1.408922
                          0.6776079
                                         0.09751166
                                                          -0.4705692
                                                                            1.204839
           1.408922
## 5
                          0.6887176
                                         0.10416449
                                                          -0.4815502
                                                                            1.222507
## 6
            1.412341
                          0.6887176
                                         0.10416449
                                                          -0.4852105
                                                                            1.218974
##
     A8.A09G.32TCGA C8.A131.32TCGA C8.A134.32TCGA A2.A0YF.33TCGA BH.A0DD.33TCGA
## 1
          -1.523343
                           2.707250
                                          0.1401818
                                                           0.3113192
                                                                          -0.6923158
## 2
          -1.512646
                           2.733832
                                          0.1260538
                                                           0.2961771
                                                                          -0.6594687
                                                                          -0.6641611
## 3
          -1.509972
                           2.737629
                                          0.1331178
                                                           0.2961771
## 4
          -1.517995
                           2.733832
                                          0.1119257
                                                           0.2961771
                                                                          -0.6571224
## 5
          -1.509972
                           2.752819
                                          0.1260538
                                                           0.2961771
                                                                          -0.6618149
## 6
          -1.512646
                            2.737629
                                           0.1260538
                                                           0.2961771
                                                                          -0.6618149
##
     BH.AOE9.33TCGA AR.AOTT.34TCGA AO.A12B.34TCGA A2.AOSW.35TCGA AO.AOJL.35TCGA
                         -0.5114212
## 1
           1.466665
                                         -0.9639039
                                                         -0.4877725
                                                                            -0.10668
## 2
           1.482283
                         -0.5260667
                                         -0.9382095
                                                          -0.4877725
                                                                            -0.10668
## 3
           1.474474
                         -0.5260667
                                          -0.9439194
                                                          -0.4877725
                                                                            -0.10668
## 4
           1.458856
                         -0.5333894
                                         -0.9353546
                                                          -0.4877725
                                                                            -0.10668
## 5
                                         -0.9353546
                                                          -0.5038532
           1.474474
                         -0.5297281
                                                                            -0.10668
## 6
            1.474474
                         -0.5297281
                                          -0.9382095
                                                          -0.4877725
                                                                            -0.10668
##
     BH.AOBV.35TCGA A2.AOYM.36TCGA BH.AOC7.36TCGA A2.AOSX.36TCGA X263d3f.I.CPTAC
        -0.06583842
## 1
                          0.6558497
                                         -0.5522120
                                                          -0.3985598
                                                                            0.5985845
## 2
        -0.05589267
                          0.6581426
                                         -0.5477494
                                                          -0.3926014
                                                                            0.6066975
## 3
        -0.06583842
                                         -0.5522120
                                                          -0.3926014
                          0.6558497
                                                                            0.6039931
##
        -0.05589267
                          0.6558497
                                         -0.5522120
                                                         -0.3926014
                                                                            0.6039931
## 5
        -0.06252317
                          0.6512639
                                         -0.5566746
                                                         -0.3955806
                                                                            0.6039931
        -0.05589267
## 6
                          0.6581426
                                         -0.5477494
                                                          -0.3926014
                                                                            0.6066975
##
     blcdb9.I.CPTAC c4155b.C.CPTAC
## 1
         -0.1912845
                          0.5669753
## 2
         -0.1839177
                          0.5787017
## 3
         -0.1860225
                          0.5767473
## 4
         -0.1860225
                          0.5767473
## 5
         -0.1670792
                          0.5767473
## 6
         -0.1839177
                          0.5787017
```

tail(proteomes)

```
##
         RefSeq_accession_number gene_symbol
## 12548
                        NP_997203
                                       OTUD6A
## 12549
                    NP_001191293
                                          <NA>
## 12550
                        NP_775791
                                          <NA>
## 12551
                        NP_004065
                                         COX8A
## 12552
                                         MIIP
                        NP_068752
##
   12553
                        NP_219494
                                     KIAA1737
##
                                                gene_name AO.A12D.01TCGA
                        OTU domain-containing protein 6A
## 12548
                                                                       NA
## 12549
                                protein FAM24B precursor
                                                                       NA
## 12550
               putative uncharacterized protein C9orf62
                                                                       NA
## 12551 cytochrome c oxidase subunit 8A, mitochondrial
                                                                       NΑ
              migration and invasion-inhibitory protein
## 12552
                                                               -0.6335172
## 12553
                        uncharacterized protein KIAA1737
                                                               12.6664882
```

##		C8.A131.O1TCGA	AO.A12B.O1TCGA	BH.A18Q.O2TCGA	C8.A130.O2TCGA
##	12548	NA	NA	-8.111243	
##	12549	NA	NA	-16.029761	1.72969151
##	12550	NA	NA	-2.046065	-0.42518234
##	12551	NA	NA	-1.778435	-0.14967335
##	12552	4.8403254	-1.965192	NA	NA
##	12553	0.1407356	-2.854835	-3.069752	-0.04799742
##		C8.A138.O3TCGA	E2.A154.O3TCGA	C8.A12L.O4TCGA	A2.AOEX.O4TCGA
##	12548	4.707022	-4.733495	NA	NA
##	12549	4.107251	-9.584499	-5.196859	-6.101005
##	12550	-3.203370	-4.786183	NA	NA
##	12551	1.971481	-3.103949	-0.933726	-1.726336
##	12552	NA	NA	NA	NA
##	12553	NA	NA	NA	NA
##		AO.A12D.O5TCGA	AN.AO4A.O5TCGA	BH.AOAV.O5TCGA	C8.A12T.O6TCGA
##	12548	NA	NA	NA	NA
##	12549	-2.5788279	0.9024874	-7.011385	-11.02102
##	12550	NA	NA	NA	NA
##	12551	1.2949255	1.7370646	-1.393788	NA
##	12552	-0.1893414	0.3614967	-3.057136	NA
##	12553	13.0664447	0.1437809	NA	NA
##		A8.A06Z.07TCGA	A2.AOCM.O7TCGA	BH.A18U.O8TCGA	A2.A0EQ.O8TCGA
##	12548	NA	NA	-11.55786	-6.373934
##	12549	NA	NA	-12.62890	-1.123160
##	12550	NA	NA	NA	NA
##	12551	NA	NA	NA	NA
##	12552	NA	NA	NA	NA
##	12553	NA	NA	NA	NA
##		AR.AOU4.O9TCGA	AO.AOJ9.10TCGA	AR.A1AP.11TCGA	AN.AOFK.11TCGA
##	12548	NA	NA	-1.073848	-3.059596
##	12549	NA	3.4097859	NA	NA
##	12550	NA	1.7632069	NA	NA
##	12551	NA	-0.3382950	NA	NA
##	12552	NA	2.7012335	NA	NA
##	12553	NA	0.6560938	-1.177280	-3.266926
##		AO.AOJ6.11TCGA	A7.A13F.12TCGA	BH.AOE1.12TCGA	A7.AOCE.13TCGA
##	12548	-3.231339	0.9698297	-7.609707	NA
##	12549	NA	NA	NA	NA
##	12550	NA	NA	NA	-1.306238
##	12551	NA	NA	NA	NA
##	12552	NA	-1.3202008	-2.006840	NA
##	12553	-3.753616	NA	NA	NA
##		A2.AOYC.13TCGA	AO.AOJC.14TCGA	A8.A08Z.14TCGA	AR.AOTX.14TCGA
##	12548	NA	NA	NA	NA
##	12549	NA	NA	NA	NA
##	12550	4.509094	2.6071730	0.48649433	-3.542726
##	12551	NA	NA	NA	NA
##	12552	NA	0.2912064	-0.05448119	-2.136516
##	12553	NA	0.5281020	-3.43647384	-10.008030
##			AO.A126.15TCGA		
##	12548	NA	NA	-9.512991	4.597606
	12549	NA	NA	-6.217378	6.179888
	12550	NA	NA	NA	NA
	12551	NA	NA	NA	NA

```
0.6720037 -6.0643526 NA
5.1056477 0.5648036 NA
## 12552
## 12553
## AR.A1AW.17TCGA AR.A1AV.17TCGA C8.A135.17TCGA A2.A0EV.18TCGA
         -1.201996 2.7849772 -11.185872 -12.2785457
## 12548
          1.468461 -4.1439828
-1.475288 -0.5724091
                                          -10.3377293
## 12549
                                -13.630031
## 12550
                                 -3.702775
                                            -0.6532514
                      NA
## 12551
            NA
                                   NA
          NA NA NA
1.113181 -5.3675279 -8.953951
## 12552
## 12553
## AN.AOAM.18TCGA D8.A142.18TCGA AN.AOFL.19TCGA BH.AODG.19TCGA
## 12548 -5.42108597 -12.337110 NA
         -1.27228306
## 12549
                      -9.546530
                                 -3.012765
                                              -0.753707
## 12550
         0.03521846
                      -4.066584
                                      NA
                                                   NA
## 12551
           NA
                        NA
## 12552
                NA
                           NA
                                       NΑ
         NA
## 12553
                           NA
                                       NA
## AR.AOTV.2OTCGA C8.A12Z.2OTCGA AO.AOJJ.2OTCGA AO.AOJE.21TCGA
## 12548 NA NA NA NA
## 12549
          -2.116583
                      -6.703657
                                 0.8147029
                                                   NΑ
                       NA
                                  NA
## 12550
            NA
                                             -1.494749
## 12551
           2.040722
                      -4.375203
                                 -0.2916098
## 12552
            NA
                       NA
## 12553
               NΑ
                           NA
                                      NΑ
## AN.AOAJ.21TCGA A7.AOCJ.22TCGA A0.A12F.22TCGA A8.AO79.23TCGA
         NA
                     NA
## 12548
                                  NA
## 12549
               NA
                           NA
                                      NA
## 12550
          -3.017946
                           NA
                                       NA
                                                   NΑ
## 12551
            NA
                            NA
                                       NA
                                             -1.351759
## 12552
                NA
                            NA
                                              2.875880
                                       NA
## 12553
               NA
                           NA
                                       NA
## A2.AOT3.24TCGA A2.AOYD.24TCGA AR.AOTR.25TCGA AO.AO30.25TCGA
## 12548 NA NA
                                       NA
## 12549
          -12.265010
                      -7.677420
                                  1.475930
                                             -0.5997658
           NA
                      NA
## 12550
                                  0.528281
                                             0.5276447
                                             0.7536714
## 12551
          -1.264179
                      -4.801442
                                  1.492514
## 12552
          -1.100403
                      -2.590516
                                  -4.098615
                                             -5.0113722
                                  NA
          -5.590348
                     -6.740437
## AO.A12E.26TCGA A8.A06N.26TCGA A2.A0YG.27TCGA BH.A18N.27TCGA
## 12548 NA NA -10.245556 -9.481603
                     -0.02350419
## 12549
           1.716341
                                  NA
                                                  NA
## 12550
            NA
                                 -1.254869
                                              4.036092
                      NA
## 12551
                NΑ
                           NA
                                                  NΑ
                                      NΑ
## 12552
                NA
                            NA
                                       NA
               NΑ
                           NA
                                       NΑ
## AN.AOAL.28TCGA A2.AOT6.29TCGA E2.A158.29TCGA E2.A15A.29TCGA
         NA NA NA
## 12548
                     -13.120988
           4.633196
                                 0.3789715
                                              -3.863634
## 12549
## 12550
            NA
                       NA
                                               NA
                                  NA
                      NA NA
1.181271 0.6657973
NA NA
## 12551
                NA
                                                   NΑ
           NA
NA
## 12552
                                              4.072432
## AO.AOJM.3OTCGA C8.A12V.3OTCGA A2.AOD2.31TCGA C8.A12U.31TCGA
## 12548 -10.289998 1.4960184 -8.324969 -2.1140524
          -9.920774 3.8697791
## 12549
                                 -4.679219
                                             1.0557838
```

##	12550	NA	NA	NA	NA
##	12551	-2.375620	-0.3370729	-1.106650	1.0265012
##	12552	NA	NA	1.100000 NA	1.0200012 NA
##	12553	NA NA	NA NA	-6.941181	0.7446567
##	12000		A8.A09G.32TCGA		
##	12548	4.7986122	NA	NA	NA
##	12549	3.0529642	-9.744192	-2.130632	0.5392989
##	12550	NA	NA	NA	NA
##	12551	1.5405403	NA	NA	NA
##	12552	NA	-4.034527	2.027516	-1.2796861
##	12553	0.9327439	NA	NA	NA
##			BH.AODD.33TCGA	BH.AOE9.33TCGA	AR.AOTT.34TCGA
##	12548	1.412905	-5.987738	-8.482188	NA
##	12549	NA	NA	NA	-2.576435
##	12550	NA	NA	NA	NA
##	12551	NA	NA	NA	NA
##	12552	NA	NA	NA	NA
##	12553	NA	NA	NA	-1.983293
##		${\tt AO.A12B.34TCGA}$	A2.AOSW.35TCGA	AO.AOJL.35TCGA	BH.AOBV.35TCGA
##	12548	NA	NA	NA	NA
##	12549	-6.66235	NA	NA	NA
##	12550	NA	NA	NA	NA
##	12551	NA	NA	NA	NA
##	12552	NA	NA	NA	NA
##	12553	-6.00286	NA	NA	NA
##		A2.AOYM.36TCGA	BH.AOC7.36TCGA	A2.AOSX.36TCGA	X263d3f.I.CPTAC
##	12548	NA	NA	NA	NA
##	12549	NA	NA	NA	-8.02007140
##	12550	NA	NA	NA	0.04960831
##	12551	NA	NA	NA	NA
##	12552	NA	NA	NA	
##	12553	NA	NA	NA	NA
##		blcdb9.I.CPTAC	c4155b.C.CPTAC		
##	12548	NA	NA		
##	12549	-3.0938223	-4.6024175		
##	12550	-0.6469766	0.2405902		
##	12551	NA	NA		
##	12552	-1.7183267	-0.3691832		
##	12553	NA	NA		

dim(proteomes)

[1] 12553 86

Combining the datasets

- $\bullet\,$ Two data sets need to be combined before analysis
- using the cbind()
- new column was formed in each data set
- the two data sets were connected based on this new column

```
# Transposing the proteome matrix will result in rows rather than columns of observations.
## save rownames
# RefSeq_accession_number : ID of proteins
n <- proteomes$RefSeq_accession_number
#Transpose all columns except the first three
proteomes <- as.data.frame(t(proteomes[,4:86]))
colnames(proteomes) <- n
#Row names in the first column,
proteomes <- cbind(rownames(proteomes), data.frame(proteomes, row.names=NULL))
colnames(proteomes)[1] <- "Complete.TCGA.ID"</pre>
```

Manipulating the format

working with the Patient IDs: - Unfortunately, the patient IDs in the clinical dataset and the proteomic data set have different formats. - To enable combining of the two data sets on this variable, this piece of code reformats the id in the clinical data set. - defined the code which does the job - used sapply() to implement the function defined

```
# To enable the combining of data sets, Complete.TCGA.ID is being reorganized into a clinical format.
# Defining the restructuring formula:

get.clinical.id <- function(proteome.id) {
    x = substr(proteome.id, 4, 7)
    y = substr(proteome.id, 0, 2)
    paste("TCGA",y,x,sep="-")
}

#sapply to proteomes' id column
proteomes$Complete.TCGA.ID <- sapply(proteomes$Complete.TCGA.ID, get.clinical.id)
proteomes_all <- proteomes</pre>
```

I have used plot intro() from the DataExplorer package to check if the missing values are present. There are 10% of missing values in the dataset. Discarded variables with more than 25% of the data missing because they wouldn't be significant for further research. Used the mean imputation method, imputed the NA values with the mean of a specific column for the remaining variables with missing data.

Exploring the dataset merging and manipulating the datasets

After merging and manipulating the datasets retrieved - The proteome data set is the final one and to be used in the further analysis using the head str dim glimpse summary functions, the dimensions of the data can be observed

```
#head(proteomes)
#dim(proteomes)
#tail(proteomes)
#summary(proteomes)
#glimpse(proteomes)
#str(proteomes)
```

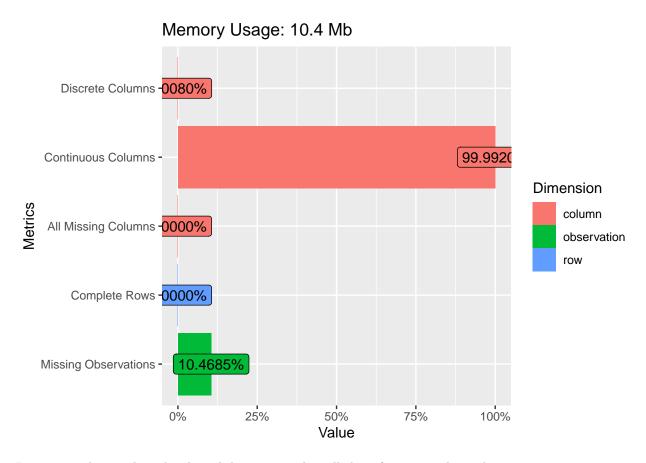
Data Exploration

- 1. exploratory data plots plot_intro(), ggplots after the feature selection
- 2. detection of outliers for continuous features box(), defined outlier() to detect the outliers
- 3. correlation/collinearity/chi-squared analysis cor(),
- 4. evaluation of distribution barplots, histograms, boxplots for each subtype, pair.panels() plot

2. Data Exploration

- I have used plot_intro function from DataExplorer package
- plot_intro provides an insight of what type of data is present along with that it provides the information about missing values
- plotted a graph to show the proportion of missing data for each variable.
- Apart from that, I have used str and summary to understand the structure of the data present
- Discarded variables with more than 25% of the data missing because they wouldn't be significant for further research.
- Used the mean imputation method, imputed the NA values with the mean of a specific column for the remaining variables with missing data.
- After cleaning the data, using the ggplot() I have analyzed the distribution of each subtype in the dataset
- 1. exploratory data plots: I used plot intro() from the DataExplorer package. plot intro gives an understanding of the sort of data present as well as information about missing values. In addition, I have used str() and summary() to comprehend the data's structure.

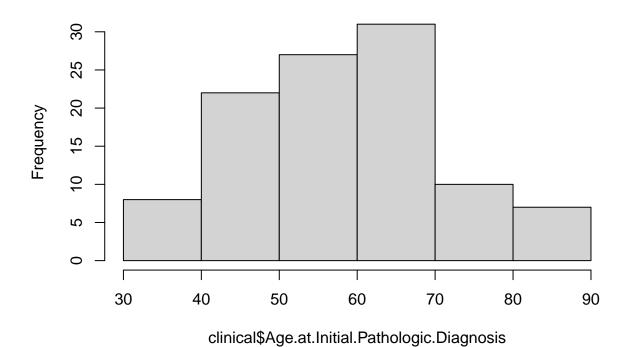
#Visualizing structure of the data set
plot_intro(proteomes)



It is essential to explore the clinical dataset as it has all the information about the pateints

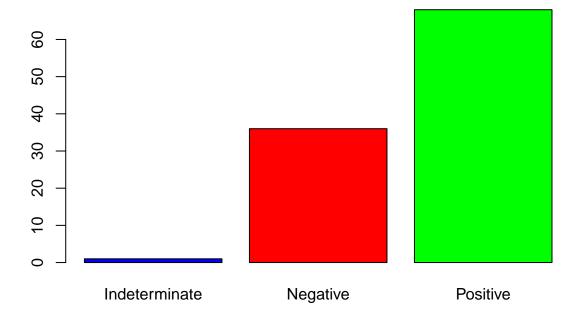
 $\label{lem:clinical} $$Age.at.Initial.Pathologic.Diagnosis $$\leftarrow$ as.numeric(clinical\$Age.at.Initial.Pathologic.Diagnosis)$$ 1 $$\leftarrow$ hist(clinical\$Age.at.Initial.Pathologic.Diagnosis)$$$

Histogram of clinical\$Age.at.Initial.Pathologic.Diagnosis



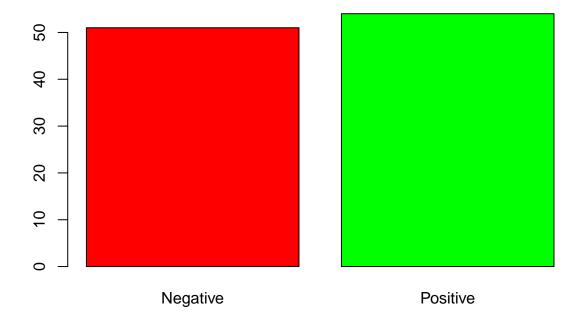
b1<- barplot(table(clinical\$ER.Status), col=c("blue", "red", "green"), main = "ER Status")

ER Status



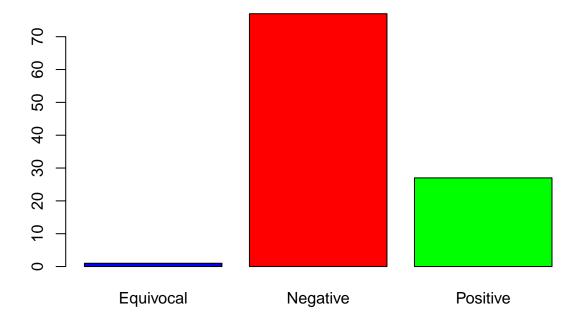
b2<- barplot(table(clinical\$PR.Status), col=c("red", "green"), main = "PR Status")

PR Status



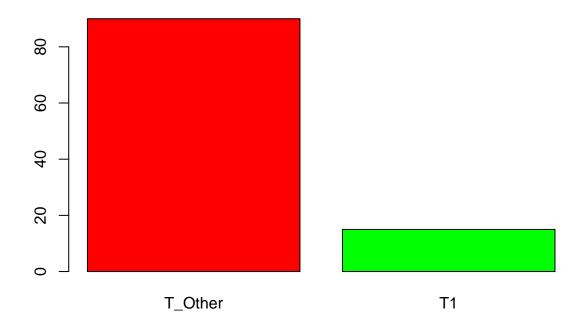
b3<- barplot(table(clinical\$HER2.Final.Status), col=c("blue", "red", "green"), main = "HER2 Final Statu

HER2 Final Status



b4<- barplot(table(clinical\$Tumor..T1.Coded), col=c("red", "green"), main = "Tumor--T1 Coded")

Tumor--T1 Coded



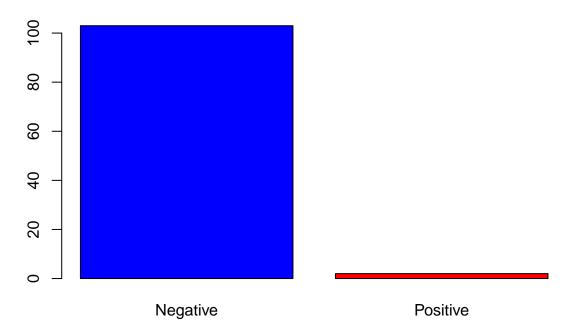
b5<- barplot(table(clinical\$Node.Coded), col=c("red", "green"), main = "Node-Coded")

Node-Coded

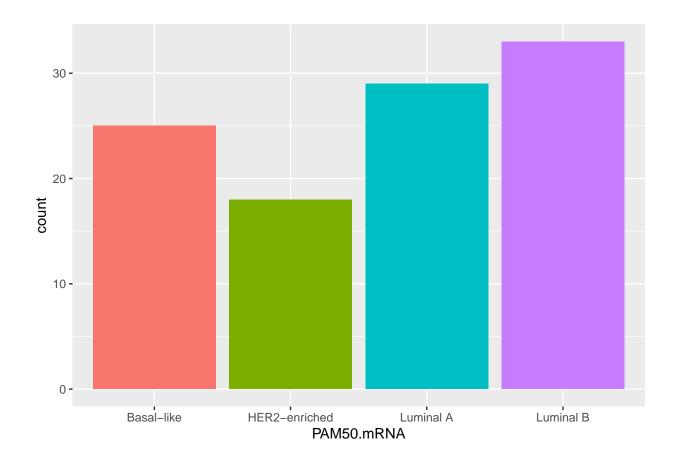


b6<- barplot(table(clinical\$Metastasis.Coded), col=c("blue", "red", "green"), main = "Metastasis-Coded"

Metastasis-Coded



b7<- ggplot(clinical,aes(x= `PAM50.mRNA`,fill=`PAM50.mRNA`))+geom_bar()+theme(legend.position = "none") b7



Missing Values

This plot_intro() plot shows that the dataset has missing observations. we can know the count of na's from summary() and also colSums()

colSums(is.na(proteomes))

##	Complete.TCGA.ID	NP_958782	NP_958785	NP_958786
##	0	0	0	0
##	NP_000436	NP_958781	NP_958780	NP_958783
##	0	0	0	0
##	NP_958784	NP_112598	NP_001611	NP_076965
##	0	0	0	0
##	NP_001367	NP_036222	NP_001138241	NP_001714
##	0	0	0	0
##	NP_002464	NP_001073936	NP_061198	NP_001074248
##	0	0	0	0
##	NP_001074419	NP_056009	NP_036355	NP_001155291
##	0	0	0	0
##	NP_002462	NP_000248	NP_036467	NP_004989
##	0	0	0	0
##	NP_001243024	NP_149043	NP_067047	NP_002461
##	0	0	19	0
##	NP_060003	NP_003793	NP_060004	NP_005954

##	0	0	0	0
##	NP_002463	NP_689914	NP_001242941	NP_060227
##	_ 0	_ 0	- 0	- 9
##	NP_055796	NP_065935	NP_008974	NP_001156806
##	2	0	_ 0	13
##	NP_000251	NP_001242970	NP_001094891	NP_057323
##	16	0	15	52
##	NP_002465	NP_005955	NP_001035202	NP_001070654
##	0	0	0	0
##	NP_079005	NP_000250	NP_878918	NP_892006
##	0	0	0	0
##	NP_149062	NP_001138722	NP_003118	NP_001123910
##	0	42	0	0
##	NP_001182461	NP_001243000	NP_004406	NP_689916
##	0	0	0	0
##	NP_060124	NP_443141	NP_008835	NP_000375
##	0	41	0	0
##	NP_001447	NP_001157789	NP_001448	NP_001449
##	0 ND 0011000E0	0 ND 003110	ND 040E6E	ND 00100000
## ##	NP_001120959	NP_003119	NP_842565	NP_001020029
##	NP_008877	NP_000338	NP_066022	NP_001243415
##	NF_008877	NF_000338	NF_000022	NF_001243413
##	NP 004478	NP_001243417	NP_005742	NP_671714
##	0 11 004470	N1_001243417	N1_000742	N1_0/1/14 0
##	NP_005520	NP 004360	NP_476507	NP_065816
##	0	0	0	0
##	NP 006280	NP 055874	NP 113584	NP 612482
##	_ 0	0	_ 0	_ 0
##	NP_006176	NP_006258	NP_055185	NP_001137485
##	0	0	9	0
##	NP_005045	NP_872394	NP_001019628	NP_872634
##	0	0	0	0
##	NP_005029	NP_003128	NP_005720	NP_004095
##	0	0	0	0
##	NP_002017	NP_997647	NP_997639	NP_009055
##	0	0	0	0
##	NP_003997	NP_004005	NP_004007	NP_001164655
##	0	0	0	15
##	NP_000055	NP_002323	NP_061027	NP_065872
##	0	0	0	0
##	NP_006827	NP_003117	NP_612429	NP_003283
##	0 ND 054722	0 ND 000460	0 ND 004364	0 ND 0040E0
##	NP_054733	NP_000468	NP_004361	NP_004850
##	0 ND 00000	0 ND 003961	ND 006604	0 00042
## ##	NP_009029	NP_003861	NP_006624	NP_839943
##	0 NP_002696	0 NP_002069	NP_001166184	0 NP_114141
##	NF_002090 0	NF_002009	0	0 NL_114141
##	NP_001254479	NP_001091738	NP_596870	NP_663304
##	NF_001234479	NF_001091738	NF_590070 7	NF_003304 0
##	NP_003179	NP_001229542	NP_001138668	NP_001014364
##	0	59	60	53
##	NP_003741	NP_003080	NP_002895	NP_067051
	000, 11		002300	00.001

##	0	0	0	0
##	NP_078873	NP_001166987	NP_065700	NP_055107
##	0	0	M1_000700	M1_000107
##	NP_002151	NP_061978	NP_006436	NP 006717
##	0	0	M _000 100	0
##	NP 001186211	NP 056493	NP_004230	NP 510880
##	0	0	0	0.00000
##	NP_004516	NP 579899	NP_038479	NP_001124459
##	0	0	M _000 17 0	0
##	NP 001124453	NP_001229752	NP 001002029	NP 009224
##	0	0	0	0
##	NP 000543	NP 057336	NP_005551	NP 937762
##	_ 0	0	_ 0	0
##	NP 004437	NP 005900	NP 002364	NP 055987
##	_ 0	0	_ 0	0
##	NP 542417	NP_037450	NP 001264	NP_001005271
##	_ 0	- 0	- 0	- 0
##	NP 056372	NP 001034679	NP 004645	NP_001008938
##	_ 0	- 0	_ 0	- 0
##	NP_005091	NP_942133	NP_001084	NP_004332
##	0	0	0	0
##	NP_002449	XP_003960896	XP_003960532	NP_203754
##	0	3	53	0
##	NP_001979	NP_001138599	NP_001072989	NP_001072990
##	0	2	0	0
##	NP_000005	NP_002855	NP_573566	NP_057427
##	0	0	0	0
##	NP_003364	NP_056083	NP_001123476	NP_001093
##	0	0	0	0
##	NP_004915	NP_001265272	NP_001095	NP_055426
##	0	0	0	0
##	NP_062536	NP_009131	NP_001257475	NP_001231509
##	0	0	0	0
##	NP_003557	NP_055710	NP_115645	NP_055872
##	0	5	31	0
##	NP_004362	NP_001091868	NP_006297	NP_004990
##	0	0	0	0
##	NP_056234	NP_056110	NP_001243194	NP_001243193
##	0	0	0	0
##	NP_006022	NP_000129	NP_001990	NP_000928
##	0	0	0	0 ND 004050
##	NP_940980	NP_068506	NP_003165	NP_001059
##	16	ND 066022	ND 00E436	ND 066967
##	NP_001058	NP_066933	NP_005436	NP_066267
##	0 ND 001101222	0 ND 001101330	0 ND 000000	ND 065010
##	NP_001191333	NP_001191332	NP_000028	NP_065210
##	0 ND 001120	ND 066197	ND 001140	ND OFF4F0
##	NP_001139	NP_066187	NP_001140	NP_055458
## ##	0 ND 733821	0 ND 733833	ND 005563	1 ND 116126
##	NP_733821 0	NP_733822	NP_005563	NP_116126
##	NP_005564	NP_001185486	NP_009049	NP_001019831
##	NF_003364 0	NF_001185486	NF_009049 0	NF_001019631
##	NP_008995	NP_001135452	NP_444253	NP_444254
πĦ	1/1 _000333	MI _001130402	Mr _444200	NF _444204

			•	•
##	20	0	0	0
##	NP_444259	NP_004949	NP_005723	NP_000867
##	0	0	0	0
##	NP_001054	NP_002283	NP_031382	NP_000019
##	0	0	12	0
##	NP_060024	NP_001193937	NP_001035089	NP_002284
##	0	0	0	0
##	NP 006050	NP 996744	NP_006274	NP_852118
	-	-	_	
##	6	0	0	0
##	NP_055490	NP_001098677	NP_009057	NP_000457
##	0	0	0	0
##	NP_006787	NP_002524	NP_000278	NP_112593
##	0	0	0	0
##	NP 065871	NP 006819	NP_071374	NP_683692
##	_ 0	_ 0	_ 0	_ 0
##	NP 612378	NP_005327	NP_001230829	NP_056197
	N1_012570 0	NI _003327	N1_001250025	
##	· ·	ŭ	•	0
##	NP_115927	NP_001258898	NP_001017963	NP_003290
##	21	0	0	0
##	NP_057376	NP_001615	NP_001092693	NP_002408
##	0	0	0	0
##	NP_055873	NP_001009814	NP_002263	NP_003371
##	0	0	0	0
##	NP_872313	NP 778253	NP_006112	NP_005373
##	0	0	0	0
##	NP 000414	· ·	•	· ·
	-	NP_006253	NP_006149	NP_002275
##	0	0	0	0
##	NP_066554	NP_002272	NP_001243211	NP_476429
##	0	0	0	0
##	NP_002046	NP_116116	NP_542785	NP_001099011
##	0	0	0	0
##	NP_057613	NP_778223	NP_005547	NP_149034
##	4	_ 0	0	_ 0
##	NP 258259	NP 778238	NP 775487	NP 149022
##	2	0	0	15
##	NP_001918	NP 000415	NP 005546	NP_005545
		NF_000413	-	
##	0	U 707000	0	0
##	NP_004684	NP_787028	NP_001609	NP_001073867
##	0	0	0	0
##	NP_055816	NP_056069	NP_006603	NP_008985
##	0	0	0	0
##	NP_004789	NP_065810	NP_002245	NP_001092763
##	0	0	0	0
##	NP_878905	NP_005541	NP_065867	NP_659464
##	7	3	12	14
##	NP_078980	NP_055889	NP_071396	NP_001099038
	-			NF_001099030
##	0	0	0	0
##	NP_001186795	NP_004312	NP_036442	NP_904325
##	0	0	0	0
##	NP_036565	NP_005886	NP_003161	NP_003237
##	0	0	0	0
##	NP_003238	NP_002102	NP_006608	NP_056986
##	0	_ 0	0	_ 0
##	NP_062535	NP_003248	NP_783297	NP_004136
	111 _002000	111 _000240	111 _100201	141 _004100

##	0	0	0	0
##	NP_008832	NP_001952	NP_001716	NP_004841
##	2	0	0	0
##	NP_005397	NP_001181875	NP_886553	NP_937887
##	0	0	0	0
##	NP 937885	NP_001185730	NP 001263347	NP_055681
##	_ 0	0	_ 0	_ 0
##	NP 056193	NP 001511	NP_072174	NP_073585
##	0	0	0	0
##	NP_056134	NP 065821	NP 002282	NP_001193583
##	0	0	0	0
##	NP 006303	NP 006302	NP 569707	NP 570924
##	NF_000303	NF_000302	NF_509707	NF_570924
	· ·	MD 000030	ND 662700	ND 056101
##	NP_002841	NP_002830	NP_663780	NP_056101
##	0	0	0	0
##	NP_000417	NP_001073291	NP_005550	NP_064505
##	0	0	12	0
##	NP_064506	NP_149072	NP_055968	NP_056121
##	0	0	0	0
##	NP_006026	NP_003598	NP_055641	NP_059995
##	0	0	0	0
##	NP_001075031	NP_055907	XP_003960679	XP_003960702
##	1	0	10	20
##	XP 003960698	NP 056993	NP_001157968	NP_078789
##	37	_ 0	21	_ 0
##	NP 006188	NP_003461	NP_000042	NP 705833
##	_ 0	_ 0	_ 0	0
##	NP 055791	NP 000480	NP 612114	NP 060473
##	0	0	0	0
##	NP 055928	NP 006466	NP 001129406	NP_001129407
##	0	0.000100	0	0
##	NP_597677	NP 002366	NP 001102	NP 056655
##	0	NI _002300	N1_001102	N1 _030033
	· ·	ND 027002	v	VD 001102120
##	NP_001234926	NP_937883	NP_115797	NP_001103132
##	UD 000444	0.00440	WD 004750	VD 050404
##	NP_006411	NP_006412	NP_004753	NP_059431
##	0	0	0	0
##	NP_004218	NP_037517	NP_002215	NP_002214
##	0	3	0	0
##	NP_001161744	NP_002213	NP_149129	NP_149130
##	0	0	0	0
##	NP_004333	NP_001124295	NP_001370	NP_056988
##	0	0	0	0
##	NP_002853	NP_002854	NP_005600	NP_001726
##	0	0	0	0
##	NP_060918	NP_001155971	NP_055847	NP_001093869
##	0	0	0	0
##	NP_003881	NP_001422	NP_001186318	NP_036439
##	0	- 0	_ 0	_ 0
##	NP_001245258	NP_001159477	NP_818932	NP_004428
##	0	0	0	0
##	NP_001245259	NP_001258928	NP_212132	NP_982272
##	0	0	0	0
##	NP_065863	NP_005517	NP_004497	NP_001035757
ππ	MI _000000	MI _003317	MI _004491	111 _001033131

	_	_	_	
##	0	0	3	22
##	NP_004512	NP_004513	NP_004975	NP_002145
##	0	0	0	0
##	NP_006635	NP_055093	NP_005349	NP_056667
##	0	0	0	0
##	NP_065741	NP_115593	NP_060217	NP_078944
##	0	0	0	0
##	NP_005119	NP_001186871	NP_001073883	NP_057688
##	0	21	0	0
##	NP_060903	NP_940953	NP_001509	NP_001003795
##	_ 0	_ 0	_ 0	- 0
##	NP 775808	NP_938149	NP_938148	NP 009041
##	0	_ 0	0	_ 0
##	NP 001258152	NP 443075	NP_695012	NP_002152
##	0	0	0	0
##	NP_940978	NP 000292	NP 005568	NP 829884
##	0	M _000232	M _000000	M _020001
##	NP_056391	NP 055454	NP 064502	NP 005210
##	NF_030391 0	NF_055454	NF_004502	NF_003210
##	NP 036533	NP_000177	ND 000104	ND 066303
	-	NP_000177	NP_002104	NP_066303
##	0 ND 005657	v	ND 002000	ND OFFOR
##	NP_005657	NP_001188479	NP_003226	NP_055951
##	0	0	0	VD 055050
##	NP_001014972	NP_009123	NP_005106	NP_055950
##	0	0	0	0
##	NP_055205	NP_060542	NP_006286	NP_065073
##	0	0	0	0
##	NP_056281	NP_057535	NP_000087	NP_004578
##	0	0	0	0
##	NP_065972	NP_002435	NP_003370	NP_002897
##	0	0	0	0
##	NP_001247425	NP_060783	NP_002999	NP_001146
##	0	0	22	0
##	NP_056090	NP_073616	NP_003852	NP_001193814
##	0	0	0	0
##	NP_001193813	NP_001322	NP_001078935	NP_005378
##	0	0	0	0
##	NP_001186308	NP_004184	NP_001155901	NP_037506
##	0	0	0	0
##	NP_002334	NP_001032412	NP_003282	NP_004482
##	0	0	0	0
##	NP_003592	NP_003060	NP_620604	NP_055644
##	0	0	0	0
##	NP_009117	NP_001123892	NP_003066	NP_003065
##	0	0	0	0
##	NP_003473	NP_733751	NP_061322	NP_001181885
##	0	0	0	0
##	NP_055535	NP_005981	NP_002532	NP_001158508
##	_ 0	_ 0	- 0	0
##	NP_060715	NP_005500	NP_001167587	NP_001005619
##	_ 0	_ 0	_ 0	
##	NP_001005731	NP_079330	NP_056240	NP_008986
##	0	_ 0	- 0	_ 0
##	NP_056111	NP_001123520	NP_653259	NP_055504
	-	<u>-</u>		

##	0	0	0	0
##	NP_872579	NP_004450	NP_061897	NP_056288
##	0	0	M1_001037	N1 _000200
##	NP 060101	NP_001012985	NP_001371	NP_004937
##	0	0	M _001011	0
##	NP 079216	NP 874365	NP_056171	NP 060684
##	0	0.00	N1_0001/1	0 N1 _00000
##	NP 940890	NP 001087	NP_942127	NP_001368
##	N1 _540030	M1_001007	N1_342127	M1_001000
##	NP 055627	NP 001035864	NP 001106197	NP 055820
##	0	0	0	0
##	NP 061903	NP 057417	NP 001894	NP 004380
##	0	0	0	0 0 0 0 0
##	NP 037398	NP 000079	NP 001835	NP_002902
##	0 007	M1 _00007 <i>9</i>	M _001000	N1 _002302
##	NP 001158287	NP 001158286	NP 055524	NP_114032
##	0	0	0	0
##	NP_004492	NP 570849	NP_060395	NP_001138295
##	0	M1_070049	M _000050	0
##	NP 004808	NP 001163886	NP 001163885	NP_001164100
##	0 004000	0	0	0
##	NP 001262	NP_001261	NP 060250	NP_115597
##	0	0	0	M _110037
##	NP_079410	NP 001257903	NP_004694	NP_001077054
##	3	19	0	0
##	NP 000170	NP 005947	NP_001229696	NP 056255
##	0	0	0	0
##	NP 002289	NP 005023	NP 002661	NP_001193980
##	0	0	0	0
##	NP 056097	NP 055912	NP_001193973	NP 036601
##	0	0	0	0
##	NP 055845	NP 075463	NP_001073869	NP 542194
##	0	_ 0	0	_ 0
##	NP 954650	NP 056025	NP 001012279	NP 003962
##	- 0	19	- 58	- 0
##	NP 001124000	NP 001123999	NP 001238900	NP 055948
##	0	0	0	_ 0
##	NP_006380	NP_001035957	NP_872270	NP_001193725
##	0	0	0	0
##	NP_001193727	NP_000289	NP_060110	NP_001167448
##	0	0	0	0
##	NP_004902	NP_005304	NP_004376	NP_941372
##	0	0	0	0
##	NP_066964	NP_001138231	NP_001127910	NP_942147
##	0	0	0	3
##	NP_060554	NP_001129069	NP_001073922	NP_071504
##	0	0	0	0
##	NP_006830	NP_001093639	NP_001093640	NP_006588
##	0	0	0	0
##	NP_005338	NP_005337	NP_001035924	NP_068814
##	0	0	0	0
##	NP_694881	NP_005518	NP_002146	NP_060898
##	0	0	0	0
##	NP_005959	NP_001596	NP_002799	NP_115784

##	0	0	0	0
##	NP_031387	NP_002499	NP_060414	NP_066997
##	0	0	0	0
##	NP_203752	NP_001139381	NP_004658	XP_003403905
##	0	0	0	16
##	NP 054831	NP 003472	NP 003931	NP_057662
##	0	0	0	0
	•	· ·	•	ND 001024020
##	NP_036425	NP_055949	NP_958431	NP_001034230
##	0	0	0	0
##	NP_619538	NP_476516	NP_078837	NP_036387
##	0	0	0	0
##	NP_057212	NP_036265	NP_060090	NP_036350
##	0	0	0	0
##	NP 060857	NP_060697	NP 620305	NP 115571
##	_ 0	_ 0	_ 0	_ 0
##	NP 004229	NP 001025177	NP 663782	NP 477352
##	N1 _004223	N1_001023177	NI _003702	N1 _ 1 77332
	ND 004005000	VD 004005004	VD 004006	VD 004005000
##	NP_001005360	NP_001005361	NP_004936	NP_001005336
##	0	0	0	0
##	NP_056384	NP_037423	NP_004125	NP_803187
##	0	0	0	0
##	NP_006428	NP_060404	NP_061836	NP_694984
##	0	0	3	6
##	NP 006833	NP_055888	NP_056371	NP 065859
##	- 0	_ 0	- 0	_ 0
##	NP 001106189	NP 001106190	NP 005076	NP 001026884
##	0	M _001100100	0	0
##	NP 071934	NP 055506	NP_056992	NP_001138678
	0 0	NF_033300	NF_030992	NF_001130070
##	•	ND 001025007	ND 001100660	ND 000017
##	NP_056154	NP_001035207	NP_001128662	NP_003617
##	0	0	0	0
##	NP_003651	NP_003616	XP_003960947	NP_055868
##	0	0	0	0
##	NP_001073953	NP_055423	NP_055191	NP_001028200
##	47	0	0	0
##	NP_006268	NP 003015	NP 001001132	NP 671494
##	- 0	_ 0	_ 0	_ 0
##	NP_005115	NP_001036015	NP_036546	NP_003613
##	0	0	M1_000010	0
	-	•	ND 001042407	·
##	NP_001185844	NP_003612	NP_001243497	NP_056280
##	0	0	0	0
##	NP_000179	NP_000180	NP_002106	NP_079406
##	0	0	0	0
##	NP_006006	NP_065783	NP_619520	NP_150648
##	0	0	0	0
##	NP_001155855	NP_001035905	NP_001177371	NP_001075019
##	0	0	0	0
##	NP_055318	NP_000499	NP_001408	NP_004477
##	0	0	0	0
##	NP_001099008	NP_000120	NP_001120683	NP_001245227
##	111 _001099000	141 _000120	141 _001120003	NI_001240221
	U ND 001070675	ND 060604	MD AFFOCA	ND 001460
##	NP_001070675	NP_060621	NP_055990	NP_001460
##	0	0	0	0
##	NP_057441	NP_001107018	NP_001107019	NP_056392

##	0	0	0	0
##	NP_001138997	NP_001138994	NP_036525	NP_006178
##	0	0	0	0
##	NP_005487	NP_065910	NP_001640	NP_065768
##	0	0	0	22
##	NP 055464	NP 002958	NP 002473	NP_001182122
##	0	0	0	0
	•	· ·	· ·	ND 001021001
##	NP_689511	NP_006169	NP_002471	NP_001231921
##	0	0	0	0
##	NP_002472	NP_001184060	NP_002652	NP_003061
##	0	0	0	0
##	NP_001122321	NP_001122317	NP_004757	NP_055456
##	0	0	0	0
##	NP_002365	NP 001034627	NP 061889	NP 000215
##	0	0	5	0
##	NP_061883	NP_000214	NP 002271	NP_002268
	-	NF_000214	NF_002271	NF_002200
##	0	0	0	4
##	NP_853515	NP_003762	NP_689562	NP_002267
##	0	1	0	0
##	NP_002269	NP_066293	NP_000413	NP_002266
##	8	43	0	0
##	NP_000517	NP_005548	NP_056330	NP_000412
##	_ 0	_ 0	_ 0	_ 0
##	NP_705694	NP_001073027	NP_003950	NP 005329
##	0	0	0	0
##	•	· ·	· ·	NP_000168
	NP_001073341	NP_001008895	NP_001265443	NP_000100
##	0	0	0	0
##	NP_001121134	NP_001121138	NP_001121135	NP_003913
##	0	0	0	0
##	NP_001185594	NP_056128	NP_008841	NP_009045
##	0	0	0	0
##	NP_006697	NP_777597	NP_056130	NP_060548
##	_ 0	_ 0	_ 0	_ 0
##	NP 005246	NP 005325	NP 037452	NP 056261
##	0	0	0	0.000201
##	NP 008860	NP 056390	NP 000909	NP 056224
	NF_000000	NF_030390	NF_000909	NF_030224
##	0	VP 000450	U	0
##	NP_006810	NP_002453	NP_002454	NP_001701
##	0	0	0	0
##	NP_071349	NP_001106963	NP_001106967	NP_997249
##	0	0	0	0
##	NP_001106965	NP_001779	NP_001011553	NP_060713
##	0	0	0	0
##	NP 004395	NP_055944	NP_653311	NP_001092281
##	0	0	0	0
##	NP 775873	NP_055966	NP_055789	NP_597734
		_	_	
##	0	0	0	0
##	NP_001122405	NP_055144	NP_060040	NP_066363
##	0	0	0	0
##	NP_065176	NP_660093	NP_001104488	NP_055718
##	0	12	12	36
##	NP_361013	NP_001004060	NP_055102	NP_079031
##	_ 0	_ 0	_ 0	_ 0
##	NP_001013865	NP_055831	NP_005177	NP_001307
	001010000	300001	0001/1	001001

	^	4	^	
##	0	1	0	0
##	NP_996810	NP_996811	NP_001156423	NP_001092
##	0	0	0	0
##	XP_003960357	NP_001264012	XP_003846703	NP_001264232
##	0	0	0	56
##	NP_001605	NP 001135417	NP_001606	NP_001091
##	_ 0	- 0	_ 0	_ 0
##	NP 001077007	NP 001017992	NP 001093241	NP 001264335
##	M1_001077007	N1_001017332	M1_001030241	N1_001204000
	VD 002706	ND 004500	ND 007FFF	ND 005270
##	NP_003706	NP_064522	NP_997555	NP_005372
##	0	0	0	0
##	NP_001005751	NP_060702	NP_056077	NP_001129477
##	0	0	0	0
##	NP_061874	NP_004387	NP_001091974	NP_001347
##	0	0	0	0
##	NP_004651	NP 077726	NP 001839	NP 004960
##	0	0	0	0
##	NP_002147	NP_006301	NP_000080	NP 009330
##	NF_002147	NF_000301	NF_000000	NF_009330
	· ·	•	0	0
##	NP_644671	NP_003142	NP_003631	NP_001092879
##	0	5	0	0
##	NP_001136086	NP_060360	NP_062538	NP_055586
##	0	3	0	0
##	NP_002636	NP_004784	NP_001076581	NP_056289
##	0	0	0	0
##	NP_001265387	NP_001265389	NP_004738	NP 064627
##	0	_ 0	_ 0	2
##	NP_055129	NP 001244957	NP 055699	NP 004810
##	_ 0	_ 0	0	_ 0
##	NP 060225	NP 001180384	NP 002262	NP 036548
##	M _000220	M1_0001100001	WI _002202	M1_000010
	ND 001020121	U 001000100	O 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	O 1020100
##	NP_001230131	NP_001030128	NP_001030126	NP_001030129
##	0	0	0	0
##	NP_001257700	NP_001139145	NP_001139144	NP_003594
##	0	0	0	0
##	NP_001139146	NP_003604	NP_001243778	NP_001399
##	0	0	0	2
##	NP_055061	NP_001420	NP_004371	NP_056989
##	_ 0	_ 0	_ 0	_ 0
##	NP_115921	NP_001419	NP_001188412	NP_001966
##	0	0	0	0
##	NP_001967	NP_003096	NP_003655	NP_004635
		_		_
##	0 ND 001065440	0 ND 001100007	ND 055501	ND 055700
##	NP_001265440	NP_001106207	NP_055521	NP_055799
##	0	0	0	0
##	NP_001138829	NP_055704	NP_001240628	NP_065845
##	0	0	0	8
##	NP_056044	NP_039269	NP_056018	NP_054722
##	0	0	0	0
##	NP 004932	NP_054828	NP_060022	NP_877423
##	_ 0	0	- 0	_ 0
##	NP_001245303	NP_067058	NP_001265119	NP_001138632
##	0	0	0	0.0000000
	-	· ·	-	NP_001018005
##	NP_001018008	NP_112600	NP_000357	ML _001010002

шш	0	0	^	^
##	0	ND 000000	ND 002004	VD 002000
##	NP_001018020	NP_998839	NP_003281	NP_003280
##	0	0	0	0
##	NP_001036817	NP_689476	NP_005110	NP_055554
##	0	0	0	0
##	NP_001070908	NP_002005	NP_005984	NP_001171914
##	0	0	0	0
##	NP_006639	NP_115763	NP_065973	XP_003960567
##	0	0	0	17
##	NP 067062	NP 001017915	NP 001558	NP_005769
##	- 0	_ 0	_ 0	_ 0
##	NP_001485	NP_001484	NP 055625	NP 036329
##	0	0	0	0
##	NP 055672	NP_001230694	NP 795352	NP 001248335
##	NF_033072	NF_001230094	NF_190002	NF_001240333
	v	· ·	UD 004404007	ND 055000
##	NP_005753	NP_003160	NP_001124297	NP_055903
##	0	0	0	0
##	NP_001020014	NP_055315	NP_055842	NP_055427
##	0	0	0	0
##	NP_060318	NP_003866	NP_001129487	NP_056073
##	0	0	0	0
##	NP_001073884	NP_937879	NP_055599	NP_002717
##	0	0	0	0
##	NP_005057	NP_031389	NP_002038	NP_056175
##	0	0	0	0
##	NP 001123917	NP 003857	NP 001681	NP 001025226
##	M _001120317	M1_000007	M _001001	M1_001020220
##	ŭ	ND 001007EE4	OOOOO	ND OGOESO
	NP_003391	NP_001007554	NP_009089	NP_060530
##	0	0	0	VD 004504
##	NP_079461	NP_002379	NP_001257401	NP_064504
##	0	0	0	0
##	NP_060707	NP_150241	NP_150243	NP_150242
##	0	0	0	0
##	NP_150247	NP_150252	NP_150249	NP_002188
##	0	0	0	0
##	NP_000405	NP_001186221	NP_005426	XP_003960466
##	0	0	0	0
##	NP_001003702	NP_073739	NP_055754	NP_001120
##	- 0	_ 0	_ 0	_ 0
##	NP 937791	NP_036333	NP_001139138	NP_056981
##	0	0	0	0
##	NP_000923	NP_055160	NP_001737	NP_004353
	-	_		_
##	0 ND 0011261E7	0 ND 001000760	ND OCEONE	ND 442100
##	NP_001136157	NP_001020769	NP_065805	NP_443180
##	0	0	0	0
##	NP_056138	NP_055768	NP_036192	NP_001265393
##	0	0	0	0
##	NP_001265392	NP_671492	NP_001004065	NP_001130034
##	0	0	0	0
##	NP_443749	NP_001409	NP_001036024	NP_001089
##	_ 0	0	0	0
##	NP_057191	NP_055596	NP_000173	NP_115826
##	0	_ 0	0	_ 0
##	NP_001092750	NP_002209	NP_001159921	NP_001035810
	001002100	002200	001100021	001000010

##	0	0	0	0
##	NP_006251	NP_004719	NP_076950	NP_898868
##	0	0	0	0
##	NP_036540	NP_898869	NP_005042	NP_057460
##	0	0	0	0
##	NP_001244928	NP_002963	NP_112224	NP_036365
##	- 0	- 0	- 0	- 0
##	NP_055592	NP_001029345	NP_001257293	NP_002878
##	0	0	0	0
##	NP_001128715	NP_001128717	NP_001128716	NP_060587
##	0	0	0	M _000001
##	NP_065801	NP_061141	NP_001034764	NP_055647
##	N1 _005001	M1_001141	N1_001034704	N1_000047
##	NP_001240841	NP_001240844	NP_003350	NP_005145
##	NF_001240041	NF_001240044	NF_003350	NF_005145
	ND 610144	ND 00063E	ND 002610	ND 00100E476
##	NP_612144	NP_002635	NP_003619	NP_001005476
##	VD 004.000	U	UD 00440400E	U 004404000
##	NP_001323	NP_001131024	NP_001131025	NP_001131022
##	0	0	0	0
##	NP_006300	NP_055018	NP_036437	NP_001229766
##	0	0	0	0
##	NP_001001937	NP_001229489	NP_001229488	NP_663719
##	0	0	0	0
##	NP_001155033	NP_001155032	NP_001155035	NP_722549
##	0	0	0	0
##	NP_733763	NP_002651	NP_877963	NP_001248755
##	0	0	0	0
##	NP_005641	NP_071376	NP_112533	NP_112420
##	0	0	0	0
##	NP_002127	NP_001011724	NP_036441	NP_573573
##	0	0	0	0
##	NP_277052	NP_065996	NP_001972	NP_001166935
##	1	0	0	0
##	NP_940927	NP_001239029	NP_060046	NP_060644
##	3	0	5	0
##	NP_001189472	NP_001904	NP_056082	NP_001020262
##	0	0	0	0
##	NP_001081	NP_005644	NP_001166923	NP_055368
##	0	0	0	0
##	NP_005906	NP_056450	NP_008971	NP_653333
##	0	0	0	0
##	NP_055484	NP_061918	NP_001239007	NP_003354
##	0	0	0	0
##	NP_004642	NP_055601	NP_075384	NP_004334
##	0	0	0	0
##	NP_001231639	NP_005101	NP_031398	NP_001119527
##	0	0	0	0
##	NP_001119526	NP_055045	NP_055750	NP_002037
##	0	0	0	0
##	NP_001243728	NP_001159581	NP_001159582	NP_115967
##	0	0	0	3
##	NP_777576	NP_689508	NP_689547	NP_056070
##	0	0	0	0
##	NP_001171730	NP_002448	NP_056107	NP_001171725

##	0	0	0	0
##	NP_055860	NP_004517	NP_001036216	NP_005853
##	0	0	0	0
##	NP_036579	NP_005757	NP_001001715	NP_001157745
##	7	0	6	0
##	NP_001349	NP_003578	NP_055578	NP_002516
##	_ 0	_ 0	_ 0	_ 0
##	NP 061160	NP 003989	NP 078938	XP_003960289
##	0	0	0	0
##	XP_003960292	XP_003846537	XP_003846569	NP 005522
##	M _0000000202	M _0000 10001	43	M _000022
##	NP_001193496	NP_689714	NP 056056	NP 073602
	NF_001193490	-	MF_030030	NF_073002
##	VD 004400476	0 ND 055447	ND 000050	VD 000074
##	NP_001129476	NP_055447	NP_036252	NP_003671
##	0	0	0	0
##	NP_620636	NP_995314	NP_005328	NP_001177893
##	0	0	0	0
##	NP_006420	NP_733765	NP_005164	NP_777615
##	0	0	0	0
##	NP_775293	NP_061936	NP_000101	NP_005086
##	0	0	0	0
##	NP 963893	NP 001164633	NP 001014437	NP_001181926
##	_ 0	0	_ 0	_ 0
##	NP 002547	NP 110385	NP_006358	NP 071362
##	0	0	0	M1_071002
##	NP_001265114	NP_001265115	NP_001002814	NP 055719
##	NF_001203114 0	NF_001203113	NF_001002014	NF_055/19
	ŭ	ND 000144	ND 045353	ND 004004
##	NP_998827	NP_003141	NP_945353	NP_004981
##	0	0	0	0
##	NP_001188474	NP_037418	NP_000242	NP_001157281
##	0	0	0	0
##	NP_003963	NP_110379	NP_689739	NP_476518
##	0	0	0	2
##	NP_006382	NP_006381	NP_065069	NP_060719
##	0	0	0	0
##	NP_001258968	NP_055242	NP_699197	NP_001188408
##	0	0	2	0
##	NP_078811	NP_056991	NP_003581	NP_056145
##	_ 0	- 0	- 0	- 0
##	XP_001713902	NP_005989	NP_001008800	NP_002559
##	0	0	0	0
##	NP_001129126	NP_003810	NP_112241	NP_001118228
##	0	0 000010	N1_112241 0	NI _001110220
	•	-	·	ND 00101700E
##	NP_543022	NP_001108206	NP_001012995	NP_001017995
##	0	VD 055500	0	0
##	NP_055446	NP_055520	NP_004938	NP_002256
##	0	0	3	0
##	NP_005111	NP_443728	NP_036205	NP_006421
##	0	9	0	0
##	NP_919223	NP_068579	NP_005137	NP_000349
##	0	0	0	0
##	NP_079390	NP_003676	NP_003893	NP_003925
##	0	0	_ 0	0
##	NP_056169	NP_006576	NP_055661	NP_000419
		_,,,,,		

##	0	0	0	0
##	NP_003469	NP_055429	NP_055417	NP_005236
##	0	0	0	0
##	NP_001008781	NP_005073	NP_004765	NP_009114
##	17	_ 0	_ 0	_ 0
##	NP 858058	NP 858059	NP 005598	NP_722560
##	0	M1_000009	M1_000050	N1_122000
	· ·	· ·	•	VD 000047
##	NP_002617	NP_000280	NP_002618	NP_036347
##	0	0	0	0
##	NP_006247	NP_998725	NP_002732	NP_001739
##	0	0	0	0
##	NP_056380	NP_004809	NP_005434	NP_004661
##	0	0	0	0
##	NP 056262	NP 055870	NP 060760	NP 002829
##	0	0	0	0
##	NP_006422	NP 056046	NP 000056	NP 000225
	NF_000422	NF_030040	NF_000030	NF_000225
##	0	0	0	0
##	NP_001243806	NP_660208	NP_002793	NP_001186913
##	0	0	0	0
##	NP_002851	NP_001017423	NP_001393	NP_001949
##	0	0	0	0
##	NP_005132	NP_055518	NP_000497	NP_060853
##	0	0	0	0
##	NP_006816	NP_005907	NP 009121	NP_001239264
##	0	0	0	0
##	NP 005691	NP 006730	NP_002201	NP 003079
##	NF_003091	NF_000730	NF_002201	NF_003079
	· ·	ND 004600	ND 065705	ND 004FF4
##	NP_004730	NP_004680	NP_065795	NP_004554
##	0	0	0	0
##	NP_002582	NP_001184223	NP_001184222	NP_001378
##	0	0	0	0
##	NP_001304	NP_001014809	NP_003277	NP_443195
##	0	0	0	0
##	NP 009176	NP 002964	NP 003743	NP_001093131
##	- 0	_ 0	- 0	- 0
##	NP 060676	NP 055915	NP 001181866	NP_001002843
##	0	0	0	0.0000000
	ND 060126	ND 006072	ND 116002	ND OOFOO1
##	NP_060136	NP_006073	NP_116093	NP_005991
##	0	0	0	0
##	NP_997195	NP_061816	NP_079079	XP_003846706
##	0	0	0	0
##	NP_001099032	NP_002068	NP_001263206	NP_001124330
##	0	0	0	0
##	NP_003899	NP_056173	NP_003583	NP_003897
##	0	0	0	0
##	NP_004604	NP_003232	NP_001810	NP_060705
##	0	20	0	0
			ND 010260	ND OOOESO
##	NP_689969	NP_066014	NP_919269	NP_000539
##	0	0	3	U 0045:0
##	NP_000201	NP_149100	NP_078856	NP_064716
##	0	0	0	0
##	NP_073744	NP_002476	NP_001073898	NP_065953
##	0	0	0	0
##	NP_001295	NP_942568	NP_055945	NP_071496

##	0	0	0	0
##	NP_114381	NP_002130	NP_062556	NP_055284
##	0	0	0	0
##	NP_001138818	NP_001158275	NP_055548	NP_001186090
##	0	0	0	0
##	NP_078893	NP_001009881	NP_055805	NP_919277
##	0	5	0	0
##	NP_065076	NP_054878	NP_000245	NP_060773
##	0	0	0	0
##	NP_000286	NP_001137293	NP_001243399	NP_055839
##	0	0	_ 0	- 0
##	NP_075048	NP_115786	NP_056972	NP_056311
##	53	_ 0	_ 0	- 0
##	NP_004439	NP_005219	NP 005226	NP_443198
##	0	_ 0	_ 0	_ 0
##	NP_055048	NP_055994	NP_055637	NP_004913
##	0	_ 0	_ 0	_ 0
##	NP_003133	NP_001116105	NP_056060	NP_690001
##	_ 0	_ 0	_ 0	- 6
##	NP_001191008	NP_068375	NP_002207	NP_055640
##	- 6	_ 0	_ 0	_ 0
##	NP_006030	NP_004035	NP 000282	NP_620061
##	- 0	- 0	- 0	_ 0
##	NP_005582	NP_001743	NP 005817	NP_001095868
##	0	_ 0	0	0
##	NP_006363	NP_001153149	NP_065905	NP_055416
##	0	0	0	_ 0
##	NP_006786	NP_644670	NP_055415	NP_005768
##	- 0	- 0	- 0	_ 0
##	NP 891553	NP_001123579	NP_005543	NP_073733
##	0	0	0	_ 0
##	NP 958929	NP_803136	NP 001888	NP_062565
##	0	_ 0	0	_ 0
##	NP_001171721	NP 008878	NP_005142	NP_113619
##	0	0	0	0
##	NP_001034558	NP_004945	NP_001156769	NP_001122390
##	0	_ 0	_ 0	_ 0
##	NP_001122392	NP_061120	NP_001186796	NP_113605
##	0	0	3	3
##	NP_079413	NP_065916	NP_945314	NP_055105
##	0	_ 0	_ 0	_ 0
##	NP_001265399	NP_055841	NP_003137	NP_653174
##	0	0	0	0
##	NP_001165780	NP_116178	NP_009293	NP_060549
##	0	0	0	0
##	NP_036286	NP_079128	NP_036331	NP_060099
##	0	0	0	0
##	NP_598000	NP_055806	NP_005883	NP_443137
##	0	0	0	0
##	NP_783863	NP_065842	NP_057365	NP_004774
##	0	N1_000042	M1_007000	0 0
##	NP_057235	NP_001129663	NP_056308	NP_694856
##	0	0	0	0
##	NP_001182073	NP_056014	NP_775866	NP_001153705
	001102010	000011	, , 0000	

##	0	0	0	0
##	NP_000693	NP_001243142	NP_653300	NP_000695
##	0 00000	0	0	M1 _000030
##	NP_001172014	NP 006368	NP_001074003	NP 001073890
##	0	3	20	M1_001070000
##	NP 996826	NP_001123616	NP_066548	NP_061184
##	0	0	0	0
##	NP 065134	NP 542166	NP_863651	NP_055828
##	0	0	M _000001	N1 _000020 0
##	NP 060506	NP_003740	NP 976324	NP 079524
##	0	0 000	0	0
##	NP_059830	NP_003541	NP 055448	NP 002936
##	0	0	0	0
##	NP 112553	NP 004564	NP 006504	NP 060191
##	M _112000	100±000_1M	11 _000004	M1_000131
##	NP 653271	NP_149107	NP 001166216	NP 006355
##	12	M _110107	0	0
##	NP 079272	NP_079033	NP 079532	NP_009178
##	0	0	0	0
##	NP 942583	NP 982284	NP 002218	NP 003322
##	0	0	0	0
##	NP_001004019	NP 001989	NP_064581	NP 001934
##	0	0	0	0
##	NP 001138439	NP_006145	NP 940682	NP 000143
##	0	0	0	0
##	NP_002678	NP_001009921	NP 001001998	NP 060167
##	0	0	0	0
##	NP 001116298	NP 001116297	NP 079215	NP 000886
##	0	0	_ 0	_ 0
##	NP 001243573	NP 115812	NP 001145	NP 001150
##	- 0	_ 0	- 0	- 2
##	NP 000370	NP 065875	NP 001186346	NP 002430
##	_ 0	_ 0	15	_ 0
##	NP 077719	NP 000426	NP 060087	XP_003960240
##	0	_ 0	_ 2	19
##	NP_000030	NP_001003698	NP_005398	NP_612808
##	0	0	30	53
##	NP_002826	NP_005104	NP_001127853	NP_001121113
##	0	0	0	0
##	NP_031383	NP_079434	NP_002085	NP_060564
##	0	0	0	0
##	NP_001003800	NP_001705	NP_001138593	NP_001124
##	0	0	54	0
##	NP_005553	NP_061486	NP_002898	NP_001230106
##	0	0	0	0
##	NP_005156	NP_001026973	NP_004178	NP_001036068
##	0	0	0	0
##	NP_006609	NP_067083	NP_001140177	NP_689854
##	0	0	0	0
##	NP_055955	NP_003326	NP_005222	NP_444507
##	0	0	0	0
##	NP_001107558	NP_058625	NP_005262	NP_036216
##	0	0	0	0
##	NP_055692	NP_055680	NP_078799	NP_001257327

##	0	0	0	0
##	NP_068831	NP_001895	NP_036419	NP_085152
##	0	_ 0	0	9
##	NP_009090	NP_002624	NP_570854	NP_060252
##	0	0	0	0
##	NP_038476	NP_001119	NP_003908	NP_001987
##	0	0	0	0
##	NP_006477	NP_006476	NP_542164	NP_659412
##	0	0	0	0
##	NP_001098057	NP_653283	NP_002261	NP_038461
##	0	9	0	0
##	NP_005539	NP_064710	NP_000086	NP_003239
##	0	0	0	0
##	NP_009043	NP_055885	NP_057508	NP_653173
##	0	3	0	0
##	NP_061915	NP_057424	NP_055062	NP_002931
##	U	0	0	VD 054000
##	NP_002941	NP_036228	NP_001185462	NP_851999
## ##	ND 00110E462	ND 000403	ND 001070060	NP_002899
##	NP_001185463	NP_002493	NP_001070962	NP_002099
##	NP_055756	NP 060487	NP 000089	NP_001005373
##	NF_033730 0	0 00407	0 00008	NF_001003373
##	NP 000866	NP_000199	NP_055030	NP_036424
##	0	0	30	0
##	NP 001245257	NP_055861	NP_006175	NP_036440
##	0	0	0	0
##	NP_001005366	NP_002206	NP_001159907	NP_056141
##	6	0	0	0
##	NP_065813	NP_055665	NP_001258816	XP_003960178
##	0	0	0	0
##	NP_056976	NP_057225	NP_006132	NP_001138887
##	43	0	0	0
##	NP_055040	NP_002707	NP_742067	NP_835461
##	0	0	0	0
##	NP_001185707	NP_002800	NP_054908	NP_789771
##	0	0 ND 000075	0 ND 000074	0 ND 004070040
##	NP_057190	NP_000875	NP_000874	NP_001073312
##	ND 003004	0 ND 005400	ND 057526	0 ND 000044
## ##	NP_003904 0	NP_005490 0	NP_057526	NP_002044
##	NP_004111	NP_060754	NP_001139575	NP_000578
##	0	0	N1_001133373	M _000378
##	NP_852664	NP_005018	NP_852556	NP_003620
##	0	0	0	0
##	NP_694985	NP_001191236	NP_060653	NP_001161328
##	0	0	0	0
##	NP_001093592	NP_001230368	NP_690868	NP_891554
##	0	_ 0	- 0	_ 0
##	NP_203744	NP_000519	NP_001073860	NP_001124160
##	0	0	0	0
##	NP_002119	NP_004500	NP_001171944	NP_612411
##	0	0	0	0
##	NP_009168	NP_787048	NP_060830	NP_005569

	_	_	_	_
##	9	0	0	0
##	NP_003312	NP_005078	NP_001013456	NP_000166
##	0	0	0	0
##	NP_000473	NP_064601	NP_008880	NP_001139280
##	0	0	0	0
##	NP_000878	NP_005344	NP_061176	NP_073609
##	0	7	0	0
##	NP_036580	NP_003143	NP_078900	NP_055865
##	0	0	0	0
##	NP 000054	NP 001171534	NP 001041666	NP_690856
##	_ 0	- 0	- 0	_ 0
##	NP_001037	NP_000329	NP 001008493	NP 001753
##	0	16	0	0
##	NP 001009186	NP 006575	NP 001180458	NP_001157632
##	M1_001003100	M1_000070	001100-100	N1_001107002
##	NP 001229827	NP 061878	NP_002100	NP_036340
	_	-	NF_002100	NF_030340
##	0	0	VD 057040	U 00400040E
##	NP_006653	NP_001986	NP_057318	NP_001009185
##	0	0	0	0
##	NP_054749	NP_476510	NP_001128686	NP_659496
##	0	0	0	0
##	NP_001395	NP_821133	NP_006079	NP_001060
##	0	0	0	0
##	NP_821080	NP_006078	NP_006077	NP_115914
##	0	0	0	0
##	NP_110400	NP_817124	NP_000009	NP_001029031
##	0	0	0	0
##	NP_055935	NP_060322	NP_001120727	NP_065812
##	4	0	0	0
##	NP_055486	NP_002825	NP_001840	NP_478054
##	_ 0	_ 0	_ 0	_ 0
##	NP 001631	NP 000273	NP 001171475	NP 003362
##	- 0	- 0	- 0	- 0
##	NP 001127870	NP 001001521	NP 006750	NP 060077
##	0	0	0	0
##	NP 001258547	NP 056969	NP 060686	NP 006245
##	0	0	0	0
##	NP_006248	NP_001229342	NP_001013725	NP_060731
##	N1 _000240 0	0	N1 _001013723	0 000731
##	NP_839955	NP_001024	NP_001005242	NP_004563
			NF_001003242	
##	0	0 ND 002074	ND 00110FF07	0 ND 065143
##	NP_001026859	NP_003971	NP_001185537	NP_065143
##	0	0	WD 007540	0 ND 007545
##	NP_079100	NP_001153576	NP_997543	NP_997545
##	0	0	0	3
##	NP_071451	NP_060469	NP_004078	NP_001191317
##	0	0	0	0
##	NP_065781	NP_066943	NP_001159750	NP_001136171
##	0	0	0	0
##	NP_001356	NP_003267	NP_001027454	NP_001027455
##	0	0	0	0
##	NP_079285	NP_002495	NP_060552	NP_003240
##	0	0	0	0
##	NP_001340	NP_001646	NP_919254	NP_055869

##	0	0	0	0
##	NP_004108	NP_004318	NP_068781	NP_001083
##	0	0	0	0
##	NP_001243776	NP_060391	NP_001136254	NP_001254703
##	0	0	0	0
##	NP_777547	NP 036475	NP_055925	NP_055642
##	- 0	- 0	- 0	_ 0
##	NP 001258604	NP 006201	NP_056178	NP_078934
##	N1_001250004	N1_000201	M1_000170	N1 _070334
	v	•		ND 001171500
##	NP_005678	NP_001171579	NP_001171577	NP_001171580
##	0	0	0	0
##	NP_001171586	NP_056013	NP_005004	NP_001191397
##	0	0	0	0
##	NP_001191395	NP_009005	NP_065857	NP_001161842
##	0	0	0	0
##	NP_055595	NP_055904	NP 001724	NP 057630
##	0	0	_ 0	_ 0
##	NP 057591	NP_003709	NP 112557	NP 001252
##	NF_057591	NF_003709	NF_112337	NF_001232
	•	v	ŭ	VD 004464076
##	NP_060418	NP_037446	NP_065116	NP_001161076
##	0	0	0	0
##	NP_036336	NP_001245218	NP_001164006	NP_114127
##	0	0	0	0
##	NP_006657	NP_631898	NP_932064	NP_109590
##	0	0	0	0
##	NP_068656	NP_001112	NP_058432	NP_005751
##	0	0	0	0
##	NP_037373	NP_000174	NP_000479	NP_001129101
##	0	0	0	0
##	NP 115518	NP 848537	NP 060141	NP 002970
##	_ 0	0	_ 0	_ 0
##	NP 060952	NP 003878	NP 004175	NP 002881
##	0	M _000070	M1_004170	N1_002001
	•	ND 004577	ND 001000000	ND OFFO11
##	NP_002944	NP_004577	NP_001006933	NP_055311
##	0	0	0	0
##	NP_031391	NP_006612	NP_001124192	NP_000678
##	0	0	0	0
##	NP_001124195	NP_001124194	NP_001684	NP_001683
##	0	0	0	0
##	NP_060137	NP_005348	NP_001547	NP_004386
##	0	0	0	0
##	NP_061862	NP_071401	NP_000293	NP_000604
##	0	0	0	0
##	NP_079199	NP_065147	NP_004514	NP_998754
##	_ 0	- 0	_ 0	_ 0
##	NP_976241	NP_061916	NP_056016	NP_001006635
##	0	0	0	0
##		NP_004388	NP_001258717	ND 001050710
	NP_055674	_		NP_001258719
##	6 ND 001260	0 ND 004400	0 ND 050104	VD 005700
##	NP_001369	NP_004402	NP_056104	NP_005766
##	0	0	0	0
##	NP_001018003	NP_065731	NP_001075	NP_001141
##	0	0	0	0
##	NP_001226	NP_004710	NP_008977	NP_004630

##	0	0	0	0
##	NP_055111	NP_778228	NP_001193723	NP_002883
##	0	0	0	0
##	NP_006383	NP_054872	NP_060060	NP_942595
##	0	0	0	0
##	NP_055726	NP_065908	NP_065761	NP_057332
##	0	_ 0	_ 0	0
##	NP_001124427	NP_001035879	NP_055844	NP_058519
##	0	0	_ 0	_ 0
##	NP_058518	NP_001116539	NP_001190181	NP_059347
##	0	0	0	0
##	NP_001157852	NP_001032405	NP_001157854	NP_006775
##	0	0	0	0
##	NP_079408	NP_000149	NP_079010	NP_001001924
##	0	0	0	0
##	NP_001001925	NP_065800	NP_001001931	NP_002874
##	0	0	0	0
##	NP_004448	NP_075266	NP_078880	NP_004715
##	0	0	0	0
##	NP_536858	NP_036450	NP_004127	NP_000680
##	0	0	0	0
##	NP_000681	NP_000683	NP_000684	NP_003879
##	WD 001103006	ND 001610	ND 005454	ND 050477
##	NP_001193826	NP_001610	NP_005151	NP_056477
## ##	NP_005410	NP_006535	NP_061155	ND 001150006
##	NF_005410	NF_000555	NF_001133	NP_001158886
##	NP_001158887	NP_002291	NP_659409	NP_002292
##	0	0	0	0
##	NP 149972	NP_573441	NP_055852	NP_071415
##	_ 0	_ 0	0	_ 0
##	NP 056456	NP_055529	NP_065939	NP_001009571
##	0	_ 0	_ 0	0
##	NP_003707	NP_899631	NP_064590	NP_570971
##	2	2	0	0
##	NP_003306	NP_001138238	NP_067022	NP_982261
##	0	0	0	0
##	NP_004530	NP_115984	NP_001159632	NP_001675
##	0	0	0	0
##	NP_001673	NP_001001331	NP_001001344	NP_056036
##	0	0	0	0
##	NP_003890	NP_004831	NP_001106984	NP_663788
##	0	0	0	0
##	NP_071348	NP_079052	NP_056201	NP_060597
##	0 ND 000440	0	0 ND 001011516	0 ND 001007030
##	NP_006448	NP_001243355	NP_001011516	NP_001007232
## ##	0 ND 055607	ND 004500	ND 001009707	ND 065174
##	NP_055697 0	NP_004588	NP_001098707	NP_065174
##			-	ND 055635
##	NP_116243 0	NP_001005337	NP_000290	NP_055635
##	NP_005419	NP_006104	NP_071757	NP_001193928
##	NF_005419 0	NF_000104 0	NF_0/1/5/ 0	NF_001193928
##	NP_001957	NP_001188306	NP_001173	NP_001189333
υπ	111 _001301	141 _001100000	141 _001110	111 _001103000

##	0	0	0	0
##	NP_115679	NP_001539	NP_001010987	NP_001008212
##	0	0	3	0
##	NP_004235	NP_981961	NP_115754	NP_001123528
##	0	0	0	0
##	NP_001182132	NP_001950	NP_009017	NP_065737
##	0	0	0	0
##	NP_055121	NP_001128471	NP_060362	NP_001026868
##	0	0	0	3
##	NP_006611	NP_001138679	NP_783328	NP_008823
##	0	0	0	26
##	NP_001193984	NP_060592	NP_277021	NP_277028
##	26	0	0	0
##	NP_076916	NP_004438	NP_060641	NP_114098
##	0	0	0	0
##	NP_001019837	NP_001076	NP_061917	NP_009223
##	0	0	0	0
##	NP_004026	NP_004028	NP_001020561	NP_000027
##	0	_ 0	0	_ 0
##	NP_115941	NP_733793	NP_006763	NP_006282
##	0	_ 0	- 8	_ 0
##	NP 001123577	NP 112729	NP_150229	NP_003184
##	0	_ 0	_ 26	_ 0
##	NP_057308	NP 065070	NP 001035847	NP 112604
##	_ 0	- 0	- 0	_ 0
##	NP 004491	NP_001139653	NP 001130033	NP_115971
##	_ 0	0	_ 0	_ 1
##	NP 004496	NP_003265	NP_002622	NP_001191676
##	9	- 0	- 0	_ 0
##	NP 078932	NP 061185	NP 003886	NP_001158213
##	_ 0	_ 0	0	0
##	NP 055555	NP 001407	NP 001958	NP 000509
##	_ 0	_ 0	0	_ 0
##	NP 000510	NP 000550	NP 000175	NP 005321
##	0	0	0	0
##	NP 597709	NP 001087240	NP 877419	NP 001248375
##	_ 0	0	_ 0	_ 0
##	NP_006431	NP_078878	NP_057280	NP_001354
##	_ 0	0	0	_ 0
##	NP 057216	NP 002705	NP_001018079	NP_001018077
##	0	_ 0	_ 0	0
##	NP 005913	NP_006517	NP 055465	NP_065704
##	0	0	14	17
##	NP 872321	NP_067092	NP_660281	NP_001254645
##	14	21	19	30
##	XP_003960003	NP_258429	NP_001230967	NP_057615
##	8	6	35	20
##	NP 612376	NP_001073886	NP_009069	NP_150376
##	35	0 0 0 10 7 3 0 0 0	NI _003003	NI _130370
##	NP 001070143	NP_612203	NP_001025168	NP_001034216
##	18	35	35	13
##	NP_065127	NP_001721	NP 004226	NP_005603
##	21	22	NI _004220	10
##	XP_001715117	XP_003960812	NP_001243100	NP_009180
11.11	VI _001110111	M _000300012	111 _001Z 1 0100	141 _003100

##	60	17	37	41
##	NP_699194	NP_003700	NP_060913	NP_067039
##	N1 _033134 6	28	M _000313	M _007039
##	NP_067678	NP_003421	NP_057354	NP_001291
##	N1 _007070 57	23	30	NI _001231 27
##	NP 872439	NP_116225	NP_872296	NP_078947
##	17	W _110223	23	N1_010541 7
##	NP_940882	NP_001201835	NP_001001520	NP_116020
##	22	16	N1 _001001020	0
##	NP 150091	NP_004485	NP_001119523	NP_066967
##	0	0	0	0
##	NP 057157	NP_001035026	NP_872346	NP_060537
##	0	0	0	0
##	NP_000932	NP_065779	NP 443066	NP_115612
##	0	_ 0	_ 0	_ 0
##	NP_001027005	NP_000284	NP_056093	NP_001243299
##	- 0	- 0	- 0	- 0
##	NP_891988	NP_000926	NP 075447	NP_001035889
##	_ 0	_ 0	_ 0	_ 0
##	NP_071900	NP_060248	NP_004706	NP_000246
##	16	0	0	0
##	NP_000359	NP_006386	NP_004159	NP_005624
##	0	0	0	0
##	NP_008870	NP_006801	NP_065796	NP_004689
##	0	0	0	0
##	NP_004704	NP_079229	NP_054768	NP_003168
##	0	0	0	0
##	NP_001128524	NP_060365	NP_000057	NP_775267
##	0	0	0	0
##	NP_775266	NP_057018	NP_061161	NP_005795
##	0	0	0	0
##	NP_647537	NP_067013	NP_001231827	NP_001157260
##	0	0	0	0
##	NP_001157262	NP_038477	NP_056157	NP_001265855
##	U ND 001065056	3 ND 000400	U 001120703	ND 001070000
##	NP_001265856	NP_060460	NP_001138783	NP_001072992
##	ND 0049E1	ND 001170011	ND 00201E	ND 026472
## ##	NP_004851 0	NP_001172011	NP_002015	NP_036473
##	NP_001001329	NP_001158136	NP_055998	NP_055927
##	NF_001001329 0	NF_001130130	NF_033998	NF_033921
##	NP_055790	NP_079093	NP_112196	NP_001121631
##	0	0	0	0
##	NP_060759	NP_001020119	NP_062826	NP_078772
##	0	0	_ 0	_ 0
##	NP_892017	NP_001070865	NP_001186210	NP_055848
##	- 0	_ 0	- 4	- 0
##	NP_115627	NP_694574	NP_001107604	NP_037473
##	_ 0	_ 0	- 0	_ 0
##	NP_006444	NP_058633	NP_001166958	NP_000908
##	- 0	- 0	- 0	_ 0
##	NP_001017962	NP_490597	NP_001186384	NP_031397
##	0	0	0	0
##	NP_002429	NP_060272	NP_957516	NP_001005333

##	^	0	0	^
##	0 ND 001000001	ND 001024704	UD 001002400	ND 000711
##	NP_001229291	NP_001034794	NP_001003408	NP_006711
##	33	0	0	0
##	NP_006449	NP_001123539	NP_004746	NP_003933
##	0	0	0	0
##	NP_002605	NP_689596	NP_001244210	NP_001026853
##	0	0	0	0
##	NP_056990	NP_148980	NP_056506	NP_958850
##	0	0	0	0
##	NP_056161	NP 055759	NP_599031	NP_599032
##	_ 0	_ 0	_ 0	_ 0
##	NP 001193946	NP 001018494	NP_055491	NP 056132
##	0	0	0	0
##	NP 065825	NP_596867	NP_391988	NP_000879
##	NF_003623	NF_590007	NF_591900	NF_000079
	· ·	U	VD 000056	VD 070570
##	NP_055523	NP_001093646	NP_000356	NP_073572
##	0	0	0	0
##	NP_003130	NP_001243546	NP_001243547	NP_060828
##	0	0	0	0
##	NP_005923	NP_056350	NP_001093894	NP_006313
##	0	0	0	0
##	NP_060458	NP_079426	NP_001258825	NP_001137474
##	0	0	0	0
##	NP 001896	NP_001096078	NP_054727	NP 000687
##	_ 0	- 6	- 0	- 0
##	NP 001317	NP 064551	NP 000081	NP_001107227
##	0	0	0	0
##	NP 060867	NP 005733	NP 112223	NP 079029
##	0 000007	NF_003733	NF_112225	NF_073023
	· ·	ND 005007	ND 050245	ND 000001
##	NP_002159	NP_005887	NP_059345	NP_006331
##	0	0	0	0
##	NP_001095886	NP_000884	NP_056004	NP_061926
##	0	0	0	0
##	NP_075067	NP_057733	NP_065899	NP_055568
##	0	0	0	0
##	NP_057507	NP_060082	NP_569829	NP_569831
##	23	0	0	0
##	NP_569827	NP_036275	NP_057306	NP_004309
##	_ 0	_ 0	_ 0	_ 0
##	NP_000628	NP_060802	NP_009173	NP_037396
##	0	_ 0	_ 0	2
##	NP_000959	NP_000267	NP_005531	NP_861447
##	0	0	0	0
##	NP_000161	NP_055211	NP_006793	NP_055684
		NF_033211 0	_	
##	0 ND 001107054	· ·	0 ND 001164000	0 ND 050047
##	NP_001127854	NP_001104595	NP_001164209	NP_056047
##	0	0	10	10
##	NP_060264	NP_037528	NP_004276	NP_001254490
##	0	0	0	0
##	NP_001254489	NP_001911	NP_008958	NP_000691
##	0	0	0	0
##	NP_694997	NP_055077	NP_065826	NP_004591
##	0	0	0	0
##	NP_060628	NP_060496	NP_056012	NP_001254500
	_	-	_	

	^	^		
##	0	U ND 004477057	0 0	U 000074
##	NP_036602	NP_001177957	NP_002387	NP_006671
##	0	0	0	0
##	NP_002638	NP_001229372	NP_006353	NP_057227
##	0	0	0	0
##	NP_037386	NP_054721	NP_055542	NP_001184033
##	0	2	0	0
##	NP 001129120	NP_001098714	NP_008981	NP_006339
##	- 0	7	- 49	_ 0
##	NP 075068	NP_002094	NP 068776	NP 060541
##	0	0	0	0
##	NP 008982	NP_620137	NP 003225	NP_003218
	-	NF_020137	NF_003225	_
##	0	U 077050	U	18
##	NP_612403	NP_277050	NP_078805	NP_004806
##	0	0	0	0
##	NP_003889	NP_065892	NP_065947	NP_005549
##	0	0	0	0
##	NP_001153682	NP_003696	NP_443202	NP_079291
##	0	0	0	0
##	NP 004550	NP 003642	NP_001138898	NP 057066
##	- 0	- 0	- 0	- 0
##	NP_001265367	NP_056007	NP_113654	NP 005580
##	0	0	M _110001	M _000000
##	NP 001265523	NP_005511	NP 062543	NP 004957
	-	-	-	NF_004957
##	0	0	0	0 0 0 0 0 0 0 0
##	NP_036339	NP_067676	NP_001144	NP_006640
##	0	0	0	0
##	NP_001159693	NP_067677	NP_000549	NP_006746
##	0	0	0	0
##	NP_003698	NP_004779	NP_001035932	NP_077006
##	0	0	0	0
##	NP_284941	NP_055689	NP_002849	NP_001116146
##	_ 0	_ 0	_ 0	_ 0
##	NP 006210	NP 001098662	NP 005069	NP 005068
##	0	0	0	0
##	NP 008936	NP 003251	NP 005134	NP 066275
##	M1 _000500	M1_000201	M1_000104	N1_000270
	ND OOG214	ND 060017	ND 005727	ND 070026
##	NP_006314	NP_068817	NP_005737	NP_078836
##	0	0	0	0
##	NP_757367	NP_000883	NP_219491	NP_004122
##	0	0	6	46
##	NP_116165	NP_005637	NP_000917	NP_065770
##	0	0	0	0
##	NP_004597	NP_722516	NP_055336	NP_006829
##	0	0	0	0
##	NP 003479	NP_003744	NP_009166	NP_075055
##	0	_ 0	_ 0	_ 0
##	NP_056348	NP_510965	NP_001258027	NP_055096
##	N1 _030340 0	N1_510505	MI_001200021	M1 _000030
			MD OFFOOF	VD 002200
##	NP_001258028	NP_060619	NP_955805	NP_003739
##	0	0	0	0
##	NP_000924	NP_006199	NP_001124436	NP_006216
##	0	0	0	0
##	NP_001124432	NP_055453	NP_055983	NP_064578

##	0	38	0	0
##	NP_002567	NP_002568	NP_001121640	NP_689749
##	0	0	0	0
##	NP 004085	NP 005403	NP_004160	NP 705831
##	_ 0	0	_ 0	0
##	NP 056370	NP_000553	NP_110517	NP 115609
##	- 0	- 0	- 0	- 0
##	NP 057657	NP_001258356	NP_001138771	NP_001138770
##	- 0	- 0	- 0	- 0
##	NP_056034	NP_054767	NP_065882	NP_001139784
##	0	_ 0	_ 0	3
##	NP_001257624	NP_001257628	NP_005920	NP_031394
##	0	0	3	0
##	NP_001001992	NP_001027582	NP_001193997	NP_001153239
##	0	0	0	0
##	NP_001096123	NP_005921	NP_001234919	NP_758441
##	0	0	0	0
##	NP_848656	NP_940897	NP_116258	NP_787072
##	0	0	0	0
##	NP_056249	NP_061854	NP_055953	NP_002363
##	0	0	0	0
##	NP_001093896	NP_001093897	NP_066982	NP_001093898
##	0	0	0	0
##	NP_066018	NP_004183	NP_055707	NP_065757
##	0	0	0	0
##	NP_036246	NP_005144	NP_055855	NP_009200
##	0	UD 070404	0 ND 007005	0
##	NP_001106177	NP_079191	NP_937825	NP_114414
##	0 ND 001100340	ND 001013640	000000 dw	ND 000021
## ##	NP_001128342	NP_001013649	NP_000889	NP_000231
##	NP 001139685	NP 073608	NP 055241	NP 689419
##	N1_001133003	N1 _073000	N1_035241 0	M1 _003419
##	NP 689973	NP 055495	NP 057019	NP 004302
##	0	0.000	0.007010	0
##	NP 079436	NP 001108585	NP 004272	NP 115890
##	0	0	0	0
##	NP_060555	NP_001121902	NP_000275	NP_001166927
##	- 0	- 0	- 0	- 0
##	NP_005381	NP_000377	NP_001159583	NP_001122089
##	0	_ 0	0	0
##	NP_114128	NP_004717	NP_001074444	NP_001157918
##	0	0	0	3
##	NP_443173	NP_940862	NP_997281	NP_003585
##	0	0	0	0
##	NP_055527	NP_055863	NP_542414	NP_056464
##	0	13	2	0
##	NP_061830	NP_963848	NP_116045	NP_004299
##	0	0	0	0
##	NP_079092	NP_065777	NP_060117	NP_003839
##	0	0	0	0
##	NP_653302	NP_002386	NP_060595	NP_055444
##	31	0 ND 055750	0 ND 055050	0
##	NP_056155	NP_055758	NP_055850	NP_000364

##	0	0	0	0
##	NP_005535	NP_055720	NP_001243239	NP_037399
##	0	0	0	8
##	NP_002331	NP_004821	NP_660202	NP_005427
##	0	0	0	0
##	NP 006586	NP_001136402	NP_055385	NP 115765
##	_ 0	- 0	_ 0	_ 0
##	NP 009032	NP 004725	NP 001182345	NP_835365
##	0	0	0	0
##	NP 001182359	NP 079523	NP_009208	NP 003397
##	NF_001102339 40	NF_079525	NF_009200	NF_003397
		U 000047	U	VD 000005
##	NP_006752	NP_006817	NP_036611	NP_003395
##	0	0	0	0
##	NP_003396	NP_006133	NP_002930	NP_006074
##	0	0	0	0
##	NP_006182	NP_037377	NP_004860	NP_001036227
##	0	0	0	0
##	NP_006591	NP_001728	NP_003452	NP_054774
##	0	0	0	0
##	NP 006203	NP 001018063	NP 004557	NP 004558
##	_ 0	_ 0	0	_ 0
##	NP_002616	NP_060598	NP 002477	NP 004199
##	0	M _000050	NI _002477	M1_004133
	NP_665811	•	ND 001620	ND 006764
##	-	NP_006654	NP_001632	NP_006764
##	0	0	0	0
##	NP_006293	NP_892116	NP_446464	NP_001138890
##	0	0	0	0
##	NP_077740	NP_077741	NP_002194	NP_000421
##	0	3	0	0
##	NP_068756	NP_004896	NP_001191354	NP_000114
##	0	0	0	0
##	NP_001970	NP_001243412	NP_060522	NP_653205
##	0	0	0	0
##	NP 073600	NP 001867	NP 001138608	NP_001229804
##	_ 0	_ 0	9	_ 0
##	NP 001116142	NP 000283	NP 006102	NP 055617
##	0	0	0	0
##	NP_997226	NP_001265178	NP_003926	NP_004609
		NF_001203178	NF_003920	•
##	0 ND 050411	·	•	ND 00E130
##	NP_258411	NP_037397	NP_001006658	NP_005130
##	0	0	0	0
##	XP_003960256	NP_004721	NP_005683	NP_004243
##	0	0	0	0
##	NP_001025053	NP_058197	NP_002526	NP_078966
##	0	0	0	0
##	XP_003960891	NP_114437	NP_003602	NP_055970
##	17	0	0	0
##	NP_001238817	NP_060102	NP_112483	NP_001164121
##	- 0	_ 0	_ 0	_ 0
##	NP_005722	NP_066278	NP_001258662	NP_057732
##	0	0	0	0
##	NP_001120792	NP_055662	NP_001245321	NP_006319
##	0	N1 _000002	0 001240321	U _000013
	_	-	•	۷D ۸۸6710
##	NP_001185765	NP_003144	NP_001171551	NP_006712

##	0	0	0	0
##	NP_060634	NP_001186848	NP_056414	NP_000706
##	0	0	0	0
##	NP_002708	NP_060931	NP_858061	NP_870991
##	0	0	0	0
##	NP 006100	NP 036448	NP 002255	NP_002260
##	_ 0	0	_ 0	_ 0
##	NP_056449	NP 068758	NP 003095	NP_001489
##	M1_000110	M1_000100	M _000000	M _001100
##	NP_001027392	NP 004703	NP 005952	NP_954712
##	N1_001027332	N1 _004705	NI _003332	N1 _304/12
##	ND 005710	ND 065170		ND 002177
	NP_005712	NP_065178	NP_057371	NP_003177
##	U	0	0	0
##	NP_056365	NP_663160	NP_663161	NP_115912
##	0	0	0	5
##	NP_149124	NP_071387	NP_006238	NP_001230280
##	0	0	0	0
##	NP_066956	NP_000052	NP_004521	NP_001121363
##	0	0	0	0
##	NP_056257	NP_848927	NP_006256	NP_001130038
##	0	0	0	0
##	NP_009034	NP_068761	NP_001091981	NP_001230882
##	0	0	0	0
##	NP 006836	NP 115948	NP_002197	NP_001138468
##	- 0	_ 0	- 2	- 0
##	NP 066952	NP_789845	NP 789842	NP_004416
##	0	0	0	0
##	NP 060649	NP_001001936	NP 115939	NP_000017
##	M1_000010	0	M _110000	M1_000017
##	NP 004452	NP_073613	NP_060254	NP_005122
##	NF_004432	NF_073013	NF_000254	NF_003122
	ND OOOO	ND OOGOAO	ND 001001000	ND 055427
##	NP_008824	NP_006249	NP_001091982	NP_055437
##	0	0	0	VD 005050
##	NP_659400	NP_060224	NP_000958	NP_005052
##	0	0	0	0
##	NP_065387	NP_004434	NP_004435	NP_004433
##	13	0	0	3
##	NP_004431	NP_004429	NP_004432	NP_115285
##	6	11	3	0
##	NP_001073917	NP_004422	NP_003650	NP_066576
##	11	11	0	0
##	NP_004678	NP_003492	NP_001245238	NP_001244066
##	45	0	0	0
##	NP_008945	NP_008944	NP_001257384	NP_001245221
##	0	0	3	0
##	NP_001245224	NP_004545	NP_001129494	NP_001265598
##	0	0	0	0
##	NP_775188	NP_056126	NP_003715	NP_000848
##	5	0	51	0
##	NP 009215	NP_002813	NP_056975	NP_001106849
##	0	0	0	0
##	NP_079483	NP_001171821	NP_000523	NP_055929
##	0 NI _075400	0	M _000020	M1_000323
##	NP_036561	NP_203740	NP_777635	NP_001069
	101 000001	MI _ ZUU / 4U	ME_111033	ML _001003

	_	_	_	_
##	0	0	0	0
##	NP_001186763	NP_644810	NP_006400	NP_005711
##	0	0	0	0
##	NP_001177925	NP_000110	NP_001020366	NP_001257
##	0	0	0	0
##	NP 060504	NP 612398	NP 001010872	NP_001231513
##	_ 0	_ 0	- 6	- 0
##	NP 057103	NP 001257572	NP 958815	NP_932068
##	N1 _007 100 0	N1_001257572	NI _550015	N1 _332000
	•	•	ND 440427	ND 002700
##	NP_000746	NP_110430	NP_110437	NP_003728
##	0	0	0	8
##	NP_066963	NP_001245185	NP_005808	NP_001157661
##	0	0	0	0
##	NP_001008707	NP_001180197	NP_001186601	NP_001186600
##	0	0	0	0
##	NP_004032	NP_064647	NP 004303	NP 000532
##	_ 0	_ 0	_ 0	3
##	NP_001258137	NP 006835	NP 071904	NP 008855
##	NF_001230137	NF_000033	NF_071304	NF_000033
	•	VD 000007	U	VD 055400
##	NP_003127	NP_060327	NP_000134	NP_055192
##	0	0	0	0
##	NP_002573	NP_006651	NP_001132	NP_066015
##	0	0	0	0
##	NP_003106	NP_997192	NP_002817	NP_443711
##	0	0	0	0
##	NP_085058	NP 071353	NP_006038	NP_001175
##	_ 0	_ 0	_ 0	3
##	NP 003669	NP 004227	NP 056417	NP_001254507
##	_ 0	_ 0	_ 0	0
##	NP 659471	NP_006614	NP 001258752	NP_001258751
##	M _0054/1	11 _00001 TM	M1_001200702	N1_001200701
	ŭ	U 00110000	ND 0040E0	ND 004077
##	NP_001258753	NP_001182220	NP_004250	NP_004277
##	0	0	0	0
##	NP_009190	NP_001266284	NP_001266282	NP_001677
##	0	0	0	0
##	NP_003648	NP_008996	NP_000195	NP_775853
##	0	0	0	0
##	NP_000276	NP_057022	NP_001070	NP_006531
##	0	0	0	0
##	NP_001092104	NP_061900	NP_659416	NP_001185748
##	- 0	_ 0	_ 0	_ 0
##	NP_783161	NP_001002909	NP_857597	NP_919226
##	0	0	11	14
##	NP_757366	NP_115949	NP_001116436	NP_056150
##	0	N1_110343	M1_001110 4 50	N1 _030130 0
		ND 001065557	ND 000160	
##	NP_005112	NP_001265557	NP_000169	NP_116021
##	0	0	0	0
##	NP_075045	NP_001238936	NP_064529	NP_054868
##	0	0	0	0
##	NP_000265	NP_001165285	NP_056274	NP_001129145
##	0	0	0	0
##	NP_056999	NP_542775	NP_004732	NP_057383
##	6	0	0	0
##	NP_057421	NP_001230022	NP_004893	NP_001070819
	_	=	_	_

0	0	0	0
NP_003053	NP_059119	NP_071330	NP_060742
0	0	0	11
NP_006337	NP_001124495	NP_001124496	NP_000310
0	0	0	0
NP_065890	NP_002423	NP_060764	NP_002766
0	0	0	0
NP_003293	NP_000403	NP_001073964	XP_003403435
0	0	0	0
NP_055399	NP_063944	NP_005794	NP_078994
0	3	0	0
NP_006364	NP_055615	NP_877496	NP_002729
0	0	0	0
NP_997700	NP_002728	NP_002730	NP_001185688
_ 0	_ 0	5	- 0
NP 002340	NP 065098	NP 001243423	NP 001243424
_ 0	_ 0	- 0	- 0
NP 000010	NP 004377	NP 071919	NP_006803
- 0	51	_ 0	- 0
NP 001248349	NP 003090	NP 001229862	NP 003091
0	0	_ 0	_ 0
NP 001265128	NP 056285	NP 057715	NP_037461
0	0	0	0
NP 000997	NP 001254628	NP 002204	NP 000202
_	0	0	0
•	NP 078824	NP 055445	NP 002942
-	0	0	0
NP 061887	NP 055430	NP 003900	NP 705898
-	0	0	0
•	NP 065990	NP 705899	NP 006023
-	0	0	0
•	NP 004985	NP 001686	NP 065750
-	0.001000	0	0
•	NP 001248763	NP 005849	NP 976240
-	M1_001240700	M1 _000049	N1_570240
ŭ	NP 000424	NP 055055	NP 005727
M1 _000009	N1_000424	M _000000	N1 _000121
NP 005726	NP 055227	NP 004220	NP_061330
_	_		0
-	· ·	•	NP_060681
	_		3
	•	•	NP_057297
	_	_	N1 _031231
	_	-	NP_001073333
			001073333
·	•	-	NP_060206
_	_		NF_000200
-	•	-	NP 006497
_	_	NF_000949	NF_000497
	•	ND 0011270F4	ND OOGSSS
_	_	_	NP_006333
	-	·	ND 055134
_	_		NP_055134
-	· ·	-	ND 000700
NP_002736	NP_002/3/	NP_002/39	NP_002738
	NP_003053	NP_003053 NP_059119 0 0 NP_006337 NP_001124495 0 0 NP_065890 NP_002423 0 0 NP_003293 NP_000403 0 0 NP_055399 NP_063944 0 0 NP_997700 NP_055615 0 0 NP_002340 NP_065098 0 0 NP_000340 NP_065098 0 0 NP_0004377 0 0 0 NP_001248349 NP_003090 0 0 NP_001265128 NP_056285 0 0 NP_000880 NP_078824 0 0 NP_061887 NP_055430 0 0 NP_689940 NP_055990 0 0 NP_06130 NP_001248763 0 0 NP_055899 NP_0055227	NP_003053 NP_059119 NP_071330 0 0 0 0 NP_006337 NP_001124495 NP_001124495 NP_001124495 0 0 0 0 NP_065890 NP_002423 NP_060764 0 0 0 0 NP_003293 NP_000403 NP_001073964 0 0 0 0 3 0 0 0 NP_055399 NP_063944 NP_087796 0 0 0 0 NP_006364 NP_055615 NP_877496 0 0 0 0 NP_997700 NP_002728 NP_002738 NP_002730 0 5 NP_002340 NP_065098 NP_001243423 0 0 0 NP_002340 NP_065098 NP_001243423 0 0 0 0 NP_002340 NP_065098 NP_001243423 NP_071919 0 0 0 0 0 0 0 0 0

	^	^	^	•
##	0	U 000400	0	U 004400007
##	NP_006843	NP_036422	NP_001130026	NP_001130027
##	U ND 055504	VP 005000	UD 000054	U 000445
##	NP_055581	NP_065883	NP_060651	NP_009145
##	0	0	18	VD 055475
##	NP_006136	NP_008965	NP_001002762	NP_055475
##	0	0	0	0
##	NP_005026	NP_079440	NP_000032	NP_005038
##	0	0	0	0
##	NP_001036010	NP_037518	NP_066974	NP_065173
##	0	0	0	0
##	NP_055398	NP_060941	NP_694540	NP_076427
##	0	0	0	0
##	NP_036433	NP_001061	NP_057521	NP_002797
##	0	0	0	0
##	NP_001186127	NP_036232	NP_001116295	NP_004675
##	0	0	0	0
##	NP_000544	NP_055483	NP_001136438	NP_001862
##	3	0	42	0
##	NP_004506	NP_006342	NP_060238	NP_004231
##	0	0	0	0
##	NP_002584	NP_115543	NP_001129123	NP_542196
##	0	0	0	0
##	NP_000084	NP_001265003	NP_542411	NP_009185
##	0	0	37	0
##	NP_000921	NP_127509	NP_055181	NP_001180414
##	0	0	0	0
##	NP_001180412	NP_001180413	NP_004978	NP_001154875
##	0	0	0	0
##	NP_001154876	NP_006236	NP_001155198	NP_848701
##	0	0	0	0
##	NP_848702	NP_006235	NP_004586	NP_071735
##	0	2	0	0
##	NP_878256	NP_005491	NP_009217	NP_001159681
##	0	0	0	0
##	NP_005852	NP_757386	NP_036342	NP_036377
##	0	0	0	0
##	NP_001035536	NP_003932	NP_055319	NP_000243
##	0	0	0	0
##	NP_057240	NP_958435	NP_036233	NP_057396
##	0	0	0	0
##	NP_061138	NP_000689	NP_003795	NP_066565
##	0	0	0	0
##	NP_004799	NP_055946	NP_036243	NP_005017
##	0	0	0	0
##	NP_892021	NP_055940	NP_055746	NP_055975
##	_ 0	_ 0	_ 0	_ 0
##	NP_699160	NP_001530	NP_001117	NP_689541
##	0	_ 0	0	0
##	NP_004955	NP_001518	NP_116185	NP_060855
##	0	0	0	0
##	NP_055263	NP_006099	NP_055830	NP_000659
##	0	0	0	0
##	NP_000660	NP_000661	NP_003356	NP_056099

##	0	0	0	3
##	NP_001180452	NP_055761	NP_112740	NP_002129
##	0	0	0	0
##	NP_001003810	NP_004490	NP_055787	NP_004676
##	0	0	0	0
##	NP_001538	NP_055449	NP_079239	NP_005754
##	0	0	0	0
##	NP 073738	NP 150634	NP_001017534	NP_037387
##	0	0	0	0
##	NP_001138826	NP_065385	NP 072096	NP_036552
##	0	0.00000	M1_072000	M _000002
##	XP_003960677	NP_078801	NP 075526	NP_060045
	_	NF_078801 4	NF_075520	_
##	0	-	0 00000	6 ND 004700
##	NP_001014447	NP_001014448	NP_000013	NP_001702
##	5	5	0	0
##	NP_001028229	NP_056335	NP_001137307	NP_001026855
##	0	0	0	0
##	NP_005899	NP_079120	NP_000241	NP_000493
##	0	0	0	19
##	NP_056189	NP_115958	NP_002212	NP_002211
##	0	0	0	42
##	NP 060493	NP_006687	NP 004389	NP_005442
##	0	0	0	0
##	NP 976227	NP_998801	NP 056377	NP_036364
##	0	0	M _000077	11 _00004
##	NP 066961	ŭ	· ·	ND 00102E020
	NF_000901	NP_057575	NP_002769	NP_001035930
##	ŭ	ND 005000	U 000400	ND 071075
##	NP_001035931	NP_005902	NP_000420	NP_071375
##	0	0	0	0
##	NP_001441	NP_061156	NP_940869	NP_060834
##	0	0	0	0
##	NP_149079	NP_065755	NP_002620	NP_000281
##	0	0	0	0
##	NP_001025062	NP_078849	NP_001244942	NP_001244944
##	0	0	0	0
##	NP_006357	NP_000475	NP_001129603	NP_006035
##	0	0	0	_ 0
##	NP_149077	NP_005498	NP_619579	NP_003070
##	0	0	0	0
##	NP_000512	NP_065991	NP_073586	NP_001180449
##	0	0	0	3
		-	-	
##	NP_002071	NP_115824	NP_000099	NP_003189
##	0	0	0	0 004440
##	NP_006556	NP_001143	NP_001627	NP_001142
##	0	0	0	0
##	NP_112581	NP_003749	NP_002892	NP_940916
##	0	0	0	0
##	NP_056277	NP_009097	NP_055656	NP_002310
##	0	0	0	0
##	NP_036244	NP_068751	NP_060627	NP_872363
##	0	0	3	- 0
##	NP_001074324	NP_001230685	NP_079433	NP_057175
##	- 0	- 0	_ 0	- 0
##	NP_775733	NP_001822	NP_078911	NP_683706
	, , 0, 00	001022	0,0011	000700

		^	^	
##	1	0	0	VP 05000
##	NP_078862	NP_000780	NP_006759	NP_056062
##	0	0	0	0
##	NP_001153171	NP_001153173	NP_001153174	NP_001153175
##	0	0	0	0
##	NP 079255	NP 689956	NP_003896	NP_000312
##	- 0	- 0	- 0	_ 0
##	NP 001015001	NP 001816	XP 003960636	NP 071741
	-	-	AF_003900030	NF_0/1/41
##	0	0	VD 000057	0
##	NP_004688	NP_009103	NP_002257	NP_002703
##	0	0	0	0
##	NP_036335	NP_055392	NP_056510	NP_001900
##	0	0	0	0
##	NP 006812	NP 001032238	NP_689544	NP 116246
##	- 0	_ 0	- 0	_ 0
##	NP 060457	NP 001032519	NP 001124181	NP 694955
##	_	N1 _001032313	NI _001124101	M1 _034333
	0	VD 004005454	U	0
##	NP_001035882	NP_001265154	NP_006378	NP_788276
##	0	0	0	0
##	NP_001244904	NP_003864	NP_005509	NP_001091742
##	0	0	0	0
##	NP 003841	NP_001229281	NP_001020376	NP 775767
##	_ 0	_ 0	_ 0	_ 0
##	NP_009135	NP 005745	NP_056211	NP 001253970
##	0	0 000	0	M1_001200010
	· ·	•	v	ND 446004
##	NP_054873	NP_004444	NP_006089	NP_116294
##	0	0	0	0
##	NP_002254	NP_001559	NP_001243307	NP_000022
##	0	0	0	0
##	NP_004887	NP_443107	NP_006273	NP_006272
##	0	0	0	0
##	NP 001129490	NP 938074	NP 062828	NP 001814
##	_ 0	0	_ 0	_ 0
##	NP 001815	NP 001512	NP 036457	NP 036458
	N1 _001013	MI _001512	N1 _030 1 37	N1 _030 1 30
##	· ·	ND 025027	U 0010120C0	ND 00000E
##	NP_115687	NP_835237	NP_001013269	NP_002385
##	0	0	0	0
##	NP_112506	NP_775752	NP_000128	NP_006416
##	0	0	0	0
##	NP_001139542	NP_996790	NP_006243	NP_055462
##	0	0	0	0
##	NP_113678	NP_001344	NP_001240837	NP_001345
##	_ 0	_ 0	_ 0	_ 0
##	XP_003960903	NP_001809	NP_001177836	NP_036466
##	_	M1_001005	M1_0001111000	M1 _000400
	0 ND 004300	O O O O O O	ND 00000	O 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
##	NP_004390	NP_690006	NP_006360	NP_003818
##	37	0	0	0
##	NP_071363	NP_001093326	NP_005227	NP_004665
##	0	0	0	0
##	NP_001248761	NP_004619	NP_036218	NP_001171114
##	_ 0	- 0	- 0	_ 0
##	NP_002148	NP_056202	NP_001094289	NP_004148
##	0	0	0	001140
	•	· ·	-	ט מען
##	NP_002727	NP_062547	NP_057541	NP_002733

##	0	0	0	0
##	NP_005804	NP_003162	NP_000998	NP_000999
##	0	0	0	0
##	NP_002056	NP_001035919	NP_004131	NP_001186110
##	0	0	0	0
##	NP_640334	NP_055271	NP_001909	NP_060216
##	0	0	0	0
##	NP_001164187	NP_001164188	NP_001159450 19	NP_001120982
## ##	NP_062817	NP 001243731	NP_001608	NP_689942
##	0	M001240701	M1_001000	N1_003542
##	NP 001182484	NP_055481	NP 115825	NP 003361
##	0	_ 0	_ 0	0
##	NP_005040	NP_001073901	NP_001212	NP_742126
##	0	0	0	0
##	NP_751911	NP_751913	NP_001191421	NP_001213
##	0	0	0	0
##	NP_742076	NP_741960	NP_001136112	NP_055309
##	0 ND 001100200	0 ND 00100E	0 ND 000004	8 ND 00010
## ##	NP_001189399	NP_001005	NP_006694	NP_068810
##	NP_001138610	NP_002950	NP_115515	NP 060422
##	0 001130010	N1 _002330	N1_110010	N1 _000422
##	NP 003316	NP_061819	NP 075556	NP_665875
##	0	0	0	0
##	NP_001003954	NP_001124163	NP_000935	NP_001135825
##	0	0	0	0
##	NP_001230903	NP_066025	NP_001164161	NP_689966
##	0	0	0	8
##	NP_766638	NP_758962	NP_004987	NP_003777
##	ND 064603	ND 000343	0 ND 055075	10 ND 445003
## ##	NP_064693 19	NP_000343 23	NP_055875	NP_115683
##	NP 004466	NP 001171900	NP 005154	NP 001617
##	0	M1_0011/1300	M1_000104	M1_001017
##	NP 005456	NP 031372	NP 817092	NP 065897
##	_ 0	_ 0	_ 0	- 0
##	NP_057023	NP_000043	NP_000044	NP_002083
##	0	4	57	0
##	NP_001091947	NP_003463	NP_060222	NP_001095896
##	0	0	0	2
##	NP_001077362	NP_055389	NP_001034966	NP_056244
## ##	0 ND 006254	0 NP 788955	0 NP_003176	0 NP_005631
##	NP_006254 0	NF_788933	NF_003176	NF_003031
##	NP_001032242	NP_004505	NP_001032248	NP_001028755
##	0	0	0	0
##	NP_005484	NP_065901	NP_001334	NP_001231800
##	_ 0	_ 0	_ 0	- 0
##	NP_001985	NP_002637	NP_003959	NP_937838
##	0	0	0	0
##	NP_004921	NP_001193469	NP_085133	NP_076417
##	0	0	0	0
##	NP_001427	NP_037522	NP_001290	NP_085059

##	0	3	0	0
##	NP_061174	NP_000259	NP_861966	NP_001091896
##	0	0	0	0
##	NP_003067	NP_001003801	NP_006581	NP_001243655
##	0	0	0	0
##	NP 001243657	NP 057205	NP_612401	NP_056474
##	0	_ 0	4	_ 0
##	NP_001664	NP 000440	NP 002208	NP_001241647
##	0	0	0	48
##	NP 001241652	NP 002716	NP_003923	NP_057213
##	48	0	0	0
##	NP_068799	NP_036614	NP 073743	NP_872294
##	NF_0007 <i>99</i>	NF_030014	NF_0/3/43	NF_072294 16
	•	ND OFCESC	ND 001000607	
##	NP_079114	NP_056536	NP_001229687	NP_005505
##	0	0	0	0
##	NP_002107	NP_002108	NP_001229971	NP_005507
##	0	0	0	0
##	NP_001091949	XP_003961043	NP_002118	NP_001158
##	0	0	0	0
##	NP_001265362	NP_002726	NP_689799	NP_036585
##	0	0	3	0
##	NP_064560	NP_109587	NP_002746	NP_115756
##	0	0	0	0
##	NP_001009555	NP_001122395	NP_055412	NP_003672
##	0	0	_ 0	- 0
##	NP_001074295	NP_056513	NP_055956	NP 003785
##	0	0	0	_ 0
##	NP 001217	NP 003746	NP 001180289	NP 056355
##	0	0	0	0
##	NP 008882	NP 037365	NP 005100	NP 612565
##	0	0007000	M1_000100	M1_012000
##	NP 004847	NP 057419	NP 006506	NP 002427
	NF_004047	NF_05/419	NF_000500	NF_002427
##	· ·	ND 001014426	ND 054700	ND 001116400
##	NP_001159933	NP_001014436	NP_054782	NP_001116428
##	VD 005550	VD 0000E0	000000	VD 004670
##	NP_005558	NP_060650	NP_002328	NP_004679
##	0	0	0	0
##	NP_004854	NP_714914	NP_001186011	NP_068813
##	0	0	0	0
##	NP_061170	NP_004901	NP_004146	NP_109591
##	2	0	0	0
##	NP_002631	NP_006208	NP_060392	NP_065799
##	0	0	0	0
##	NP_057422	NP_055620	NP_005079	NP_068800
##	0	0	0	0
##	NP_005909	NP_690869	NP_001128647	NP_055204
##	0	0	0	0
##	NP_001186170	NP_001028170	NP_006008	NP_001139322
##	0	0	_ 0	0
##	NP_003622	XP_003960526	NP_000119	NP_079478
##	0	19	_ 0	_ 0
##	NP_077305	NP_005914	NP_001001671	NP_065994
##	0	0	11	0
##	NP_001073981	NP_004752	NP_006257	NP_001034437
	0010,0001	001102	000201	001001101

##	5	0	16	0
##	NP_115618	NP_059984	NP_065962	NP_079455
##	3	8	21	9
##	NP 115544	NP_079366	NP_001619	NP_697021
##	0	0	0	0
##	NP_057632	NP 079425	NP_000391	NP_009194
##	0	0	0	0
##	NP 079477	NP 002087	NP_004351	NP_001784
##	0	0	0	3
##	NP 006525	NP 005871	NP 055428	NP_006027
##	0	0	0	0
##	NP 443111	NP_954658	NP_055772	NP_001029288
##	0	0	0	0
##	NP 056451	NP 001073888	NP_001180317	NP 005267
##	0	0	2	0
##	NP_006028	NP 001015053	NP_000963	NP 072047
##	0	0	0	0
##	NP 057085	NP_076968	NP_005266	NP_742068
##	0	_ 0	_ 0	_ 0
##	NP 001094058	NP_997657	NP_005479	NP_001129204
##	2	_ 0	- 0	- 0
##	NP 001129201	NP 115791	NP_434700	NP_001248319
##	_ 0	- 2	- 0	- 0
##	NP 001248324	NP 060688	NP_001406	NP_002070
##	0	_ 0	_ 0	_ 0
##	NP_065804	NP_065815	NP_000304	NP_004508
##	0	30	0	_ 0
##	NP_001265370	NP_001139110	NP_037428	NP_001219
##	0	0	11	0
##	NP_001073594	NP_001388	NP_001124154	NP_690902
##	0	0	0	0
##	NP_004531	NP_065802	NP_002171	NP_062552
##	2	0	0	0
##	NP_001230209	NP_000072	NP_037417	NP_443174
##	0	6	0	0
##	NP_057300	NP_001078846	NP_002378	NP_003787
##	0	0	0	0
##	NP_001239570	NP_002887	NP_113680	NP_001185774
##	0	0	0	0
##	NP_001185773	NP_001185772	NP_003132	NP_001123512
##	0	0	0	0
##	NP_892113	NP_003020	NP_001189788	NP_976224
##	0	0	0	14
##	NP_036567	NP_079170	NP_001071	NP_005572
##	5	0	0	0
##	NP_057159	NP_015561	NP_003615	NP_073564
##	0	0	0	0
##	NP_000662	NP_002371	NP_085072	NP_060150
##	0	0	0	0
##	NP_002684	NP_005752	NP_001026895	NP_714916
##	0 ND 001101600	0 ND 001177317	ND 001107654	ND 004404400
##	NP_001121699	NP_001177317	NP_001107654	NP_001121100
##	MD 003030	ND 90010E	ND 001007560	ND 005075
##	NP_003828	NP_899195	NP_001007562	NP_005875

##	0	0	0	0
##	NP_064553	NP_065074	NP_001124384	NP_006695
##	6	0	0	0
##	NP_001036038	NP_001020766	NP_055262	NP_037368
## ##	NP_001171846	NP 004298	NP_663722	0 NP_001002019
##	0-11/1040	N1 _004230	N1 _003722	N1_001002019
##	NP 079491	NP 001266	NP_004306	NP_001070020
##	- 0	5	_ 0	- 0
##	NP_004300	NP_149112	NP_997237	NP_689558
##	0	0	0	3
##	NP_066301	NP_003147	NP_001162589	NP_057737
##	0 ND 500407	0 ND 056156	ND 070001	ND 079020
## ##	NP_598407	NP_056156	NP_078881	NP_078939
##	NP 077734	NP 037395	NP 004929	NP_001339
##	0	0	0	0
##	NP_055141	NP_000145	NP_055630	NP_001127696
##	0	0	0	0
##	NP_055958	NP_065881	NP_001973	NP_001248340
##	0	0	1	0
##	NP_056520	NP_065823	NP_003440	NP_808821
## ##	0 NP_002697	0 NP 001028729	0 NP 808907	0 NP_036256
##	NF_002097 0	NF_001028729	NF_000907	NF_030230
##	NP_005383	NP_055922	NP 001171826	NP_055299
##	0	0	0	0
##	NP_001099010	NP_003637	NP_004708	NP_001095124
##	0	0	2	0
##	NP_036230	NP_004664	NP_114123	NP_002258
##	0	18	48	0 ND 007004
## ##	NP_002259	NP_078920	NP_003555	NP_037391
##	NP 001129064	NP 001029102	NP_001186226	NP 005376
##	0	0	0	0
##	NP_000526	NP_872578	NP_056516	NP_786924
##	0	0	0	0
##	NP_060495	NP_003357	NP_009109	NP_001104262
##	0	0	0	0
##	NP_001156907	NP_004535	NP_060341	NP_001001976
## ##	0 ND 009073	0 ND 005009	ND 064520	0 NP_852608
##	NP_008972 0	NP_005908	NP_064520 0	NF_652606
##	NP 001017973	NP_061848	NP_073713	NP 001320
##	0	_ 0	_ 0	_ 0
##	NP_001012632	NP_001319	NP_002309	NP_112180
##	0	0	0	0
##	NP_001185951	NP_056127	NP_055177	NP_689929
##	3	7	0	0
##	NP_116584	NP_004626	NP_004025	NP_001240752
## ##	0 NP_001113	NP_001012533	0 NP_005008	NP_001156736
##	NF_001113 0	NF_001012555	NF_003008	U _001130130
##	NP_036361	NP_060939	NP_001034662	NP_006566

	_	_		
##	0	3	0	0
##	NP_060796	NP_001231001	NP_689647	NP_004541
##	0	11	0	0
##	NP_003819	NP_001138422	NP_000433	NP_060501
##	0	0	0	0
##	NP_055659	NP_004805	NP_006747	NP_003186
##	0	0	0	0
##	NP 004451	NP 859063	NP_056017	NP_071347
##	- 0	- 0	- 5	_ 0
##	NP_060679	NP_006092	NP_065132	NP 002694
##	0	2	0	0
##	NP 001185885	NP 001185886	NP 001027467	NP 057626
##	0	N1_001105000	N1 _001021401	N1 _037 020
	ŭ	ND 0010E000C	U 000000	VD 000400
##	NP_001035917	NP_001258906	NP_002809	NP_000402
##	0	0	0	0
##	NP_067064	NP_001135869	NP_055587	NP_002452
##	0	0	0	0
##	NP_061167	NP_149094	NP_057381	NP_054859
##	0	0	0	0
##	NP_110395	NP_003694	NP_002305	NP_038460
##	0	0	0	0
##	NP_009297	NP_005148	NP_009298	NP_056541
##	_ 0	_ 0	3	_ 0
##	NP_001017930	NP 001130005	NP 777567	NP 588615
##	18	17	_ 0	2
##	NP 060439	NP_001138303	NP 057322	NP_005735
##	0	0	0	0
##	NP 055487	NP 004374	NP 002369	NP 647611
##	NF_033487	NF_004374	NF_002309	10
	· ·	ND 001161006		
##	NP_055857	NP_001161206	NP_000088	NP_683877
##	3	0	0	0
##	NP_001491	NP_005085	NP_940982	NP_005877
##	0	0	0	0
##	NP_001269	NP_055735	NP_001229314	NP_954592
##	0	8	14	0
##	NP_058632	NP_001248337	NP_005990	NP_055246
##	3	5	0	0
##	NP_060289	NP_003891	NP_002871	NP_001645
##	0	0	0	0
##	NP_001243125	NP_004324	NP_006779	NP_061877
##	- 0	_ 0	_ 0	_ 0
##	NP_058431	NP_005197	NP_004625	NP_997198
##	0	0	0	0
##	NP_002035	NP_001473	NP_150644	NP_036356
##	M1_002000	M1_001410	0	1
	-	MD 071020	-	ND 0E6303
##	NP_001186326	NP_071938	NP_036468	NP_056393
##	1	0	3	0
##	NP_653304	NP_000261	NP_078941	NP_005638
##	0	0	0	0
##	NP_150600	NP_001077083	NP_443134	NP_057321
##	0	0	0	0
##	NP_631961	NP_004951	NP_002604	NP_055531
##	0	0	0	0
##	NP_001245308	NP_002330	NP_005409	NP_056504

	_	_	_	_
##	0	0	0	5
##	NP_008998	NP_115564	NP_006452	NP_056076
##	0	0	6	0
##	NP_444505	NP_061955	NP_057025	NP_001171501
##	0	0	0	0
##	NP_973726	NP_653081	NP_005326	NP_085046
##	0	0	0	0
##	NP_001001851	NP_079197	NP_001185824	NP_059122
##	0	0	0	0
##	NP 001005743	NP 004747	NP 005408	NP 005424
##	- 0	_ 0	_ 0	_ 0
##	NP_005347	NP 002028	NP 835224	NP 003636
##	3	_ 0	0	0
##	NP_001123493	NP 001123492	NP 444271	NP_115759
##	0	0	0	0
##	NP_612453	NP_065819	NP_938167	NP 000333
##	N1_012400	M1_000019	M1 _5500107	M1_000000
##	NP_079032	NP_065797	NP 066553	NP_001001503
##	NF_079032	NF_003797	NF_000555	NF_001001303
##	NP_055186	NP_055815		ND 0010402E7
	_		NP_009202	NP_001248357
##	0	0	0 0.00	VD 000045
##	NP_055461	NP_057164	NP_116219	NP_060615
##	0	0	0	0
##	NP_005179	NP_733762	NP_055051	NP_003175
##	0	0	3	0
##	NP_001092094	NP_001180207	NP_036348	NP_001130046
##	0	0	20	0
##	NP_056216	NP_002010	NP_001153392	NP_001041659
##	0	0	2	0
##	NP_859048	NP_005800	NP_006397	NP_777573
##	0	0	0	0
##	NP_003812	NP_004472	NP_003765	NP_001186711
##	0	0	0	0
##	NP_056268	NP_065937	NP_009115	NP_001263223
##	0	0	0	0
##	NP_987101	NP_987100	NP_689753	NP_001157
##	0	0	0	0
##	NP_001156	NP_005566	NP_787116	NP_006316
##	2	0	0	0
##	NP_001245244	NP_004783	NP_006215	NP_036531
##	0	0	0	0
##	NP_001600	NP_006209	NP_001157207	NP_065941
##	0	0	0	0
##	NP_005237	NP_001996	NP_055362	NP_055363
##	0	0	0	0
##	NP_001078880	NP_039234	NP_005782	NP_038470
##	0	0	0	0
##	NP_061994	NP_620164	NP_997245	NP_002885
##	- 0	_ 0	- 6	- 26
##	NP_995315	NP_001231931	NP_001231934	NP_005588
##	- 0	- 0	_ 0	- 0
##	NP_001257972	NP_001177666	NP_002492	NP_005587
##	- 0	- 0	- 0	- 0
##	NP_005586	NP_060037	NP_002632	NP_065206

	_	_		
##	0	0	32	35
##	NP_004528	NP_005960	NP_006370	NP_148935
##	0	0	0	0
##	NP_000445	NP_775837	NP_076936	NP_000035
##	0	10	0	0
##	NP_004748	NP_001015048	NP_001015049	NP_057305
##	0	0	0	0
##	NP_001129115	NP_003016	NP_003017	NP_005318
##	0	0	0	0
##	NP_001805	NP_057951	NP_001093862	NP_001026902
##	_ 0	_ 0	_ 0	- 0
##	NP_006461	NP 001032407	NP 005489	NP_001123996
##	0	0	0	0
##	NP 955445	NP_001254509	NP 001104792	NP_076977
##	0	0 001204009	N1_001104/32	N1 _010311
##	· ·	•	v	ND 0011E004E
	NP_060821	NP_056029	NP_001181	NP_001158245
##	0	0	0	VD 00400FF00
##	NP_115551	NP_001092108	NP_001135766	NP_001035526
##	0	0	0	0
##	NP_057123	NP_659419	NP_002444	NP_057377
##	0	3	0	0
##	NP_647597	NP_001626	NP_002625	NP_001135858
##	0	0	0	0
##	NP_002601	NP_002602	NP_055918	NP_001248342
##	0	0	0	0
##	NP_001248341	NP_000007	NP_056342	NP_008993
##	0	0	0	0
##	NP_001876	NP_001316	NP_056050	NP_055152
##	0	0	0	0
##	NP 680481	NP 057393	NP 060205	NP 060204
##	- 0	_ 0	- 0	- 0
##	NP 006776	NP 776216	NP 001154806	NP 066407
##	0	0	0	0
##	NP 003511	NP 003512	NP 542160	NP 003519
##	0	N1_000012	N1_042100	M1_000013
##	NP_778225	NP 733759	NP 002630	NP 003898
##	NF_110225	NF_133139	NF_002030	NF_003030
	ND 00401E	ND 004916	ND 004671	ND OFFESA
##	NP_004815	NP_004816	NP_004671	NP_055534
##	0	7	7	0
##	NP_653319	NP_808881	NP_886552	NP_006351
##	0	0	0	0
##	NP_001880	NP_055118	NP_001020114	NP_075559
##	0	0	0	0
##	NP_005048	NP_001180201	NP_115635	NP_000194
##	0	0	0	0
##	NP_060115	NP_005561	NP_005707	NP_573568
##	0	0	0	0
##	NP_060125	NP_004823	NP_891993	NP_002721
##	0	0	0	0
##	NP_002722	NP_001229787	NP_002723	NP_066272
##	0	0	0	_ 0
##	NP_001101	NP_000962	NP_057122	NP_955392
##	_ 0	_ 0	- 0	- 0
##	NP_955393	NP_003278	NP_071905	NP_653296
			_ = = = = = = = = = = = = = = = = = = =	

##	0	0	0	0
##	NP_006861	XP_001717731	NP_001164014	NP_001164015
##	0	0	0	0
##	NP_002884	NP_005601	NP_001185648	NP_001128727
##	- 0	- 0	- 0	- 0
##	NP 001128728	NP 059990	NP_000925	NP_061961
	N1_001120720 0	N1 _003330	NI _000323	
##	· ·	0	0	0
##	NP_055866	NP_071682	NP_612477	NP_775882
##	0	0	31	0
##	NP_056290	NP_000933	NP_003895	NP_056147
##	0	0	0	0
##	NP 060220	NP 006259	NP_689928	NP_055639
##	_ 0	_ 0	_ 0	_ 0
##	NP 001265099	NP_003430	NP_009080	NP 001128650
##	6	M1_000100	WI _0000000	M1_001120000
	•	ND COOCOO	ND 001000401	O O O O O O
##	NP_001265104	NP_689688	NP_001008401	NP_008886
##	6	6	2	6
##	NP_009084	NP_001103159	NP_001018854	NP_008900
##	18	3	3	0
##	NP_079038	NP_001131080	NP_001240728	NP_001123992
##	0	0	14	6
##	NP_001013768	NP_001191385	NP_899061	NP_055604
##	- 8	- 3	- 6	- 6
##	NP_001075949	NP_001078837	NP_065838	NP_001001662
##	6	29	0.00000	3
	-		· ·	_
##	NP_942596	NP_001258777	NP_001243582	NP_940859
##	5	3	30	19
##	NP_056936	NP_116217	NP_612383	NP_689816
##	27	9	0	6
##	NP_003407	NP_001166109	NP_872330	NP_001035275
##	11	15	3	2
##	NP 001004301	NP_001166161	NP 612143	NP_001070146
##	_ 0	_ 0	- 8	21
##	NP 001252526	NP 065906	NP_061983	NP_001265048
##	M1_001202020	18	M _001308	7
	-			ND 001001660
##	NP_008887	NP_116313	NP_001018855	NP_001001668
##	0	0	15	15
##	NP_443084	NP_003416	NP_066971	NP_689839
##	3	16	12	6
##	NP_001166142	NP_001092096	NP_689570	NP_001138461
##	26	18	18	0
##	NP 112234	NP_003406	NP_666019	NP_689697
##	42	_ 6	49	_ 6
##	NP_001008801	NP_848639	NP_008922	XP_003846493
##	0	N1_040003	7	16
			•	
##	NP_055532	NP_001008727	NP_001018857	NP_001129971
##	34	0	4	8
##	NP_997216	XP_003960942	NP_001121695	NP_149350
##	0	2	6	6
##	NP_848618	NP_008890	NP_001007249	NP_008889
##	2	6	6	7
##	NP_689668	NP_009076	NP_001096127	NP_079009
##	29	_ 0	10	3
##	NP_001229609	NP_001138906	NP_862828	NP_001191747
πĦ	141 _001223009	MI _001190900	ME _002020	111 _001131141

##	3	12	14	1
##	NP_003399	NP_065708	NP_001156863	NP_699189
##	7	6	7	9
##	NP_001265102	NP_597730	NP_689475	NP_116078
##	- 6	34	- 8	_ 0
##	NP_055333	NP_872415	NP_940941	NP_694563
##	34	16	0	- 6
##	NP_597717	NP_005764	NP_001129507	NP_065156
##	6	6	0	0
##	NP_060297	NP_001026971	NP_005560	NP_619634
##	0	0	0	0
##	NP_004270	NP_065861	NP_000300	NP_056114
##	0	0	0	0
##	NP_065392	NP_009201	NP_000700	NP_001258395
##	0	0	0	0
##	NP_005474	NP_055238	NP_068746	NP_055871
##	0	0	0	0
##	NP_002336	NP_031377	NP_002681	NP_001659
##	0	0	0	0
##	NP_001184254	NP_055677	NP_001026976	NP_000685
##	0 ND 000690	27	0 ND 000700	ND 060E61
##	NP_000682 7	NP_000686 8	NP_002709	NP_060561
## ##	NP 003001	NP 071334	NP 071356	NP 000840
##	NF_003001 0	NF_071334 0	NF_071330	NF_00040
##	NP_003747	NP 852478	NP 004292	NP 057272
##	0	0	0	0
##	NP_443122	NP 006615	NP 057594	NP 000033
##	_ 0	_ 0	0	_ 0
##	NP_036266	NP_059982	NP_001119595	NP_056937
##	0	0	0	0
##	NP_004248	NP_056340	NP_116162	NP_001155251
##	0	6	0	0
##	NP_006288	NP_002824	NP_001013012	NP_000496
##	0	0	0	0
##	NP_001830	NP_004359	NP_002341	NP_001104567
##	0	0	0	0
##	NP_002101	NP_060919	NP_001121093	NP_057712
##	0	0	0	0
##	NP_567823	NP_055496	NP_004690	NP_036267
##	0 ND 000113	3	0 ND 00100E020	ND 000600
##	NP_000113	NP_001107563	NP_001095838	NP_000629
## ##	0 ND 000084	0 ND 063037	0 NP_001177	ND 001729
##	NP_002084 0	NP_063937 0	NF_001177	NP_001738
##	NP_001185702	NP_001803	NP_060401	NP_060767
##	0	3	0	0 000707
##	NP_056210	NP_001138765	NP_006550	NP_006549
##	0	0 0011007	0	M _000049
##	NP_689901	NP_079359	NP_003829	NP_001106279
##	0	0	0	0
##	NP_006491	NP_000511	NP_001124437	NP_001165415
##	- 0	_ 0	0	0
##	NP_001165416	NP_001107579	NP_055291	NP_001244892

##	0	0	0	0
##	NP_005654	NP_060788	NP_001073919	NP_061973
##	- 0	_ 0	3	- 0
##	NP_060547	NP_078988	NP_001034813	NP_056123
##	0	0	0	0
##	NP_060579	NP_001243232	NP_001185867	NP_001185871
##	0	0	0	0
##	NP_000256	NP_075461	XP_003960990	NP_671704
##	0	43	58	52
##	NP_115758	NP_001138244	NP_054829	NP_001157685
##	0	0	0	0
##	NP_001230892	NP_004542	NP_006144	NP_001004720
##	19	0	0	0
##	NP_001074	NP_246273	NP_076982	NP_001180418
##	4	4	0	0
##	NP_001120686	NP_001096654	NP_001264242	NP_005882
##	0	0	0	0
##	NP_003791	NP_004138	NP_001379	NP_000176
##	0 ND 001171000	0 ND 001165008	ND 055003	ND 001171001
## ##	NP_001171890	NP_001165908	NP_055883	NP_001171891
##	NP_001153651	NP_003194	NP_001188264	NP_000745
##	NF_001133031	NF_003194 0	NF_001186204	NF_000745
##	NP_443157	NP_056538	NP_004068	NP_057114
##	U _440107	M1_030330	0 0 4000	N1_007114 0
##	NP 004777	NP_036556	NP_443188	NP_000713
##	0	0	3	0
##	NP 002606	NP 002755	NP 002756	NP 008941
##	- 0	_ 0	_ 0	_ 0
##	NP_071381	NP_003720	NP_079160	NP_001616
##	0	0	0	0
##	NP_001186128	NP_055859	NP_619525	NP_037497
##	0	0	0	0
##	NP_000960	NP_000427	NP_060273	NP_001598
##	0	0	0	0
##	NP_003968	NP_056444	NP_149061	NP_071350
##	0	0	0	0
##	NP_054735	NP_001531	NP_060631	NP_004652
##	0	0 ND 775470	0	0 0
##	NP_005860	NP_775179	NP_060515	NP_060138
##	0 ND 11042E	0 ND 001290	ND 076412	0 002000
## ##	NP_110435 0	NP_001389 0	NP_076412	NP_003892
##	NP_055331	NP_006691	NP_277047	NP_001123544
##	0	0	0	0
##	NP_060427	NP 683723	NP_055779	NP_689725
##	0	0	0	54
##	NP_001159403	NP_056528	NP_056041	NP_001230332
##	0	0	0	0
##	NP_001216	NP_001129584	XP_003960209	NP_001096045
##	0	5	0	_ 0
##	NP_443102	NP_002216	NP_057068	NP_001186190
##	0	0	0	0
##	NP_003725	NP_033720	NP_003375	NP_001074012

##	0	0	0	0
##	NP_057307	NP_001121700	NP_001139490	NP_001139488
##	_ 0	0	- 0	- 0
##	NP_005834	NP_003464	NP_061940	NP_057231
##	0	0	0	0
##	NP_002175	NP_001171619	NP_001028219	NP_001171621
##	0	0	0	0
##	NP_055686	NP_005198	NP_060139	NP_116764
##	0	0	9	0
##	NP_054745	NP_078894	NP_038466	NP_444295
##	0	0	0	0
##	NP_064516	NP_038472	NP_059509	NP_005556
##	0	0	5	0
##	NP_056170	NP_443094	NP_079012	NP_659731
##	0 ND 000747	0 ND 000740	0	0 ND 445700
##	NP_002747	NP_002749	NP_001003962	NP_115706
## ##	NP 001171867	NP 065960	NP_060894	NP 444504
##	NF_001171807	NF_005900 0	NF_000094	NF_444504
##	NP 001922	NP 066923	NP 055566	NP_612392
##	0	0	0	N1_012032
##	NP_005602	NP_624311	NP 060230	NP_078813
##	4	0	0	0
##	NP 071760	NP 006633	NP_006704	NP 057195
##	_ 0	- 0	- 0	_ 0
##	NP_064575	NP_114415	NP_001139632	NP_570602
##	0	0	0	0
##	NP_002004	NP_055965	NP_037415	NP_060334
##	0	0	0	0
##	NP_001099026	NP_055752	NP_060873	NP_438171
##	0	0	0	0
##	NP_002442	NP_006107	NP_036421	NP_998734
##	0	0	0	0
##	NP_006139	NP_006384	NP_001258537	NP_001120675
##	0 ND 070030	U 004040204	ND 000000	ND 000001
## ##	NP_078930	NP_001012321	NP_006828	NP_008921
##	NP_006637	NP_003790	NP_536350	NP_001095856
##	1	0	0	M1_001030000
##	NP_892023	NP_005263	NP_000163	NP_056994
##	0	0	0	22
##	NP_536351	NP_001070957	NP_006563	NP_002061
##	_ 0	- 0	- 0	_ 0
##	NP_006487	NP_002060	NP_031379	NP_620073
##	0	0	0	0
##	NP_001035997	NP_891552	NP_057058	NP_060586
##	0	0	0	0
##	NP_001243461	NP_060640	NP_067038	NP_113626
##	0	0	0	0
##	NP_065949	NP_071908	NP_006539	NP_006538
##	0	0	0	3
##	NP_006537	NP_036519	NP_057317	NP_847884
##	ND 115706	0 ND 003360	16	ND 004070407
##	NP_115796	NP_003360	NP_001279	NP_001076437

##	0	0	0	0
##	NP_057327	NP_057452	NP_002991	NP_976325
##	0	_ 0	0	0
##	NP_149991	NP_001165131	NP_037390	NP_660183
##	3	0	0	0
##	NP_006222	NP_002906	NP_005350	NP_006308
##	11	0	0	0
##	NP_006532	NP_062539	NP_001731	NP_004920
##	0	0	0	0
##	NP_689588	NP_065952	NP_060498	NP_056404
##	25	0	0	0
##	NP_077748	NP_006266	NP_005617	NP_001034554
##	0	0	0	0
##	NP_001372	NP_536739	NP_006627	NP_001935
##	0	0	0	32
##	NP_573571	NP_001136022	NP_009106	NP_006450
##	0 ND 0010E7440	ND 056160	ND 0011002E	ND 000167
## ##	NP_001257449 5	NP_056160 2	NP_001188356	NP_000167
##	NP_000892	NP_002907	NP_002660	NP_001129695
##	NF_000692 8	NF_002907	NF_002000	NF_001129093
##	NP 055971	NP_001129696	NP 683759	NP 006126
##	0	0 001123030	N1 _003733	N1_000120 0
##	NP 006127	NP 001071168	NP_078888	NP_006409
##	0	0	0	0
##	NP 006065	NP_065970	NP 001018146	NP_001018148
##	_ 0	_ 0	0	0
##	NP_937818	XP_003960919	NP_004628	NP_116189
##	0	0	0	0
##	NP_006138	NP_075392	NP_001070254	NP_005056
##	0	0	9	0
##	NP_065701	NP_775818	NP_060825	NP_031396
##	0	0	10	0
##	NP_001123584	NP_073745	NP_001003722	NP_056473
##	0	0	0	0
##	NP_006187	NP_065389	NP_127501	NP_004286
##	0	0	0	0
##	NP_996667	NP_821068	NP_065751	NP_005158
##	0	0	0 ND 004075	22
##	NP_000150	NP_839946	NP_001875	NP_075378
##	0 ND 0010134EE	ND 001640	ND 001650	12 ND 001651
## ##	NP_001013455 0	NP_001649	NP_001650	NP_001651
##	NP_001653	NP 057660	NP_006868	NP_004044
##	0	0	M1_000000	0
##	NP_848018	NP 777568	NP_003894	NP 068806
##	0	0	0	0
##	NP_037453	NP_055886	NP_006076	NP_000535
##	0	6	0	0
##	NP_055099	NP_001014841	NP_001121187	NP_055795
##	0	0	0	35
##	NP_001020424	NP_001020423	NP_057531	NP_001265299
##	0	0	0	3
##	NP_004337	NP_963293	NP_963294	NP_060177

		•	•	
##	0	0	0	0
##	NP_258261	NP_653232	NP_689473	NP_079055
##	28	45	1	0
##	NP_003064	NP_001007469	NP_001887	NP_001091990
##	0	0	0	0
##	NP_001350	NP_055623	NP_066953	XP_003960097
##	0	0	0	0
##	NP 001116540	NP 001137504	XP 003960579	XP_001129774
##	- 0	_ 0	_ 0	- 0
##	XP 001718975	XP_003960166	NP_036277	NP 064715
##	0	3	0	0
##	NP_689481	NP_078954	NP 055203	NP 003869
##		NF_070334	NF_000200	NF_003009
	0	ND AFFORE	U 004455040	VD 770000
##	NP_000254	NP_055985	NP_001155313	NP_778236
##	0	0	0	0
##	NP_008937	NP_001244259	NP_001244258	NP_001076582
##	0	0	0	0
##	NP_663318	NP_689862	NP_065393	NP_722550
##	0	0	0	0
##	NP 997403	NP_001924	NP_068598	NP 066012
##	- 0	_ 0	- 0	- 0
##	NP 004782	NP_001258685	NP_443108	NP_653254
##	0	0	0	M _000201
##	NP_001013009	NP_613258	NP_613075	NP 061119
	-	NF_013230	-	NF_001119
##	0	0	0	U 004005477
##	NP_620707	NP_620634	NP_620448	NP_001265477
##	0	0	0	0
##	NP_008920	NP_005919	NP_001108086	NP_000058
##	5	0	0	0
##	NP_689522	NP_002655	NP_005850	NP_001128119
##	0	0	0	0
##	NP_079517	NP_037511	NP_001132929	NP_000159
##	0	0	0	10
##	NP 005261	NP 005260	NP 001123484	NP 004776
##	40	29	_ 0	_ 0
##	NP 001239002	NP 071354	NP 001272	NP 009141
##	0	M _011001	MI_0012/2	MI_000111
	NP_004473	ND 001000011	ND 00104E202	ND OF 610F
##	_	NP_001229811	NP_001245383	NP_056195
##	0	0	0	0
##	NP_079152	NP_005650	NP_003644	NP_071896
##	0	0	0	0
##	NP_001189431	NP_003565	NP_919415	NP_004729
##	0	0	0	0
##	NP_005948	NP_055892	NP_001542	NP_001227
##	0	0	0	0
##	NP_001748	NP_056089	NP_219482	NP_443070
##	_ 0	_ 0	_ 0	_ 0
##	NP_002679	NP_004565	NP_001009939	NP_015556
##	0	0	0	3
##	NP_060599	NP_001474	NP_001189398	NP_006741
	_	_	MI _001103930	_
##	0 ND 006460	ND 0E10E2	ND 001127416	ND 001410
##	NP_006460	NP_851853	NP_001137416	NP_001410
##	0	0	0 ND 000774	0
##	NP_004423	NP_001138247	NP_068771	NP_115657

##	0	0	0	51
##	NP_003123	NP_001012339	NP_919259	NP_002426
##	0	0	0	0
##	NP_060440	NP_001338	NP_006673	NP_004392
##	4	0	0	0
##	NP_998731	NP_077285	NP_001032622	NP_001185832
##	_ 0	_ 0	- 0	- 0
##	NP 808227	NP 001243615	NP_004584	NP_037425
##	0	0.0	0	0
##	NP 001229825	NP_001229826	NP_005594	NP_077816
##	3	3	1	56
##	NP_079113	NP 958839	NP_003635	NP_001191379
	_	NF_930039	NF_003033	
##	30	0	0	18
##	NP_004739	NP_004695	NP_060720	NP_001186064
##	20	0	0	0
##	NP_079023	NP_005521	NP_005830	NP_904324
##	0	0	0	0
##	NP_060789	NP_001193856	NP_001360	NP_003681
##	0	0	3	0
##	NP_001132989	NP_004259	NP_076436	NP_001136148
##	0	0	0	0
##	NP 821074	NP 001230405	NP 821075	NP 055478
##	0	0	0	0
##	NP_000552	NP_000839	NP_000841	NP 000842
##	0	M1_000009	0 IF0000_ IM	N1 _000042
##	· ·	NP 054894	ŭ	ND 004447
	NP_666533	NP_054694	NP_001190176	NP_004447
##	0	VD 000540	U 000400	VD 0000E0
##	NP_001982	NP_036513	NP_009193	NP_000053
##	3	0	0	0
##	NP_110414	NP_076991	NP_055472	NP_001405
##	0	0	0	0
##	NP_056437	NP_003365	NP_005653	NP_652759
##	0	0	0	0
##	NP_652760	NP_009096	NP_005264	NP_002065
##	3	0	0	0
##	NP 067642	NP 004274	NP 002860	NP 940892
##	- 0	- 0	_ 0	_ 0
##	NP_057614	NP_006852	NP_057238	NP_002861
##	0	0	0	0
##	NP 004569	NP_055303	NP_002857	NP_002858
##	0 ND 00107110F	0 ND 040500	0 ND 110506	0 ND 001050067
##	NP_001071105	NP_942599	NP_112586	NP_001258967
##	0	0	0	0
##	NP_612462	NP_741995	NP_004785	NP_941959
##	0	0	17	0
##	NP_059986	NP_783865	NP_057406	NP_057215
##	0	0	0	0
##	NP_004152	NP_005361	NP_001020471	NP_112243
##	0	0	0	0
##	NP_057661	NP_001182144	NP_079174	NP_037460
##	- 0	- 0	- 0	_ 0
##	NP_008838	NP_722546	NP_067045	NP_001032208
##	0	0	0	0
##	NP_114152	NP_001137472	XP_003846541	NP_004611
π π	WI _II4102	MI _001101412	vi	MI _004011

	_	_	_	_
##	0	8	8	0
##	NP_060282	NP_060668	NP_003609	NP_001257354
##	0	0	3	3
##	NP_004570	NP_006036	NP_940933	NP_001186125
##	0	1	10	0
##	NP_079105	NP_001012768	NP_001171590	NP_001171587
##	0	0	0	0
##	NP_005750	NP_002782	NP_620581	NP_001306
##	0	0	0	0
##	NP 149082	NP 060895	NP 065733	NP 689853
##	- 0	_ 0	_ 0	_ 0
##	NP 001139669	NP_005504	NP_000192	NP 036344
##	0	0	0	0
##	NP_065194	NP_003304	NP_002731	NP_001166110
##	M1_000134	F00000_ IN	M _002701	0 - 001100110
##	ND 001000767	ND OOGEEO	ND 001000047	NP_699204
	NP_001020767	NP_006552	NP_001020247	NP_099204
##	0	VD 004477044	U	U 004040404
##	NP_031401	NP_001177344	NP_003081	NP_001012426
##	0	0	0	0
##	NP_612466	NP_001231737	NP_001231739	NP_001231744
##	0	0	0	0
##	NP_008816	NP_078997	NP_055979	NP_115521
##	10	24	0	0
##	NP_001093165	NP_031399	NP_077024	NP_005996
##	0	0	0	3
##	NP_003633	NP_004663	NP_036370	NP_001001484
##	0	2	0	0
##	NP_060295	NP_689485	NP_003174	NP_079172
##	0	0	0	0
##	NP_001032209	NP_060708	NP_149974	NP_079348
##	23	_ 0	_ 0	_ 0
##	NP 852136	NP 071413	NP 060648	NP 060275
##	_ 0	- 0	- 0	- 0
##	NP 061072	NP 065789	NP 001258766	NP 005443
##	0	9	0	0
##	NP 001505	NP_055899	NP 001248363	NP 001129514
##	0	0	0	0
##	NP_055824	NP_115901	NP_569119	NP_002827
##	N1_000024	18	N1 _303113	0 002027
##	NP 004394	NP_055282	· ·	
	-	-	NP_002757	NP_002758
##	0 ND 001030071	ND 005044	0 ND 0040E7004	ND 0000CE
##	NP_001230871	NP_005044	NP_001257291	NP_002865
##	0	0 ND 000670	0	0 ND 007007
##	NP_005724	NP_060672	NP_003110	NP_037367
##	16	0	3	3
##	NP_000622	NP_060268	NP_001166599	NP_689971
##	0	0	0	0
##	NP_006600	NP_002392	NP_054879	NP_006442
##	0	0	0	0
##	NP_877590	NP_899152	NP_006181	NP_004413
##	0	0	0	0
##	NP_004412	NP_004414	NP_004083	NP_060762
##	6	0	0	0
##	NP_005710	NP_060382	NP_055421	NP_114431

##	0	6	8	0
##	NP_003724	NP_006312	NP_001015055	NP_149035
##	0	0	0	0
##	NP 001846	NP 150093	NP_872407	NP_001180363
##	_ 0	_ 0	- 0	- 0
##	NP 001029097	NP 003243	NP_071505	NP_002943
	0	N1 _003243	N1 _07 1303	N1 _002343
##	•	· ·	U	0
##	NP_758515	NP_006864	NP_006087	NP_002783
##	0	0	0	0
##	NP_003757	NP_060026	NP_001019386	NP_001009552
##	0	0	0	0
##	NP 002711	NP 002706	NP 114108	NP 078817
##	_ 0	_ 0	_ 0	_ 0
##	NP 057562	NP 006088	NP_001138417	NP_006462
##	0	0	M1_001100117	M1_000102
	ŭ	· ·	ND 071414	ND 070400
##	XP_373042	NP_003468	NP_071414	NP_079498
##	0	0	0	0
##	NP_000843	NP_065997	NP_061820	NP_001005741
##	0	0	0	0
##	NP_001243692	NP_110424	NP_006800	NP_059118
##	0	0	0	0
##	NP 057572	NP_958923	NP 001137259	NP 694453
##	_ 0	_ 0	_ 0	_ 0
##	NP_000503	NP_005300	NP 597700	NP 057094
##	0	0	M1_037700	M1_007034
	•	· ·	ŭ	VD 004005000
##	NP_005785	NP_878912	NP_073742	NP_001035830
##	0	0	0	0
##	NP_054742	NP_115517	NP_006454	NP_115867
##	0	0	0	0
##	NP_006500	NP_620693	NP_001884	NP_001885
##	0	0	0	0
##	NP 055563	NP 001253989	NP 005643	NP 001202
##	_ 0	_ 0	_ 0	15
##	NP 057373	NP 112187	NP 848632	NP_004716
##	0	NI_IIZIO7	M _040002	M1_004/10
	ŭ	ND 001100171		VD 001070726
##	NP_647473	NP_001166171	NP_001139478	NP_001070736
##	0	6	0	0
##	NP_064576	NP_000121	NP_003760	NP_006523
##	0	7	0	0
##	NP_001124335	NP_006200	NP_001804	NP_004897
##	0	0	0	0
##	NP_001229510	NP_940873	NP_004993	NP_056386
##	_ 0	_ 0	0	9
##	NP 065705	NP 079467	NP_003817	NP_064583
##	11	M1_075407	M1_000017	000£000
		ND 000031	ND OCOCOO	ND 610257
##	NP_000062	NP_002031	NP_060829	NP_612357
##	0	0	0	0
##	NP_001254523	NP_055712	NP_660149	NP_000384
##	0	0	3	0
##	NP_115580	NP_115717	NP_001734	NP_006283
##	0	0	0	0
##	NP_002077	NP_987102	NP_000910	NP_004453
##	0	0	8	0
##	NP_058651	NP_060353	NP_001159898	NP_004648
π#	MI _000001	Mt _000303	MI _001199990	ML _004040

##	0	0	0	0
##	NP_077015	NP_001002009	NP_001002010	ND 002207
##	NF_077013 12	NF_001002009	NF_001002010	NP_002297
##		· ·	NP_005255	ND 665726
	NP_002490 0	NP_001975	NP_005255	NP_665736
##	•	· ·	ND 067652	0 ND 076070
##	NP_056364	NP_076994	NP_067653	NP_076970
##	0 ND 070050	0	ND 600657	U 0.000.40
##	NP_078958	NP_001161694	NP_689657	NP_066949
##	0	ND 026260	ND 00220E	ND 000FF0
##	NP_002600	NP_036369	NP_003305	NP_002558
##	0	0	VD 705000	0
##	NP_062555	NP_057086	NP_705690	NP_689659
##	0	0	0	0
##	NP_036587	NP_031375	NP_001687	NP_542384
##	0	0	0	0
##	NP_005307	NP_000381	NP_001812	NP_002955
##	0	0	0	0
##	NP_112178	NP_060518	NP_065987	NP_001163931
##	0	0	0	0
##	NP_002586	NP_002587	NP_001976	NP_001560
##	0	0	0	0
##	NP_005090	NP_079131	NP_075047	NP_004123
##	0	0	0	0
##	NP_002347	NP_075385	NP_620124	NP_001028740
##	0	0	0	0
##	NP_002877	NP_001258115	NP_066361	NP_006408
##	0	0	0	0
##	NP_001120690	NP_057071	NP_653176	NP_061004
##	0	0	1	0
##	NP_075066	NP_001788	NP_006589	NP_057366
##	0	0	0	0
##	NP_056648	NP_001103	NP_598368	NP_009133
##	0	0	0	0
##	NP_057038	NP_071739	NP_064587	NP_005836
##	0	0	0	32
##	NP_009119	NP_775829	XP_003960280	NP_073741
##	0	0	6	0
##	NP_056005	NP_006636	NP_945327	NP_006766
##	0	0	0	0
##	NP_001190991	NP_036203	NP_995319	NP_006120
##	0	0	0	0
##	NP_036597	NP_036596	NP_060232	NP_073620
##	_ 11	23	1	0
##	NP_001008661	NP_004050	NP_036269	NP_055577
##	0	3	_ 0	- 0
##	NP_000775	NP 055670	NP_002909	NP_602304
##	- 0	_ 0	- 0	_ 11
##	NP_000626	NP_060012	NP_000704	NP_612409
##	35	3	0	3
##	NP_055199	NP_008830	NP_064530	NP_001121181
##	0	0	0	0
##	NP_005333	NP_000117	NP_001121188	NP_006796
##	0	0	0	0
##	NP_001139443	NP_057390	NP_001137299	NP_001137298
	001100110	001000		

##	0	0	0	0
##	NP_775102	NP_055785	NP_116759	NP_057132
##	0	0	0	0
##	NP_001166	NP_036253	NP_001116857	NP_001116856
##	0	0	13	13
##	NP 001230228	NP 000138	NP_001263181	NP_002956
##	- 0	_ 0	_ 0	_ 0
##	NP 478059	NP 073589	NP_005472	NP_002125
##	0	0	0	0
##	NP_001035998	NP_060197	NP_055744	NP_631908
##	NF_001033998	NF_000137	NF_033744	NF_031300
	ŭ	VD 004000500	ND 000505	0 10 10 10 10 10 10 10 10 10 10 10 10 10
##	NP_001193988	NP_001008530	NP_000525	NP_001121616
##	0	0	6	6
##	NP_569120	NP_004705	NP_060671	NP_055332
##	0	3	0	0
##	NP_002195	NP_006853	NP_001077433	NP_064595
##	2	0	0	0
##	NP_997006	NP_078787	NP_054861	NP_036530
##	0	0	0	0
##	NP 003548	NP_001129108	NP_003549	NP 857593
##	_ 0	_ 0	14	_ 0
##	NP 001027539	NP_055330	NP 003289	NP 005780
##	0	0.00000	N1 _000205	M1_000700
	· ·	NP_000116	ND COOSES	ND 057527
##	NP_001025055	-	NP_690852	NP_057537
##	24	0	0	0
##	NP_775840	NP_002583	NP_001035787	NP_003307
##	3	0	0	6
##	NP_004102	NP_115818	NP_001026897	NP_861448
##	0	3	0	0
##	NP_006206	NP_659486	NP_001729	NP_078859
##	0	0	0	0
##	NP_057444	XP_003960482	NP_001280	NP_005538
##	0	0	0	9
##	NP 056651	NP 036534	NP 065114	NP 060052
##	_ 0	_ 0	_ 0	0
##	NP_775740	NP 000368	NP 006406	NP 001789
##	0	0	0	0
##	NP_001777	NP_001249	NP_036557	NP_733779
	_	_	_	•
##	0 ND 663736	0 ND 004704	ND 050035	ND 002202
##	NP_663736	NP_004791	NP_056235	NP_003303
##	0	0	0	0
##	NP_060245	NP_071324	NP_060228	NP_006218
##	0	0	0	0
##	NP_001961	NP_001137232	NP_065123	NP_055095
##	0	0	0	0
##	NP_036234	NP_001136040	NP_001263215	NP_003013
##	0	0	9	0
##	NP_002932	NP_001122401	NP_079107	NP_998776
##	10	36	3	- 0
##	NP_004926	NP_003150	NP_036434	NP_055736
##	0	18	3	37
##	NP_000447	NP_940876	NP_001129511	NP_001185808
##	0	0.0040	NF_001129311 0	001100000
	-	-	-	ND OCSEE
##	NP_004880	NP_899064	NP_863654	NP_863655

##	0	0	0	0
##	NP_056273	NP_003807	NP_060559	NP_036270
##	0	3	0	0
##	NP_001164675	NP_002745	NP_002960	XP_003846692
##	0	0	3	3
##	NP 057030	NP 064632	NP_000893	NP_775949
##	0	0	2	0
##	NP_001123498	NP_694953	NP_060317	NP_002018
	NF_001125498	_	NF_000317	
##	•	19	U	0
##	NP_001899	NP_114110	NP_008929	NP_001052
##	0	0	0	0
##	NP_001138612	XP_003960877	NP_001107590	NP_001231
##	24	46	0	0
##	NP_001026890	NP_079194	NP_006234	NP_078889
##	0	0	0	0
##	NP 077286	NP_001264125	NP_940915	NP 116119
##	_ 0	_ 0	0	_ 0
##	NP 060544	NP 065921	NP_689763	NP_001092759
##	0 000	N1 _003321	NI _003703	N1_001032733
	•	•		
##	NP_060497	NP_055938	NP_001257904	NP_001257906
##	0	0	0	0
##	NP_006808	NP_001137532	NP_116164	NP_001123617
##	0	0	0	0
##	NP_001181973	NP_006097	NP_057386	NP_066983
##	0	0	0	0
##	NP_006075	NP_002779	NP_996892	NP_000008
##	0	0	0	0
##	NP_991403	NP_001247418	NP_004610	NP_003291
##	0	0	0	13
##	NP 056182	NP 006320	NP 064709	NP 064512
##	_ 0	_ 0	_ 0	_ 0
##	NP 001119809	NP 076413	NP 689876	NP 060517
##	0	M1_070410	M1_005070	M1_000017
	•	ND 112640	VD 002107	ND 000E17
##	NP_620147	NP_113642	NP_003187	NP_002517
##	0	0	0	004000
##	NP_003394	NP_996806	NP_001116698	NP_003482
##	0	0	5	0
##	NP_116082	NP_113673	NP_005917	NP_060761
##	0	0	0	0
##	NP_113668	NP_002724	NP_057287	NP_059127
##	0	0	0	43
##	NP_031373	NP_001193581	NP_001193580	NP_001193582
##	2	0	0	0
##	NP_056461	NP_002875	NP_008896	NP_001186743
##	0	0	0	14
##	NP_689567	NP_001123468	NP_037493	NP_065858
##	44	39	32	43
##	NP_037530	NP_085137	NP_001027544	NP_001076804
##	32	NF_003137	NF_001027344 24	141 _001010004
				ND OFOACC
##	NP_037494	NP_659570	NP_689650	NP_852466
##	34	30 ND 005005	25 ND 001420	ND 005000
##	NP_958357	NP_005885	NP_001439	NP_005699
##	5	0	0	0
##	NP_055908	NP_006068	NP_000020	NP_006281

	_			_
##	0	0	0	3
##	NP_067643	NP_789847	NP_001432	NP_071440
##	0	0	0	0
##	NP_067067	NP_976060	NP_001265488	NP_079017
##	0	0	6	57
##	NP_060200	NP_000894	NP_000365	NP_001127703
##	_ 0	_ 0	_ 0	- 0
##	NP 073568	NP 112179	NP_005872	NP_002538
##	0	0	0	3
##	NP 005717	NP 001139013	NP_079139	NP_525129
##	0	M1_001103010	M1_073103	M1_020125
##	•	ND 0610E7	ND 002747	ND 775000
	NP_003192	NP_061857	NP_203747	NP_775922
##	0	5	18	0
##	NP_976225	NP_001238978	NP_055660	NP_057260
##	0	0	0	0
##	NP_057631	NP_000160	NP_006362	NP_006446
##	0	0	0	0
##	NP_036399	NP_742000	NP_001138827	NP_001138828
##	0	0	0	0
##	NP_001078926	XP_003960503	NP_671717	NP_001186621
##	0	0	_ 0	5
##	NP 963868	NP_005524	NP 001184224	NP 055717
##	20	0	0	0
##	NP 892118	NP 057407	NP_060244	NP 787050
##	N1_032110 0	2	M _000244 3	1 _707000
##	NP 055439	NP_055313	_	ND 004006
	-	NP_055313	NP_004555	NP_004986
##	0 ND 056400	VD 004644	U 000000	U 00440447
##	NP_056480	NP_004614	NP_002022	NP_001184147
##	0	0	0	0
##	NP_001184149	NP_001098101	NP_006194	NP_001032417
##	0	0	0	0
##	NP_001104777	NP_001184151	NP_000914	NP_055986
##	28	9	38	0
##	NP_001186036	NP_065756	NP_004484	NP_009035
##	0	0	0	0
##	NP 003529	NP 000217	NP 060367	NP 002901
##	- 0	_ 0	- 0	- 0
##	NP_203125	NP_203124	NP_001253986	NP_003183
##	0	0	0	0
##	NP 004228	NP 073591	NP_112233	NP_036345
##	0 004220	M1_073331	NI_II2255	
		-	-	ND 001003
##	NP_071933	NP_002200	NP_001116080	NP_001883
##	0	0	0	0
##	NP_001258670	NP_001258671	NP_660204	NP_068811
##	0	0	0	0
##	NP_002948	NP_005225	NP_003288	NP_008848
##	0	0	43	0
##	NP_055422	NP_710154	NP_963906	NP_444279
##	0	0	0	5
##	NP_000181	NP_005392	NP_008970	NP_060566
##	_ 0	- 4	- 41	- 0
##	NP_057604	NP_064569	NP_060119	NP_001668
##	0	0	16	0
##	NP_996895	NP_006794	NP_945344	NP_000966
ππ	MI _330033	111 _000134	MI _340044	141 _000300

	^	•	4.0	
##	0	0	10	0
##	NP_001186731	NP_003646	NP_079087	NP_001136037
##	0	0	0	0
##	NP_116249	NP_057041	NP_001153416	NP_006296
##	0	0	0	0
##	NP_006392	NP_036535	NP_079506	NP_060533
##	0	0	0	3
##	NP 006004	NP_001243506	NP_055054	NP_009126
##	_ 0	- 0	_ 0	_ 0
##	NP 001119575	NP 055764	NP 387506	NP_001186779
##	0	0	0	0
##	NP_000476	NP 001003	NP 005821	NP_997080
##	0 14-000 - 14	NF_001003	NF_003021	
	ŭ	ND 00000	ND ACOFOR	3 ND 0000E0
##	NP_001248394	NP_060606	NP_068593	NP_000850
##	3	0	0	10
##	NP_114402	NP_071922	NP_006807	NP_115805
##	0	0	0	0
##	NP_055546	NP_060337	NP_690870	NP_001746
##	18	0	0	0
##	NP_074036	NP_113649	NP_002856	NP_116235
##	0	17	0	0
##	NP 001004	NP 068749	NP 001028693	NP 000467
##	_ 0	_ 0	_ 0	_ 0
##	NP_777283	NP_061981	NP_000115	NP_001263987
##	0	0	0	M1_001200001
##	NP 057070	NP_079431	NP_002785	NP_001186708
		NF_0/9431	NF_002765	NF_001100700
##	0 ND 004406700	VD 005770	U 000000000	VD 440000
##	NP_001186709	NP_005779	NP_001138639	NP_110396
##	0	0	0	0
##	NP_001153408	NP_002506	NP_002507	NP_006294
##	0	0	0	0
##	NP_149075	NP_060366	NP_002572	NP_006232
##	6	0	24	0
##	NP_076415	NP_001139016	NP_000961	NP_003969
##	0	0	0	0
##	NP 060092	NP 001153803	NP 001004439	NP 116262
##	_ 0	15	- 6	- 4
##	NP_061834	NP_149975	NP_003680	NP_036199
##	40	0	0	0
##	NP_056272	NP_037505	NP_001171953	NP_004636
##	N1 _030212 0	19	N1_001171333	N1 _004030
##	NP_115558	NP_065956	NP_001186032	NP_001186016
##	0	0	0	57
##	NP_037524	NP_689794	NP_004314	NP_002574
##	0	3	0	0
##	NP_003363	NP_057314	NP_065080	NP_851851
##	0	6	0	3
##	NP_001017526	NP_001185650	NP_001028733	NP_060726
##	3	22	0	0
##	NP_775738	NP_001073948	NP_001231293	NP_001931
##	_ 0	8	8	3
##	NP_115689	NP_705902	NP_078870	NP_683878
##	0	0	0	0
##	NP_059120	NP_001171564	NP_005495	NP_001171565
##	ML _009120	NF_0011/1004	NF_005495	Mt _00111,1909

##	0	0	0	0
##	NP_036392	NP_001869	NP_001892	NP_001003652
##	0	0	0	0
##	NP_005891	NP_005894	NP_005893	NP_001120689
##	0	0	0	0
##	NP_077315	NP_004795	NP_005138	NP_001128582
##	0	0	0	0
##	NP_060329	NP_060531	NP_002780	NP_001096138
##	0	0	0	0
##	NP_065773	NP_001017405	NP_005165	NP_001001973
##	9	0	0	0
##	NP_001096645	NP_000916	NP_001191066	NP_057039
##	3	0	0	0
##	NP_002063	NP_002058	NP_004288	NP_006451
##	_ 0	_ 0	_ 0	- 0
##	NP_006388	NP_055706	NP_055926	NP_001070993
##	_ 0	_ 0	- 8	- 0
##	NP 001070994	NP 001026871	NP_059142	NP_055512
##	_ 0	_ 0	_ 0	- 0
##	NP 054790	NP 659407	NP_057078	NP_060666
##	0	0	0	_ 0
##	NP_001036083	NP 064519	NP_001801	NP_006346
##	3	4	0	0
##	NP_077308	NP 001229569	NP_683720	NP_219499
##	0	0	0	10
##	NP_001002	NP_112594	NP_060979	NP_060585
##	0	0	0	0
##	NP 001001669	NP 002784	NP_000124	NP 149098
##	0	0	0	0
##	NP 001036182	NP 054699	NP 003678	NP 849193
##	0	0	0	0
##	NP 057021	NP 060417	NP 071447	NP 068594
##	0	M1_000117	M _0/ 111/	0
##	NP 660341	NP 055323	NP 612467	NP 001165774
##	0 IF 0000 IN	N1 _000020	N1_012407	0
##	NP_116233	NP 056142	NP 000598	NP 000599
##	N1_110200	3	M1 _000030	M1 _000033
##	NP_001008219	NP_066188	NP_057392	NP_004587
##	20	20	N1_007032	0 NI _004007
##	NP_937863	NP_057117	NP_071439	NP_060421
##	0	0	0	0
##	NP 060221	NP_001139130	NP_055900	NP_073617
##	0	N1_001103100	1	0
##	NP_036555	NP_075567	NP_060018	NP 078969
##	0	0	M1_000010	M1_070303
##	NP_110425	NP_079223	NP_932157	NP_055829
##	NF_110425	NF_079223	NF_932137	_
##	NP 003734	NP_055832	NP_777565	NP 061905
		_	NF_111303	NF_001903
## ##	0 NP 008975	0 NP 001014402	NP_001020331	77 \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \
	NP_008975	NP_001014402	_	NP_038475
## ##	0 ND 570711	NP 002570	0 NP 056167	V ممهدی۱
##	NP_570711 0	NP_002570	NP_056167	NP_996531
		· ·	-	0 ND 058133
##	NP_004459	NP_001092264	NP_037458	NP_058132

##	0	0	0	0
##	NP_002525	NP_001027581	NP_057313	NP_061188
##	0	0	0	0
##	NP_004308	NP_060842	NP_079094	NP_003553
##	0	0	0	0
##	NP_001158889	NP_612364	NP_671517	NP_056362
##	0	0	0	0
##	NP_001185987	NP 872422	NP_005820	NP_060791
##	- 0	_ 0	_ 0	_ 0
##	NP 938204	NP_001164221	NP 006298	NP 006585
##	0	0	0	6
##	NP 002819	NP 110390	NP 055467	NP_149078
##	N1 _002019	N1_110550	WI _000407	N1 _143070
	•	ND 000470	ND OOFOOA	ND 00004E
##	NP_076915	NP_060476	NP_005824	NP_000345
##	0	0	0	0
##	NP_064572	NP_057126	NP_003822	NP_004613
##	0	0	0	0
##	NP_060692	NP_037459	NP_037389	NP_056258
##	0	0	0	2
##	NP_006540	NP_000627	NP_002196	NP_002067
##	0	0	0	0
##	NP 001130002	NP 006207	NP_008938	NP 002451
##	_ 0	0	_ 0	2
##	NP_076424	NP 057256	NP_009077	NP_001036004
##	0	0	0	0
##	•	v	ŭ	ND 0/E10E
	NP_002374	NP_066575	NP_061943	NP_945185
##	•	-	VD 004065505	VD 00000E
##	NP_665857	NP_060513	NP_001265525	NP_208325
##	0	14	32	5
##	NP_653310	NP_003666	NP_003850	NP_001624
##	0	0	0	0
##	NP_073624	NP_945339	NP_112197	NP_932174
##	0	0	0	0
##	NP_055594	NP_904358	NP_004080	NP_006013
##	0	0	0	0
##	NP 001015881	NP 001230726	NP 006010	NP_001006934
##	- 0	_ 0	_ 0	- 0
##	NP_001012997	NP_002803	NP_003550	NP_005019
##	0	0	0	0
##	NP_065071	NP_113637	NP_060729	NP_009016
##	0	0	M1_000129	WI_003010
		-	-	
##	NP_001494	NP_004883	NP_001137160	NP_859066
##	0	0	3	0
##	NP_001002030	NP_001159884	NP_005623	NP_004696
##	0	0	12	0
##	NP_001265432	NP_057439	NP_001138304	NP_001106962
##	0	0	0	0
##	NP_570899	NP_057060	NP_060158	NP_056131
##	3	0	0	3
##	NP_001230575	NP_056345	NP_060703	NP_076956
##	0	_ 0	_ 0	_ 0
##	NP_001186744	NP_001186745	NP_001180237	NP_001123237
##	0	0	0	0
##	NP_079148	NP_001138347	NP_001138348	NP_037498
ππ	141 _019140	111 _001100041	MI _001100040	141 _021 430

##	0	0	3	0
##	NP_073615	NP_116038	NP_060716	NP_065715
##	3	0	0	0
##	NP_006419	NP_003911	NP_008909	NP_849192
##	0	3	0	20
##	NP_004778	NP_003052	NP_443097	NP_001978
##	10	55	0	0
##	NP 938022	NP 705718	NP_065130	NP_003647
##	0	0	0	0
##	NP_001191732	NP 001136151	NP 066284	NP_932349
##	0	0	0	0.0000
##	NP_002073	NP 005299	NP 892027	NP 258260
##	NF_002073	NF_003299	NF_092027	NF_230200
	v	U		Z 0.40.40
##	NP_073557	NP_002296	NP_060058	NP_061942
##	0	0	0	0
##	NP_789782	NP_005616	NP_001007070	NP_001092084
##	0	0	0	10
##	NP_057037	NP_689545	NP_060563	NP_076952
##	0	0	0	0
##	NP_057370	NP_067636	NP_064547	NP_002845
##	0	0	3	2
##	NP_690866	NP_001834	NP_000098	NP_004290
##	_ 0	- 30	_ 0	_ 0
##	NP 942600	NP 060286	NP 006527	NP 004172
##	0	0	32	0
##	NP 079089	NP_036300	NP 001837	NP_001186512
##	M _073003	M1_000000	M _001007	0 001100012
##	NP 005861	NP 068750	NP 060106	NP 663615
##	NF_003001 0	NF_000730	NF_000100	NF_003013
	ŭ	· ·	_	ND 11///0
##	NP_060140	NP_112485	NP_055138	NP_114440
##	0	0	6	0
##	NP_005099	NP_001035374	NP_001165276	NP_003120
##	0	3	32	8
##	NP_899646	NP_057584	NP_058634	NP_828847
##	0	0	0	15
##	NP_001020092	NP_077306	NP_000285	NP_001245388
##	0	0	0	5
##	NP_002416	NP_002413	NP_009214	NP_056399
##	2	0	0	0
##	NP_872282	NP_057665	NP_758454	NP_001400
##	0	0	0	- 6
##	NP_079108	NP_612451	NP_004119	NP_001257817
##	_ 0	_ 0	_ 0	0
##	NP_001257882	NP_077321	NP_001123952	NP_758961
##	0	0	0	0
##	NP_006865	NP 973728	NP 001412	NP_056326
##	N1 _000003	N1 _373720	NI _001412 17	N1 _030320
	-	-		ND 6001E0
##	NP_006315	NP_001352	NP_060654	NP_620150
##	0	0	0 ND 050450	0
##	NP_031385	NP_071941	NP_659453	NP_001180500
##	0	0	0	0
##	NP_001312	NP_064508	NP_000095	NP_002475
##	0	0	7	0
##	NP_003600	NP_001184252	NP_002834	NP_001189488

##	0	0	0	0
##	NP_002019	NP_001190992	NP_075050	NP_542199
##	0	0	0	0
##	NP_000382	NP_006103	NP_982281	NP_001018119
##	0	0	0	0
##	XP_003846661	NP_000066	NP_057301	NP_004872
##	0	0	0	0
##	NP_001638	NP_005964	NP_060850	NP_085131
##	- 0	_ 0	- 0	- 0
##	NP_663612	NP 004077	NP_612373	NP_115737
##	0	50	0	0
##	NP_001092315	NP_001035173	NP 689594	NP_001186072
##	0	0	M _000001	46
##	NP_001258762	NP_079177	NP_006811	NP_055161
##	NF_001256762	NF_0/91//	NF_000011	NF_033101
	VD 001120007	U 00000	ND 004640	ND 026520
##	NP_001139287	NP_003004	NP_004640	NP_036539
##	0	0	0	0
##	NP_775813	NP_001001974	NP_004830	NP_004263
##	0	0	3	0
##	NP_005814	NP_037536	NP_000178	NP_001303
##	53	53	0	0
##	NP_001257766	NP_000172	NP_057111	NP_777571
##	0	0	0	0
##	NP_001433	NP_002668	NP_000108	NP_004888
##	0	0	0	0
##	NP_079527	NP_443115	NP_057637	NP_002778
##	15	7	0	0
##	NP 055488	NP 006758	NP_001446	NP_002006
##	- 0	_ 0	- 0	- 5
##	NP_005929	XP_002342143	XP_003960170	NP_002183
##	20	8	8	0
##	NP_001116293	NP_003002	NP_001234930	NP_115547
##	0	M1_000002	0	M3
##	NP 001070667	NP_060616	MD 002700	ND 00110/107
##	NF_001070007	NF_000010	NP_002788	NP_001124197
	ND 000171	ND 001129650		ND 004772
##	NP_009171	NP_001128659	NP_000299	NP_004773
##	VD 076000	U 00000F	000000	VD 007405
##	NP_076983	NP_003695	NP_003266	NP_037485
##	0	8	0	45
##	NP_060376	NP_005527	NP_005993	NP_001135442
##	0	0	0	0
##	NP_055682	NP_149989	NP_009125	NP_665861
##	9	0	0	0
##	NP_060826	NP_060423	NP_079150	NP_005302
##	0	0	0	3
##	NP_008879	NP_612441	NP_004881	NP_001000
##	0	0	0	0
##	NP_005567	NP_113608	NP_060864	NP_000953
##	- 0	- 0	- 0	- 0
##	NP_003721	NP_064711	NP_653324	NP_064535
##	0	0	0	0
##	NP_001028675	NP_071446	NP_001129465	NP_694943
##	0	0	0	0 0 10 10
##	NP_000395	NP_078782	NP_598001	NP_001159643
π π	141 _000333	141 _070702	141 _030001	MI _001103040

##	0	0	0	0
##	NP_001159639	NP_002072	NP_003783	NP_694880
##	0	0	0	0
##	NP_002564	NP_001519	NP_005031	NP_003976
##	0	0	0	4
##	NP_001099009	NP_076418	NP_037462	NP_932076
##	3	3	0	0
##	NP_079217	NP_064555	NP_001035517	NP_005317
##	0	0	0	0
##	NP_060176	NP_115575	NP_689821	NP_060851
##	0	14	0	0
##	NP_061153	NP_005013	NP_000430	NP_004481
##	0	0	54	10
##	NP_000037	NP_001033073	NP_005340	NP_055091
##	_ 0	8	_ 0	16
##	NP_003965	NP_490647	NP 006727	NP_660157
##	_ 0	_ 0	_ 0	16
##	NP 078974	NP 113669	NP_660186	NP_036443
##	_ 0	- 0	_ 0	_ 0
##	NP_003674	NP 006708	NP 001010862	NP_001006682
##	_ 0	0	_ 0	16
##	NP 054892	NP_060874	NP_006860	NP_037520
##	0	0	0	0
##	NP 001136028	NP 000779	NP 037478	NP_003786
##	3	0	0	0
##	NP 000876	NP_003732	NP 872349	NP 067027
##	0	22	0	0
##	NP 004115	NP 004868	NP 056381	NP 006862
##	0	0	11	0
##	NP 006417	NP_001263199	NP_055690	NP 005462
##	8	3	21	0
##	NP 612208	NP 001257810	NP 001612	NP 000588
##	0	0	0	0
##	NP 060190	NP 001013863	NP 004829	NP 064621
##	N1 _000130	M1_001010000	N1 _004025	NI_004021
##	NP 000024	NP_005155	NP 000182	NP 001046
##	0	36	N1 _000102	0-1010
##	XP_003960686	NP_001045	NP_803880	NP_057267
##	M _000500000	0-001040	M1_000000	0
##	NP 005330	NP 067077	NP_006347	NP_658988
##	0	0	0	9
##	NP_695003	NP_002012	NP_060813	NP_001015877
##	0	14	M1_000018	0 001010011
##	NP_689759	NP_004790	NP_001243359	NP_004049
##	3	N1_004730	3	M1_004045
##	NP_814444	NP_004596	NP_006464	NP_001152862
##	0 0	NF_004530	NF_000404 3	NF_001132002
##	NP_001229768	NP_001229769	NP_000844	NP_002801
##	NF_001229708	NI _001223109	NF_00044 0	NF_002001
##	-	MD VVEUEO	-	MD VV311E
	NP_003366	NP_005958	NP_005957	NP_003115
## ##	0 ND 073753	0 NP 001655	7 ND 786886	רטעטט מו <i>ו</i>
##	NP_073753 3	NP_001655	NP_786886 0	NP_004031
##		-	-	·
##	NP_055503	NP_066959	NP_996734	NP_060481

##	3	0	6	0
##	NP_001074930	NP_114409	NP_057724	NP_006399
##	0	0	0	0
##	NP_789783	NP_115727	NP_003381	NP_001035982
##	0	0	8	17
##	NP_001245297	NP_001245299	NP_060947	NP_079510
##	17	23	0	0
##	NP_000971	NP_612450	NP_056139	NP_001613
##	0	3	0	0
##	NP_056338	NP_039268	NP_073606	NP_112601
##	0	0	0	0
##	NP_001660	NP_001011719	NP_000038	NP_008925
##	24	48	59	0
##	NP_008979	NP_660215	NP_008978	NP_060543
##	_ 0	3	16	25
##	NP_001723	NP 006757	NP_036462	NP_055410
##	- 15	- 3	_ 0	_ 0
##	NP 003918	NP_000969	NP 001155133	NP_001670
##	_ 0	0	_ 0	_ 0
##	NP 056534	NP 060733	NP 085135	NP 060601
##	0	0	24	2
##	NP 005432	NP_002805	NP 005207	NP_001034669
##	6	0	0	20
##	NP_005656	NP 037379	NP_659540	NP_444272
##	M _000000	N1 _03/3/3	N1 _000040	N1 _ 111 2/2
##	NP 079222	NP_065828	NP_006061	NP 057087
##	NF_0/9222 0	NF_003028	NF_000001	NF_037087
##	NP 001028691	NP 001098016	NP 001258933	NP_056445
##	NF_001028091 0	NF_001090010	NF_001236933	NF_030443 26
##	NP 001007238	NP 071801	NP 003392	NP_060528
	NF_001007238	NF_0/1001	NF_003392	NF_000526
##		ND 026516	v	VD 004270
##	NP_006573	NP_036516	NP_149073	NP_004379
##	0	0	WD 057604	U 004000
##	NP_783195	NP_543151	NP_057691	NP_004323
##	5	5	0	0
##	NP_005116	NP_005063	NP_005126	NP_001128243
##	0	3	9	14
##	NP_000294	NP_002667	NP_055418	NP_001041675
##	0	0	2	0
##	NP_005994	NP_066972	NP_872620	NP_003433
##	0	0	0	0
##	NP_003418	NP_056509	NP_001041631	NP_741996
##	13	30	41	57
##	NP_002959	NP_001004309	NP_000358	NP_001172023
##	56	35	0	0
##	NP_001230	NP_056125	NP_001011	NP_659484
##	0	18	0	0
##	NP_996668	NP_060277	NP_443080	NP_689953
##	5	0	0	0
##	NP_001035937	NP_689891	NP_001001794	NP_001243893
##	0	0	27	0
##	NP_115744	NP_071935	NP_057166	NP_937784
##	0	0	3	24
##	NP_057097	NP_002410	NP_653172	NP_056270

шш	0	2	0	1
##	0 ND 060340	ND 001107000	ND 07004E	ND 001101010
##	NP_060340	NP_001107000	NP_072045	NP_001191819
##	0	0	0	0
##	NP_001001	NP_954976	NP_061913	NP_002848
##	0	0	0	0
##	NP_003738	NP_079511	NP_002664	NP_001156729
##	8	17	10	58
##	NP_060909	NP_001138430	NP_001026866	NP_001363
##	0	0	0	0
##	NP 001264044	NP 775899	NP_001070913	NP_056103
##	- 0	_ 0	5	_ 1
##	NP_955352	NP_073210	NP 079463	NP_001189367
##	58	0	0	4
##	NP 005975	NP_001243463	NP_037499	NP_002954
	- .	NF_001243403	NF_03/499	- .
##	0	· ·	WD 065700	0
##	NP_789793	NP_110438	NP_065728	NP_005912
##	0	0	0	15
##	NP_940852	NP_001027536	XP_003846351	NP_877439
##	0	0	2	3
##	NP_001254711	NP_001864	NP_071412	NP_001028255
##	13	5	0	0
##	NP_071902	NP_001139267	NP_001265618	NP_078940
##	2	0	0	0
##	NP_620148	NP_003559	NP_116186	NP 659434
##	- 0	- 0	- 0	- 9
##	NP_658985	NP 001010867	NP_036249	NP 057671
##	0	0	0	0
##	NP 001120700	NP 001138349	NP 060383	NP 077001
##	NF_001120700 0	NF_001130349	NF_000363	NF_077001
	ŭ	U 000000100	v	· ·
##	NP_002412	XP_003960188	NP_001451	NP_789794
##	2	11	22	/
##	NP_000938	NP_057270	NP_077084	NP_002627
##	0	0	0	0
##	NP_001165081	NP_009009	NP_001073583	NP_001165082
##	2	2	2	27
##	NP_001921	NP_001193903	NP_078925	NP_002790
##	0	0	1	0
##	NP_060380	NP_003745	NP_001027451	NP_003211
##	_ 0	_ 0	_ 0	_ 0
##	NP_001035890	NP_003213	NP 003212	NP_848643
##	_ 0	0	_ 0	_ 0
##	NP_758438	NP_116196	NP_065691	NP_001014305
##	3	0	0	6
##		NP 003788		NP_997242
	NP_115677	_	NP_001231634	-
##	0 ND 057000	0 ND 057705	0	51
##	NP_057292	NP_057705	NP_060810	NP_055552
##	0	3	0	0
##	NP_114412	NP_003012	NP_001253975	NP_006582
##	0	0	1	0
##	NP_116098	NP_115500	NP_037457	NP_612385
##	10	0	0	0
##	NP_057303	NP_115698	NP_005497	NP_001191184
##	0	_ 0	_ 0	0
##	NP_036380	NP_060156	NP_000968	NP_004840

##	0	0	0	0
##	NP_066973	NP_542391	NP_061863	NP_085130
##	5	0	0	0
##	NP_001170854	NP_001171608	NP_055727	NP_742128
##	0	0	0	0
##	NP 003627	NP_001186791	NP_003462	NP_683710
##	_ 0	_ 0	- 5	_ 0
##	NP_001138836	NP_689618	NP_116114	NP_001153208
##	0	5	0	0
##	NP_001026880	NP 997229	NP 078922	NP_115690
##	22	N1 _557 225	N1 _070322	N1_110030
##		ND 001000040	NP 004126	ND 7772E0
	NP_037533	NP_001099040	NP_004126	NP_777358
##	0	0 ND 500640	VD 0700FF	VD 007040
##	NP_061832	NP_588613	NP_079355	NP_997243
##	0	8	0	22
##	NP_203748	NP_001010888	NP_078887	NP_065778
##	14	17	17	6
##	NP_919299	NP_061896	NP_001171753	NP_005819
##	0	1	1	0
##	NP_006501	NP_004984	NP_060293	NP_036220
##	0	0	0	0
##	NP_055031	NP_060866	NP_859076	NP_009104
##	0	0	0	0
##	NP_002086	NP_849189	NP 055224	NP 056094
##	_ 0	- 0	- 0	- 9
##	NP 598399	NP 821077	NP 001191435	NP_690618
##	0	4	9	11
##	NP 001087194	NP 976049	NP 057710	NP_115622
##	0	0	5	5
##	NP 000584	NP 006055	NP 006561	NP_001161067
##	P00000_1M	M _000000	M1_000001	M1_001101007
##	NP 976316	NP 005633	NP 071909	NP 031368
	NF_970310 2	NF_003033	NF_071909	NF_031300
##	_	ND 070470	ŭ	VD 000047
##	NP_003121	NP_079470	NP_036515	NP_060647
##	0	0	0	4 ND 005044
##	NP_001006	NP_005310	NP_005312	NP_005311
##	0	0	0	0
##	NP_005313	NP_005316	NP_005314	NP_000311
##	0	0	0	0
##	NP_443183	NP_064695	NP_001074007	NP_003336
##	0	0	0	0
##	NP_001160163	NP_004319	NP_000137	NP_061139
##	13	0	0	0
##	NP_000378	NP_115885	NP_060930	NP_001243111
##	0	0	0	11
##	NP_056023	NP_000590	NP_001013416	NP_077078
##	15	0	0	0
##	NP_003285	NP_001654	NP_001647	NP_001186558
##	0	0	0	_ 0
##	NP_066299	NP_524147	NP_000249	NP_001002841
##	0	0	0	21
##	NP_524144	NP_077310	NP 444251	NP_001129508
##	NI _324144 0	N1_077310 0	NI _444231 0	N1_001125506
##	NP_004862	NP_001007026	NP_055819	NP_009139
##	Nr_004002	Mt_001001020	ML_000019	ML_009198

##	0	0	0	0
##	NP_612444	NP_001562	NP_001184051	NP_002078
##	0	0	0	0
##	NP_003840	NP_056526	NP_073570	NP_001030312
##	0	10	0	0
##	NP_001240693	NP_057113	NP 036337	NP_071918
##	_ 0	_ 0	0	11
##	NP_006327	NP_001032726	NP_001198	NP_004682
##	M1_000021	0	M _001130	M _00 1002
##	NP_689778	NP_004681	NP_001257448	NP_055387
##	N1 _003770	N1_00 1 001	N1_001237440	43
	ND 005001	4 A C C C C C C C C C C C C C C C C C C	ND 001040212	
##	NP_005021	NP_002939	NP_001240313	NP_002422
##	8	0	0	0
##	NP_006160	NP_008976	NP_001008	NP_005702
##	0	0	0	28
##	NP_055814	NP_060386	NP_001244119	NP_001092885
##	0	0	0	0
##	NP_001177779	NP_077007	NP_055880	NP_005554
##	0	0	31	0
##	NP_008960	NP_055268	NP_114143	NP_071357
##	0	0	0	0
##	NP_036313	NP_001185727	NP_001185728	NP_000323
##	_ 0	0	_ 0	_ 0
##	NP_079007	NP 006528	NP_001265612	NP_057095
##	- 0	_ 0	- 4	- 0
##	NP_004291	NP_009030	NP_113663	NP_112486
##	0	0	6	0
##	NP 037364	NP_001012777	NP_055513	XP_003960281
##	T-00100-	N1_001012///	NI _000010	AT _000500201
##	VD 002060202	NP_612404	NP_055112	NP_588614
	XP_003960282	NF_012404	NF_055112	_
##	ND 001002014	ND OFFESO	ND 004706	11 ND 000010
##	NP_001093214	NP_055538	NP_001786	NP_808818
##	0	3	0	0
##	NP_003109	NP_005189	NP_666499	NP_036212
##	0	0	0	0
##	NP_001171606	NP_653337	NP_660306	NP_001240
##	0	0	14	3
##	NP_004634	NP_001186793	NP_006616	XP_003960063
##	0	0	0	0
##	NP_542781	NP_036379	NP_057491	NP_038478
##	0	0	0	13
##	NP_001137498	NP_002786	NP_003711	NP_005393
##	0	0	0	0
##	NP_002872	NP_079202	NP_001243439	NP_000104
##	_ 0	2	- 0	_ 0
##	NP_001078816	NP_001238947	NP_835368	NP_058626
##	0	5	3	0
##	NP_056226	NP_001035934	NP_002712	NP_056347
##	0	N1_001033334 0	N1 _002/12 0	NI _000547
##		NP_057218		
	NP_001018100	_	NP_067639	NP_689787
##	0 ND 060060	0 ND 055755	ND 000301	ND 07990E
##	NP_060962	NP_055755	NP_000301	NP_078895
##	0	0	0 ND 000405	0 ND 700000
##	NP_115910	NP_001177396	NP_000495	NP_783322

шш	0	0	^	^
##	0 ND 071761	ND 001100177	VD 770042	ND 004400
##	NP_071761	NP_001182177	NP_778243	NP_004499
##	0	0	0	0
##	NP_150286	NP_006699	NP_006547	NP_689494
##	3	0	0	0
##	NP_056238	NP_001017926	NP_001025041	NP_079472
##	0	0	46	0
##	NP_001103447	NP_055545	NP_057104	NP_006205
##	11	0	0	0
##	NP_071424	NP 443169	NP_008933	NP_006749
##	- 0	_ 0	_ 0	_ 0
##	NP_001020374	NP_659424	NP_065894	NP_001177386
##	0	0	23	0
##	NP 113631	NP 001025006	NP_000207	NP_004674
##	N1_113031 0	0 001023000	NI _000207	12
	•	· ·		
##	NP_000362	NP_001907	NP_004369	NP_002023
##	0	0	0	0
##	NP_001327	NP_001919	NP_037408	NP_000811
##	0	0	0	0
##	NP_001793	NP_004112	NP_002937	NP_057147
##	4	0	0	0
##	NP_077295	NP_006345	NP_065683	NP_002781
##	0	0	22	0
##	NP 001129224	NP 001129225	NP 002873	NP 848604
##	0	0	_ 0	_ 0
##	NP 056982	NP 776185	NP_002966	NP_001863
##	0	0	0	0
##	NP 003244	NP 000572	NP 003155	NP 116201
##	12	0	0	0
##	NP 060309	NP 006545	NP 054893	NP 954994
	NF_000309	NF_000343	NF_054695	NF_954994
##	· ·	ND 001100004	ŭ	ND 000167
##	NP_079422	NP_001129294	NP_001995	NP_009167
##	0	0	0	0
##	NP_115873	NP_001159743	NP_001166951	NP_056135
##	0	5	32	35
##	NP_694999	NP_954664	NP_004890	NP_053733
##	0	0	0	0
##	NP_001010883	NP_001080	NP_061985	NP_079066
##	0	32	22	14
##	NP_001679	NP_062559	NP_079368	NP_006305
##	0	0	1	0
##	NP_001092906	NP_001614	NP_004838	NP_056117
##	- 0	_ 0	- 0	- 0
##	NP_060551	NP_001157814	NP_116181	NP_000469
##	0	2	18	0
##	NP_005014	NP_659436	NP_444284	NP_689775
##	0 000014	12	M _444204 6	12
##	NP_003077	NP_006824	NP_001034702	NP_112496
##	6	0	0	0
##	NP_001036189	NP_002847	NP_005848	NP_803882
##	0	0	0	0
##	NP_660331	NP_665877	NP_005708	NP_001073331
##	19	0	0	0
##	NP_001192048	NP_001258965	NP_065998	NP_001013728

##	0	0	0	14
##	NP_004055	NP_003780	NP_115553	NP_064536
##	0	0	0	0
##	NP_000895	NP_001265233	NP_149023	NP_057001
##	0	0	0	0
##	NP 066564	NP_003875	NP_005693	NP_004556
##	_ 0	14	_ 0	_ 0
##	NP 060848	NP_114160	NP_066960	NP_849194
	0	0 0	NF_000300	
##	· ·	· ·	ND 001275	ND 005031
##	NP_848659	NP_077297	NP_001375	NP_005931
##	0	0	0	0
##	NP_055952	NP_005689	NP_542397	NP_443160
##	0	0	0	19
##	NP_001003891	NP_065955	NP_775745	NP_057457
##	0	22	0	0
##	NP_976035	NP 006626	NP_775751	XP 003960920
##	_ 0	_ 0	58	8
##	NP_659413	NP_003408	NP 001124317	NP_998763
##	NF_059413 58	NF_003408	31	NF_990703
		_		· ·
##	NP_710155	NP_115919	NP_067040	NP_060122
##	6	9	33	9
##	NP_659448	NP_114068	NP_055603	NP_077003
##	36	2	0	0
##	NP_932332	NP_004177	NP_057055	NP_003301
##	0	17	0	0
##	NP 001093634	NP_620602	NP 689628	NP_001164432
##	- 0	_ 0	_ 0	- 17
##	NP 055921	NP 116264	NP 036289	NP_002890
##	23	14	32	0
##	NP 001124464	NP 004802	NP 036307	NP 612361
	NF_001124404	NF_004002	NF_030307	NF_012301
##	ŭ	U 004000	VD 050506	VD 057004
##	NP_001071087	NP_001893	NP_859526	NP_057224
##	0	0	0	0
##	NP_116259	NP_065107	NP_001108488	NP_620156
##	0	0	6	0
##	NP_001171649	NP_004121	NP_006744	NP_001265871
##	0	0	0	9
##	NP_002677	NP_000422	NP_003338	NP_004273
##	_ 0	_ 0	_ 0	_ 0
##	NP 002843	NP_000428	NP_000091	NP_055991
##	0	3	0	0
##	NP 064524	NP_064533	NP_005763	NP_001010853
	-		_	NF_001010033
##	0	3	0 ND 004056	T
##	NP_003595	NP_004963	NP_004956	NP_001364
##	54	6	1	3
##	NP_579917	NP_036226	NP_113664	NP_056970
##	2	0	0	49
##	NP_001091977	NP_001027464	NP_009191	NP_997320
##	0	0	0	0
##	NP_001243404	NP_115587	NP_898871	NP_005245
##	35	50	0	0
##	NP_002032	NP_963839	NP_001129671	NP_001031
	NF_002032 2	NF_903039		
##		-	ND 002014	0 ND 055760
##	NP_612452	NP_115814	NP_003014	NP_055760

##	0	6	3	0
##	ND 001005971	-	NP_068597	NP_115528
##	NP_001095871 5	NP_570721 0	NF_000597	NF_115526
			-	
##	NP_002640	NP_004627	NP_006014	NP_002643
##	ND 110204	5 ND 001699	ND 600120	ND 600600
##	NP_110394	NP_001688	NP_620130	NP_689600
##	11	0	2	6 ND 056056
##	NP_079351	NP_006244	XP_003960220	NP_056356
##	0 ND 7003E6	0 ND 110574	0 ND 056176	33
##	NP_700356	NP_112574	NP_056176	NP_055740
##	54	0	4 ND 004507	9
##	NP_057384	NP_001713	NP_064507	NP_054891
##	21	0	0	0
##	NP_001129333	NP_000388	NP_060590	NP_001193594
##	0	0	0	0
##	NP_001094346	NP_777593	NP_003759	NP_112217
##	0	0	0	14
##	NP_002124	NP_065118	NP_653182	NP_065207
##	0	0	0	3
##	NP_443149	NP_653168	NP_055634	NP_660095
##	22	0	0	9
##	NP_003005	NP_000437	NP_000931	NP_004070
##	0	0	0	0
##	NP_000537	NP_612393	NP_060389	NP_689786
##	9	0	0	16
##	NP_001002836	NP_055037	NP_000253	NP_001139400
##	0	0	0	3
##	NP_005081	NP_001108104	NP_057090	NP_004539
##	0	0	0	0
##	NP_065128	NP_006109	NP_001018238	NP_060036
##	0	0	0	0
##	NP_067026	NP_002922	NP_009143	NP_064528
##	0	0	0	0
##	NP_054786	NP_001005339	NP_057651	NP_510868
##	0	0	0	0
##	NP_001006618	NP_001258	NP_004454	NP_055431
##	0	0	2	18
##	NP_002576	NP_002577	NP_006186	NP_079521
##	0	0	0	47
##	NP_997227	NP_006558	NP_055100	NP_057726
##	0	0	0	0
##	NP_060633	NP_149080	NP_000582	NP_004662
##	0	0	0	0
##	NP_057250	XP_003960206	XP_003960203	NP_077005
##	0	3	3	0
##	NP_001193769	NP_001030331	NP_005810	NP_003756
##	6	0	- 0	_ 0
##	NP_001258538	NP_001243549	NP_001243548	NP_004759
##	0	3	3	0
##	NP_055969	NP_001171940	NP_113614	NP_060104
##	0	5	0	0
##	NP_000615	NP_006212	NP_003249	NP_112240
##	0	0	0	_ 0
##	NP_008826	NP_068744	NP_000967	NP_060241

	_	_	_	_
##	0	0	0	0
##	NP_001902	NP_997244	NP_006051	NP_001207695
##	0	29	0	0
##	NP_057344	NP_071910	NP_038474	NP_005655
##	4	7	0	11
##	NP_115766	NP_071895	NP_057558	NP_004568
##	0	3	0	0
##	NP_005863	NP_002688	NP_001193954	NP_055167
##	0	0	0	0
##	NP_001035784	NP_620149	NP_892117	NP_001435
##	- 17	_ 0	- 0	- 0
##	NP_001161726	NP_004996	NP_056140	NP_003670
##	0	0	20	26
##	NP_005460	NP 001071131	NP_001071132	NP_056299
##	0	1	NI _001071132	N1 _030233
##	· ·	_	MD VEEVEO	ND 070014
	NP_055059	NP_001265576	NP_055058 49	NP_078914
##	19	30		VD 007070
##	NP_116199	NP_002792	NP_005092	NP_937872
##	0	0	0	10
##	NP_060609	NP_057258	NP_689478	NP_003409
##	6	1	0	0
##	NP_001120665	NP_001120664	NP_055984	NP_001128529
##	0	0	0	0
##	NP_002353	NP_001011543	NP_004979	NP_001159872
##	11	53	43	43
##	NP_722517	NP_079130	NP_612461	NP_001010854
##	0	3	0	5
##	NP_116558	NP_055178	NP_005774	NP_004903
##	0	10	0	9
##	NP_067075	NP_001243340	NP_001078	NP_055020
##	- 0	0	_ 0	- 0
##	NP 055070	NP 001116224	NP 851939	NP_001030024
##	- 0	- 3	- 0	_ 10
##	NP 006330	NP 114140	NP 003378	NP_775781
##	0	0	0	3
##	NP_055567	NP 001032354	NP 057226	NP_060933
##	0	0	0	0
##	NP_005925	NP_003078	NP_003076	NP_001243384
##	0	0	N1 _003070	N1_001243304 0
##	NP 075387	NP_114091	NP_002155	NP_001258975
	- _	_	- .	_
##	7	ND 000065	ND 000000	33
##	NP_008850	NP_002965	NP_060283	NP_001116801
##	51 ND 000050	50	0	14
##	NP_006658	NP_006311	NP_001002295	NP_116027
##	0	0	0	0
##	NP_002043	NP_536721	NP_851996	NP_000606
##	2	2	44	44
##	NP_116204	NP_060368	NP_073583	NP_060285
##	0	0	0	0
##	NP_055646	NP_036432	NP_378663	NP_001243785
##	7	3	7	6
##	NP_004595	NP_055502	NP_036311	NP_055482
##	0	5	0	0
##	NP_001121691	NP_060521	NP_075394	NP_000777

##	0	11	0	3
##	NP_061845	NP_002544	NP_054888	NP_066552
##	0	3	0	0
##	NP_036204	NP_036595	NP_006471	NP_000185
##	0	3	_ 2	0
##	NP 065928	NP 803173	NP_115700	NP_001070914
##	8	0	_ 0	_ 0
##	NP 060453	NP_079345	NP_116161	NP_056465
##	0	0.000	24	0
##	NP 001009894	NP_071926	XP_003119578	NP_001029366
##	0	18	M _000110070	0
##	NP_060023	NP 115729	NP_000937	NP_057480
##	N1 _000023	NI_II0/25	NI _000337	N1_057400
##		ND COOOSO	ND 110507	ND OOESO1
	NP_055731	NP_689939	NP_112597	NP_005391
##	U 000046	VD 540000	WD 000050	U 000400
##	NP_006246	NP_542398	NP_006053	NP_060480
##	0	0	0	3
##	NP_060545	NP_001191763	NP_001191761	NP_787103
##	0	0	0	0
##	NP_940887	NP_114429	NP_001870	NP_624302
##	2	3	0	0
##	NP_001027019	NP_071441	NP_057490	NP_065108
##	0	0	0	0
##	NP_055610	NP_001167567	NP_005709	NP_001185722
##	0	52	0	0
##	XP_003960151	NP_004672	NP_851808	NP_001229344
##	0	0	1	12
##	NP_443723	NP_001138501	NP_060379	NP_056091
##	3	0	0	0
##	NP_001138545	NP_958443	NP_055250	NP_006784
##	5	31	0	0
##	NP_001157567	NP_001172110	NP_056020	NP_006629
##	0	0	16	0
##	NP_000046	NP_065142	NP_059993	NP_954590
##	_ 0	_ 0	_ 0	_ 0
##	NP_006684	NP_001123357	NP_005855	NP_115719
##	- 0	- 57	- 3	_ 0
##	NP 848591	NP_006434	NP 954653	NP_001265280
##	_ 0	0	_ 0	29
##	NP_004801	NP_060745	NP_056284	NP_060846
##	_ 0	0	_ 0	15
##	NP_002008	NP_001161153	NP_001129626	NP_001257939
##	2	2	13	2
##	NP_001129627	NP_059991	NP_001020419	NP_001122080
##	13	36	0	22
##	NP_000296	NP_653265	NP_689873	NP_001137384
##	NF_000290 0	NF_055205	NF_009073	
## ##	NP_060089	NP_004090		2 NP_005985
	_	-	NP_005987	
##	0 ND 001199665	0 ND 001196506	0 ND E7011E	0 ND 000070
##	NP_001182665	NP_001186506	NP_570115	NP_000970
##	0 ND 000040	0 ND 00103EE40	ND 036606	0 ND 112600
##	NP_060849	NP_001035540	NP_036606	NP_113620
##	0 ND 446470	2 ND 005054	0	0
##	NP_116172	NP_065851	NP_001026882	NP_006001

##	0	3	3	0
##	NP_001129206	NP_689706	NP_055288	NP_076959
##	N1_001123200 0	4	N1 _000200 0	N1_070303
##	NP_001034937	NP_061989	NP_003253	NP_056325
		NF_001989 0	NF_003233	NF_030323
##	0 ND 665007	-	-	•
##	NP_665807	NP_004547	NP_006369	NP_055585
##	0	0	8	3
##	NP_001116244	NP_057269	NP_001002033	NP_689505
##	2	0	0	1
##	NP_000987	NP_653210	NP_057118	NP_068779
##	0	0	0	0
##	NP_056256	NP_872293	NP_003917	NP_085148
##	0	26	0	0
##	NP_055533	NP_612366	NP_612142	NP_085151
##	17	0	0	0
##	NP_073571	NP_808807	NP_665858	NP_036359
##	3	1	0	0
##	NP_003816	NP_115584	NP_477520	NP_116032
##	_ 0	_ 0	- 28	_ 0
##	NP 061158	NP_689518	NP 003934	NP_037438
##	- 0	_ 0	- 0	- 0
##	NP 000190	NP 056144	NP 060445	NP_054808
##	6	0	0	0
##	NP 006338	NP 116265	NP 443089	NP_036231
##	0	8	0	0
##	NP 056061	NP 443148	NP 061895	NP_001138503
##	0	12	0	0
##	NP 659477	NP 060242	NP 443204	NP 057358
##	0	0	0	2
##	NP 002308	NP 919424	NP 077284	NP_056950
##	0	N1_515424	0	21
##	NP 036363	NP 073567	NP 068572	NP_444252
##	NF_030303 21	NF_073507	NF_000372	NF_444252 0
		ND 00100000	·	· ·
##	NP_002619	NP_001002294	NP_079428	NP_775099
##	0 ND 001121	55 ND 775106	0 ND 000500	ND 056204
##	NP_001131	NP_775106	NP_060560	NP_056324
##	34	3	0	0
##	NP_000975	NP_006135	NP_001238968	NP_004153
##	0	0	0	0
##	NP_001238965	NP_001231630	NP_008818	NP_004917
##	0	0	0	0
##	NP_001656	NP_061485	NP_002863	NP_005043
##	0	0	0	0
##	NP_001782	NP_426359	NP_036381	NP_065714
##	0	0	0	0
##	NP_001001563	NP_004647	NP_620412	NP_733839
##	0	14	0	0
##	NP_002433	NP_076943	NP_060017	NP_057034
##	0	0	0	0
##	NP_005677	NP_008871	NP_001139291	NP_036610
##	4	26	26	0
##	NP_057154	NP_057152	NP_001002913	NP_073581
##	0	0	0	3
##	NP_057616	NP_056137	NP_005508	NP_001188292

	•	-	^	
##	0	5	0	377.004500
##	NP_006344	NP_004233	NP_001188291	NP_001536
##	6	3	3	0
##	NP_787117	NP_001138467	NP_001185624	NP_006285
##	0	3	0	0
##	NP 001186904	NP 055547	NP_002804	NP_001161212
##	- 0	- 0	_ 0	- 0
##	NP 077304	NP 057110	NP 001186752	NP_006147
##	111 _01100±	N1_007110	M1_001100702	M1_000147
	ND 0011067F0	· ·	ND 776106	ND 054670
##	NP_001186750	NP_116193	NP_776186	NP_954673
##	0	0	0	0
##	NP_003341	NP_001027459	NP_068823	NP_001244324
##	0	0	0	0
##	NP_001244326	NP_954580	NP_001123556	NP_003456
##	0	27	5	48
##	NP_009129	NP 060372	NP_689449	NP_078976
##	0	0	0	0
##	NP 060338	NP_001020279	NP_002052	NP 057731
	NF_000558	NF_001020279	NF_002032	NF_037731
##	_	•	0	VD 000750
##	NP_005172	NP_000144	NP_001138582	NP_060758
##	0	0	48	0
##	NP_006080	NP_079106	NP_060069	NP_002446
##	9	0	9	0
##	NP_001108224	NP_004381	NP_612377	NP_075566
##	0	0	3	0
##	NP_689488	NP_689730	NP 005555	NP 055730
##	- 7	- 7	_ 0	_ 0
##	NP 612486	NP 000845	NP 004209	NP 004654
##	13	0	0	0
##	NP 003977	NP 065109	NP 001879	NP 004214
	-	WI _000103	N1 _0010/3	N1_004214
##	20	000000 011	_	VD 004045
##	NP_003152	NP_003286	NP_997305	NP_004315
##	0	0	0	0
##	NP_001138792	NP_001186284	NP_057479	NP_000478
##	25	0	0	0
##	NP_056028	NP_659410	NP_001122400	NP_997263
##	1	6	0	9
##	NP_055557	NP_057107	NP_006782	NP_996670
##	_ 0	- 0	- 0	_ 0
##	NP_001135891	NP 612505	NP_065228	NP_689968
##	0	M _012000	36	0
		-		
##	NP_005680	NP_001136202	NP_001245151	NP_001123466
##	WD 600600	0 ND 445705	0	VD 004050
##	NP_689608	NP_115725	NP_001165377	NP_694950
##	0	0	0	0
##	NP_663781	NP_001545	NP_001073982	NP_001013
##	0	0	0	0
##	NP_002563	NP_149163	NP_001107609	NP_001107607
##	0	18	2	2
##	NP_078846	NP_008987	NP_919261	NP_001243370
##	0	3	12	0
##	NP_003038	NP_060116	NP_001689	NP_005477
##	NF_003038	NF_000110	NF_001009	WI _000±11
	_		-	VD 000140
##	NP_065843	NP_001013679	NP_114430	NP_002140

##	0	0	0	0
##	NP_061882	NP_005435	NP_000601	NP_057042
##	0	0	0	0
##	NP_115898	NP 694587	NP_055836	NP_116245
##	_ 0	3	15	1
##	NP_056494	NP_003901	NP_998766	NP_006822
##	0	0	0	0
##	NP_001124198	NP_001123553	NP_002653	NP_006469
##	3	26	26	2
##	NP_071942	NP_036373	NP_001229756	NP_065095
##	0	0	0	0
##	NP_689741	NP_663729	NP_001035350	NP_208382
##	0	0	7	56
##	XP_003960218	NP_057128	NP_955368	NP_001186102
##	0	0	0	0
##	NP_733468	NP_055685	NP_001092198	NP_001506
##	0	16	0	0
##	NP_003784	NP_061720	NP_060292	NP_057047
##	0 ND 06000E	ND 112620	ND 027541	ND 065000
## ##	NP_060085 8	NP_113638	NP_037541 3	NP_065090
##	NP_937802	NP 001171896	NP 006512	NP_001161299
##	NF_937602 0	NF_001171890 3	NF_000312	NF_001101299
##	NP_060093	NP_663786	NP 001010	NP_006067
##	2	M1 _000700	M _001010	M _000007
##	NP_277037	NP_679211	NP_057569	NP_115721
##	27	3	0	0
##	NP_001137325	NP_061326	NP_002370	NP_115598
##	0	- 0	58	36
##	NP_001076047	NP_061193	NP_006322	NP_057673
##	0	0	0	0
##	NP_036330	NP_006735	NP_001912	NP_005015
##	0	0	0	2
##	NP_055321	NP_001131147	NP_619649	NP_003003
##	0	0	0	0
##	NP_056054	NP_689549	NP_000305	NP_055939
##	3	55	0	5
##	NP_689619	NP_060667	NP_955459	NP_065177
##	1	5	5 ND 604046	0 ND 000504
##	NP_005772	NP_001010938	NP_694946	NP_000594
## ##	ND 0030E1	8 8	4 ND 001177002	12 ND 001015052
##	NP_003851 0	NP_003423 8	NP_001177203	NP_001015052
##	NP_006806	NP_061901	NP_056151	NP_620162
##	0	M1_001301 0	M _000101	0
##	NP_149014	NP_057044	NP_000982	NP_001129606
##	23	0	0	0
##	NP_004093	NP_001185986	NP_004848	NP_004822
##	0	0	12	0
##	NP_612423	NP_112182	NP_543023	NP_001026854
##	0	_ 0	14	0
##	NP_003008	NP_859067	NP_000851	NP_005185
##	0	0	0	0
##	NP_059980	NP_872353	NP_694942	NP_001001890

##	0	0	21	0
##	NP_001019801	NP_001026850	NP_055053	NP_004130
##	0	0	10	0
##	NP_001108225	NP_001231319	NP_078838	NP_003148
##	0	0	15	36
##	NP_055565	NP_060009	NP_443167	NP_000651
##	9 ND 000000	3	3	0 ND 004005400
## ##	NP_003229 3	NP_001981 0	NP_002686	NP_001005498
##	NP 078875	NP_005608	NP_115813	NP_001003681
##	6	0	20	7
##	NP_000262	NP 004335	NP 004057	NP_085153
##	3	- 0	- 6	_ 0
##	NP_002014	NP_004375	NP_001257502	NP_071331
##	0	4	4	3
##	NP_001310	NP_079091	NP_775965	NP_003874
##	4	0	17	0
##	NP_006834	NP_116253	NP_006261	NP_036382
## ##	0 NP_076998	NP_065945	0 NP 001250	0 NP_009152
##	NF_070998	10	NF_001230	NF_009132
##	NP 001186583	NP 002344	NP 116321	NP 055515
##	0	0	0	3
##	NP_714913	NP_060818	NP_005898	NP_003493
##	0	5	0	8
##	NP_219484	NP_699179	NP_056136	NP_115568
##	0	2	0	0
##	XP_003846564	NP_061967	NP_060629	NP_054758
##	0 ND 00100000	0 ND 056440	0 ND 000147	ND 064709
## ##	NP_001009880 0	NP_056442	NP_000147	NP_064708
##	NP 060670	NP 076425	NP 005473	NP_001166038
##	0	0	0	0
##	NP_001136	NP_694968	NP_000597	NP_057487
##	0	0	0	0
##	NP_690000	NP_001025067	NP_954655	NP_001135
##	0	0	0	8
##	NP_076990	NP_037359	NP_001271	NP_001264003
##	3 ND 653347	9 ND 070020	0 ND 0703E3	0 ND 445736
## ##	NP_653317 0	NP_078832 0	NP_079353	NP_115736
##	NP_001747	NP_001928	NP_036613	NP_777553
##	0	0	4	0
##	NP_001136271	NP_006768	NP_001177809	NP_056286
##	0	7	0	0
##	NP_006847	NP_001871	NP_001243020	NP_037392
##	0	0	0	0
##	NP_002487	NP_060510	NP_056481	NP_002415
##	0	0	3	14
##	NP_055383	NP_001020330	NP_004346	NP_001182237
## ##	17 NP_001128658	0 NP_057512	0 NP_061919	0 NP_115514
##	NP_001120050	NP_057512 0	NP_061919 0	NP_115514 7
##	NP_690867	NP_001164606	NP_005359	NP_003401

##	9	0	0	6
##	NP_003402	NP_068838	NP_057134	NP_001116
##	6	31	0	0
##	NP_003007	NP_054783	NP_079111	NP_000146
##	0	0	7	0
##	NP 003269	NP 110439	NP_000952	NP_115742
##	_ 0	_ 10	_ 0	_ 0
##	NP 001035250	NP 036271	NP_056453	NP_001185619
##	0	1	14	0
##	NP_060605	NP_061984	XP 003846515	XP_003120317
##	0	111_001304	AF_003040313	AF_003120317
	•	VD 057006	VD 000005	VD 005645
##	NP_060611	NP_057026	NP_066285	NP_005645
##	0	3	0	2
##	NP_001138627	NP_066300	NP_005743	NP_001231684
##	0	7	0	0
##	NP_056369	NP_006805	NP_057279	NP_001230707
##	25	0	17	0
##	NP_001159715	NP_001299	NP_631918	NP_115735
##	0	_ 0	3	_ 0
##	NP_073582	NP 734467	NP 996898	NP_000972
##	0	0	8	0
##	NP 076962	NP_003205	NP_068780	NP 001243589
##	0	N1 _003203	N1_000700	N1 _001240009
	NP 003204	v	ŭ	ND 001036065
##	- ' ' ' ' .	NP_003246	NP_004660	NP_001036065
##	0	0	0	4
##	NP_057329	NP_004522	NP_001510	NP_001488
##	0	0	17	0
##	NP_004767	NP_001027	NP_004857	NP_001493
##	7	4	0	0
##	NP_073576	NP_001186776	NP_065162	NP_057185
##	0	6	6	0
##	NP_006560	NP_001159713	NP_006321	NP_001266289
##	3	37	0	0
##	NP 001266288	NP_065969	NP 060897	NP_003093
##	0	13	8	0
##	NP 000366	NP 057474	NP_775811	NP 919258
##	0	3	M _ 1 1 0 0 1 1	3
	NP_001108479	NP_116575	NP_116231	NP_005315
##				
##	3	30	5	0
##	NP_003520	NP_066403	NP_001013721	NP_000036
##	0	0	0	17
##	NP_004242	NP_055643	NP_001073899	NP_055544
##	0	0	0	12
##	NP_001092268	NP_002417	NP_001026	NP_000531
##	3	16	27	17
##	NP_005792	NP_005866	NP_653244	NP_004283
##	0	0	12	8
##	NP_001361	NP_112242	NP_036357	NP_057033
##	6	0	0	0
##	NP_006690	NP_055671	NP_001374	NP_067052
##	7	M _000071	M1_001374 0	N1 _007032
##			XP_003119309	
	NP_060786	NP_001185787	_	NP_072049
##	10 ND 00011E	54	0 VD 002046E10	0 ND 000116
##	NP_002115	NP_001230894	XP_003846510	NP_002116

	•		•	
##	0	0	0	0
##	NP_001230890	NP_002112	NP_001230891	NP_079225
##	3	0	3	6
##	NP_060911	NP_001093136	NP_078852	NP_006705
##	0	0	0	2
##	NP_060832	NP_004655	NP_071448	NP_002529
##	_ 0	_ 0	_ 0	_ 0
##	NP 001001795	NP 689880	NP 002680	NP_075046
##	M _001001730	0	M1_002000	0 _010040
	•	ŭ	-	•
##	NP_001154	NP_005494	NP_065995	NP_065912
##	5	10	14	5
##	NP_005756	NP_940983	NP_009027	NP_001824
##	0	0	0	0
##	NP_113611	NP_057253	NP_057692	NP_055148
##	8	0	3	0
##	NP 001020272	NP 002376	NP 055135	NP 004461
##	0	4	0	0
##	NP_004467	NP_710163	NP 001020787	NP 112595
	_	_	-	NF_112090
##	46	57	2	5
##	NP_077025	NP_003657	NP_005331	NP_112235
##	0	6	0	0
##	NP_059128	NP_000048	NP_000782	NP_001182572
##	5	0	0	3
##	NP_079514	NP_060267	NP_258412	NP_002820
##	1	4	0	25
##	NP 001171829	NP 001106678	NP_056006	NP_001229774
##	_ 0	_ 0	23	_ 0
##	NP 001128245	NP 003754	NP 001001433	NP_115810
##	0	M1_000701	001001100	16
##	NP 597719	NP 001601	NP 055157	NP_003230
	-	-	NF_033137	NF_003230
##	5	0	VD 005101	2
##	NP_006597	NP_001847	NP_065101	NP_004168
##	0	0	0	0
##	NP_006328	NP_056485	NP_001267	NP_056518
##	21	0	43	0
##	NP_060613	NP_003019	NP_001263990	NP_001263991
##	3	14	0	0
##	NP_037468	NP_076999	NP_776152	NP_001136336
##	_ 0	_ 0	16	23
##	NP 653171	NP 003355	NP_003912	NP_689805
	-	-	- .	_
##	0	0	0 ND 000050	33
##	NP_036553	NP_775821	NP_060259	NP_006818
##	31	23	8	0
##	NP_001544	NP_620775	NP_006432	NP_001186689
##	0	5	0	0
##	NP_037481	NP_000995	NP_000994	NP_006336
##	0	0	0	6
##	NP 002504	NP 056251	NP_612426	NP_001107803
##	0	0	0	6
##	NP_733746	NP_001138740	NP_006233	NP_054765
		_	_	
##	0 ND 000030	ND 006006	ND 0010E46E0	ND 056045
##	NP_009039	NP_996896	NP_001254652	NP_056245
##	0	0	0	20
##	NP_612430	NP_612432	NP_000425	NP_996925

##	6	11	0	0
##	NP_006619	NP_996928	NP_619727	NP_006859
##	0	0	15	0
##	NP_065724	NP_005773	NP_006521	NP_001139643
##	0	0	0	0
##	NP_150280	NP 055302	NP_000132	NP_001138385
##	4	_ 0	11	11
##	NP 001138387	NP_001167538	NP_002002	NP_001156685
##	11	9	11	M _001100000
##		NP 003815	NP_877429	NP_653180
	NP_659547	-	NP_011429	NF_000100
##	0	0	0	2
##	NP_859070	NP_055753	NP_001138947	NP_001008712
##	0	7	23	0
##	NP_001008710	NP_919248	NP_004598	NP_001231496
##	0	0	0	0
##	NP_075565	NP_003560	NP_077002	NP_667339
##	0	0	0	0
##	NP 115644	NP 848613	NP_064582	NP 065726
##	- 53	5	_ 0	_ 1
##	NP 079437	NP_065191	NP 057714	NP 004271
##	20	0	0	0
##	NP 004837	NP 001124439	NP 055230	NP_001630
##	NF_004037	NF_001124439	NF_033230 23	NF_001030
	•	•		ND 001065400
##	NP_054781	NP_689813	NP_057698	NP_001265409
##	0	0	0	/
##	NP_001094896	NP_061159	NP_065104	NP_001553
##	11	38	0	0
##	NP_060218	NP_002203	NP_058642	NP_037374
##	3	0	0	0
##	NP_001092872	NP_003860	NP_690049	NP_775786
##	0	0	0	27
##	NP_001230631	NP_665741	NP_291022	NP_056511
##	6	5	_ 0	_ 0
##	NP 001968	NP 937859	NP 000978	NP 057177
##	11	0	0	0
##	NP 848599	NP_741999	NP 001861	NP 065137
##	0	M1_141000	M _001001	M1_000107
	•	XP_003846787	VD 001701020	OCOCACOO AV
##	NP_004468		XP_001721838	XP_002343932
##	0	6	6 ND 074006	6 ND 074000
##	XP_003403890	NP_653218	NP_071936	NP_071898
##	6	0	0	0
##	NP_002791	NP_116229	NP_115647	NP_004992
##	0	0	3	0
##	NP_005176	NP_057600	NP_620048	NP_004873
##	0	6	44	2
##	NP_001034782	NP_085139	NP_005701	NP_652766
##	0	10	0	0
##	NP_296373	NP 057399	NP_001026883	NP 006379
##	2	0	0	9
##	NP_872418	NP_055494	NP_001230705	NP_057228
##	0	3	3	001220 ∩
##	NP_006577	NP_004976	NP_203524	NP_002515
		_	NP_203524 0	
##	ND 00E224	ND 057602	-	0 ND 004048
##	NP_005334	NP_057623	NP_113641	NP_004948

##	0	0	0	0
##	NP_002154	NP_006113	NP_542937	NP_004383
##	- 11	_ 11	- 3	- 3
##	NP_444511	NP_060255	NP_775826	NP_033665
##	26	29	3	0
##	NP_001035167	NP_071906	NP_003300	NP_001094089
##	0	2	0	2
##	NP_001941	NP_001942	NP_950245	NP_000203
##	0	3	0	13
##	NP_942590	NP_002566	NP_057494	NP_113651
##	0	44	0	0
##	NP_056410	NP_612510	NP_054797	NP_002401
##	34	0	0 ND 057504	17
##	NP_036411	NP_002832	NP_057524	NP_071766
##	0 ND 004706	12 ND 001035111	3 VD 002402776	ND 061006
## ##	NP_004796	NP_001035111	XP_003403776	NP_061026
##	NP_006056	NP_055703	NP_055097	NP_997187
##	NF_000030 3	NF_033703	NF_033097	0 - 101
##	NP 997176	NP_997175	NP 060858	NP_055851
##	0	0	0	0
##	NP_463460	NP 000716	NP_001193844	NP_954855
##	0	4	6	4
##	NP_963891	NP_963890	NP_000717	NP_009130
##	19	19	4	46
##	NP_078975	NP_001164267	NP_710156	NP_006605
##	0	0	3	25
##	NP_116023	NP_001120650	NP_079480	NP_001974
##	8	24	4	1
##	NP_065822	NP_002823	NP_963922	NP_001018000
##	17	0	0	0
##	NP_001074002	NP_005139	NP_077307	NP_060793
##	0	0	0	9
##	NP_055259	NP_478144	NP_079090	NP_115560
##	ND 001003910	25 NP_001003818	0 ND 777540	8 NP_115724
## ##	NP_001003819	NP_001003616 27	NP_777549	NP_115/24 15
##	NP_001138732	NP_775956	NP_000206	NP_003214
##	27	0	20	0
##	NP_060482	NP_056471	NP_002220	NP_001185890
##	27	51	0	0
##	NP_057222	NP_078943	NP_001825	NP_009028
##	0	0	0	0
##	NP_689996	NP_849157	NP_219485	NP_003359
##	0	25	0	4
##	NP_001284	NP_065943	NP_001035807	NP_002096
##	0	2	0	0
##	NP_003503	NP_778235	NP_002097	NP_619541
##	0	0	0	0
##	NP_079350	NP_061888	NP_899050	NP_071925
##	1	34	0	37
##	NP_689729	NP_001865	NP_115726	NP_001508
##	0 ND 0010300E1	4 ND 0010200E2	ND 001050543	0 ND 010047
##	NP_001239051	NP_001239053	NP_001252543	NP_919247

##	14	14	6	8
##	NP_001035247	NP_000387	NP_002397	NP_002735
##	0	0	2	0
##	NP_060739	NP 852480	NP_775782	NP_612435
##	0	0	3	0
	· ·	· ·	_	· ·
##	NP_001171597	NP_003921	NP_060357	NP_001007
##	0	0	0	0
##	NP_078822	NP_116296	NP_001252518	NP_006045
##	23	0	0	0
##	NP_078982	NP_001120716	NP_115646	NP_079337
	_	_	_	_
##	10	16	30	31
##	NP_001127948	NP_004487	NP_004488	NP_001263236
##	12	0	0	0
##	NP_001263248	NP_001171986	NP_004260	NP_001240811
##	0	0	0	0
##	NP_055210	NP_690876	NP 001098673	NP_055143
	0	0 0000	M1_001030075	_
##	· ·	· ·	_	3
##	NP_001265157	NP_057310	NP_476528	NP_853553
##	5	0	0	0
##	NP_114408	NP_695005	NP_997715	NP_954699
##	2	0	0	0
##	NP 001137381	NP_056385	NP 056227	NP_001011713
##	0	M _000000	0	0
	· ·	· ·	· ·	•
##	NP_001346	NP_001138403	NP_689629	NP_003339
##	0	0	0	0
##	NP_001258558	NP_001165365	NP_005578	NP_001180276
##	0	6	6	6
##	NP 002388	NP 057455	NP 061867	NP_060674
##	_ 6	0	8	26
##	NP 036294	NP 006485	NP 001138784	NP_005231
	-	NF_000405	-	_
##	6	0	12	47
##	NP_036250	NP_060741	NP_059116	NP_787071
##	27	13	17	6
##	NP_694992	NP_002519	NP_031388	NP_005080
##	0	0	0	_ 2
##	NP 003089	NP 079140	NP_057517	NP_079356
##	3	3	0	0
				•
##	NP_056278	NP_001078906	NP_079137	NP_612367
##	0	25	29	0
##	NP_005844	NP_150366	NP_077312	NP_077313
##	0	3	3	3
##	NP 001265562	NP_060342	NP_079021	NP_061856
##	3	0	20	0
##	NP_060738	NP_057088	NP_056502	NP_077733
##	5	7	1	0
##	NP_006546	NP_036308	NP_059107	NP_001026849
##	0	0	0	0
##	NP 490595	NP_001766	NP_002612	NP 057136
##	0	0	0	0
##	NP_056492	NP_055679	NP_065850	NP_001203
##	10	0	2	0
##	NP_000973	NP_078912	XP_003960227	NP_001238994
##	0	0	0	21
##	NP_444281	NP_115498	NP_060923	NP_036463
	_	_	_	_

##	0	0	0	0
##	NP_061036	NP_005841	NP_001138785	NP_001025030
##	9	0	43	0
##	NP 001171172	NP_001007528	NP_775858	NP_001207417
##	_ 0	12	3	
	•		-	ND 0041E4
##	NP_001073005	NP_079504	NP_001153880	NP_004154
##	0	9	1	0
##	NP_899058	NP_001129673	NP_001129674	NP_001032895
##	0	0	0	9
##	NP_001002926	NP 055113	NP_060516	NP_001026887
##	_ 0	_ 0	- 12	- 0
##	NP_005988	NP_002787	NP_005928	NP_001036046
	-	NF_002767	NF_003920	NF_001030040
##	0	0	б	0
##	NP_061029	NP_001186673	NP_954630	NP_110381
##	0	0	0	0
##	NP_001120852	NP_001035147	NP_000573	NP_001035149
##	3	_ 0	- 0	- 0
##	NP_001026881	NP_008966	NP_005611	NP 057230
	_		-	NF_037230
##	0	0	0	0
##	NP_001005505	NP_000371	NP_079198	NP_003834
##	37	0	9	10
##	NP_659001	NP_001154178	NP_115522	NP_110386
##	10	10	0	0
##	NP 001852	NP_055501	NP_005379	NP_078800
##	0	17	0	26
	·	=-	v	
##	NP_002654	NP_001007524	NP_009146	NP_071409
##	23	0	0	0
##	NP_001258570	NP_009197	NP_057049	NP_054779
##	3	0	0	0
##	NP 570981	NP 115861	NP 056263	NP 006592
##	0	0	0	0
##	NP 003705	NP 665683	NP 000837	NP_000838
	-	NF_003003	NF_000037	_
##	0	0	0	13
##	NP_714543	NP_037470	NP_002345	NP_060714
##	13	0	0	0
##	NP_066932	NP_066926	NP_064580	NP_001001548
##	0	0	0	0
##	NP_060310	NP_444280	NP_054899	NP_055153
##	0	0	0	0
	-		-	
##	NP_002403	NP_055190	NP_060263	NP_004424
##	0	0	0	0
##	NP_001193545	NP_068577	NP_057543	NP_001165161
##	2	3	0	0
##	NP_001001414	NP 001285	NP_037486	NP_001108072
##	_ 0	3	_ 0	- 0
##	NP 001903	NP 000026	NP_000518	NP_660308
	- .	_	_	N1 _000300
##	0	0	32	0
##	NP_112487	NP_057395	NP_569055	NP_006058
##	0	0	25	0
##	NP_009101	NP_055728	NP_116211	NP_443109
##	22	17	_ 0	0
##	NP_001009	NP_003621	NP_006105	NP_055891
##	0	2	0	6
	-		•	
##	NP_001251	NP_006415	NP_001171469	NP_068766

##	9	40	39	0
##	NP_001003789	NP_009013	NP_057059	NP_003178
##	0	0	0	0
##	NP_056346	NP_057174	NP_000051	NP_001073906
##	21	0	0	24
##	NP_036296	NP 112199	NP_064703	NP_001157940
##	2	4	2	0
##	NP 001008240	NP_659492	NP_079469	NP_006850
##	0	0	24	10
##	NP_001010923	NP_077718	NP 689837	NP_001152850
##	22	0	6	0
##	NP_001152849	NP 001152852	NP_060308	NP_054905
##	0	0	0	0
##	NP_060378	NP_001243338	NP 258257	NP_005362
##	6	5	9	0
##	NP_075422	NP_001168	NP_060734	NP_001138912
##	15	0	19	30
##	NP_067036	NP 056265	NP_001026870	NP_001010982
##	0	5	8	0
##	NP_660358	NP 689497	NP_000980	NP_001005340
##	_ 0	_ 0	- 0	- 0
##	NP_001243353	NP_599030	NP_005536	NP_116120
##	12	_ 0	- 0	- 0
##	NP 004256	NP 008950	NP 054885	NP_114172
##	3	_ 0	- 0	- 6
##	NP_079204	NP_002075	NP_004713	NP_569057
##	- 6	_ 0	- 8	15
##	NP_001092741	NP_075388	NP_002888	NP_058520
##	17	0	0	0
##	NP_002889	NP_001003793	NP_071344	NP_057092
##	0	0	0	13
##	NP_003991	NP_001543	NP_008846	NP_116254
##	53	0	0	40
##	NP_653251	NP_055277	NP_055463	NP_001062
##	24	0	5	5
##	NP_059988	NP_003215	NP_068753	NP_006375
##	0	0	0	0
##	NP_001157650	NP_542407	NP_001028259	NP_001258804
##	5	23	48	0
##	NP_079349	NP_005649	NP_036524	NP_003055
##	0	5	0	6
##	NP_000812	NP_003960	NP_006043	NP_620152
##	10	0	0	0
##	NP_056163	NP_001813	NP_940888	NP_803877
##	38	9	0	0
##	NP_955469	NP_079224	NP_001036148	NP_057501
##	22	0	6	0
##	NP_057178	NP_001128712	NP_001078844	NP_037366
##	0	0	0	0
##	NP_060557	NP_000936	NP_671709	NP_872375
##	4	0	0	6
##	NP_443730	NP_001002837	NP_115488	NP_789762
##	0	26	0	0
##	NP_005673	NP_699163	NP_003234	NP_079058

##	0	0	18	0
##	NP_068759	NP_001037770	NP_478126	NP_057532
##	0	0	0	14
##	NP_055751	NP_848638	NP_003126	NP_001836
##	0	28	0	0
##	NP_203699	NP_000082	NP_612390	NP_061893
##	3	3	N1_012030	19
##	NP_000472	NP 003358	NP_001263302	NP_659500
##	N1 _000472	0	0	N1 _000000
##	NP_835739	NP_079048	NP_001254772	NP_001185882
##	8	10	21	0
##	NP_064550	NP_000085	NP_001167013	NP_001167014
##	12	23	6	6
##	NP 001165358	NP 056448	NP_653247	NP_006536
##	0	0	0	12
##	NP_002486	NP 064543	NP_056312	NP_001103973
##	0	0 00 00 10	0	0
##	NP_219489	NP_006017	NP 115570	NP 001034444
##	0	0	M _110070	11
##	NP 001135865	NP_001230301	NP 776297	NP_001121371
##	0	4	4	8
##	NP_001821	NP_001827	NP_009042	NP_001171610
##	44	0	0	31
##	NP_001073861	NP_001006659	XP 003960485	NP_878908
##	11	26	0	0
##	NP_001156793	NP 444283	NP_078804	NP_001007532
##	14	0	0	0
##	NP 009014	NP 660207	NP 003668	NP 064631
##	6	_ 0	_ 0	3
##	NP 001012772	NP 003320	NP 009172	NP_001096080
##	- 0	_ 0	- 0	35
##	NP 003376	NP 036366	NP 115670	NP_001124151
##	_ 0	_ 0	- 5	_ 0
##	NP_115710	XP_003960183	NP_055130	NP_079049
##	0	6	4	20
##	NP_006507	NP_060717	NP_001017970	NP_612355
##	2	8	52	13
##	NP_835227	NP_001954	NP_660289	NP_002420
##	0	43	6	8
##	NP_612410	NP_114111	NP_001154481	NP_001135942
##	0	0	0	0
##	NP_115722	NP_653287	NP_001138775	NP_443182
##	0	3	5	3
##	NP_059110	XP_001718697	NP_001121557	NP_004865
##	4	6	24	0
##	NP_078898	NP_060932	NP_001120708	NP_001120707
##	25	0	5	5
##	NP_037375	NP_001158729	NP_001444	NP_005242
##	30	28	41	48
##	NP_003101	NP_872319	NP_005552	NP_068595
##	3	0	0	0
##	NP_536856	NP_071327	NP_115992	NP_938020
##	0	0	27	22
##	NP_002438	NP_001003796	NP_006773	NP_061335

	0.5	^		
##	25	0	0 ND 004700	8
##	NP_001161771	NP_001161773	NP_001790	NP_004426
##	3	11	0	0
##	NP_009165	NP_149103	NP_003767	NP_940919
##	0	3	0	2
##	NP_005430	NP_001003701	NP_001137434	NP_981951
##	0	0	17	3
##	NP_113676	NP_060113	NP_001078956	NP_061881
##	29	18	19	5
##	NP 067017	NP 005169	NP_006734	NP_003575
##	- 3	_ 0	- 0	_ 10
##	NP 002513	NP_055856	NP_001154816	NP_001154819
##	37	11	22	22
##	NP 008961	NP 075561	NP_001159435	NP_000213
##	N1 _000301	N1_075301	15	N1_000213
	· ·	-		
##	NP_001123500	NP_006371	NP_065811	NP_055867
##	3	18	1	4
##	NP_004208	NP_003591	NP_005679	NP_076971
##	8	18	59	0
##	NP_060816	NP_004336	NP_958928	NP_001169
##	22	0	1	19
##	NP_783859	NP_620132	NP_004169	NP_000977
##	0	0	0	0
##	NP_115728	NP_000156	NP_660286	NP_056287
##	0	8	1	0
##	NP_036274	NP 000303	NP 054884	NP_078812
##	- 3	_ 0	- 0	19
##	NP 001004127	NP 001245235	NP 001231191	NP_536316
##	14	28	10	0
##	NP 003454	NP 938408	NP 619729	NP_002436
##	0	3	12	24
##	NP 077292	NP 001853	NP 008833	NP_038463
	-	-	-	-
##	6	0	0	37
##	NP_703149	NP_006483	NP_116142	NP_057391
##	31	37	37	0
##	NP_001073930	NP_008913	NP_003915	NP_620689
##	20	37	28	34
##	NP_005160	NP_068745	NP_003021	NP_835471
##	31	37	37	0
##	NP_001186976	NP_689822	NP_877437	NP_001028204
##	0	8	0	0
##	NP_055683	NP_004251	NP_001275	NP_115997
##	17	18	0	0
##	NP_001010904	NP_690610	NP_690611	NP_002957
##	52	3	31	_ 0
##	NP_005783	NP 060699	NP_036223	NP_004321
##	0	0	25	0
##	NP_056040	NP_001129277	NP_005005	NP_000679
##	23	43	NI _003003	33
##		NP_006361		NP_001072987
	NP_001106674		NP_612508	_
##	ND 006003	ND 065120	17 ND 004575	0 VD 002060229
##	NP_006003	NP_065120	NP_004575	XP_003960338
##	0	0	0	2
##	NP_060755	NP_005605	NP_612433	NP_057475

##	0	0	56	0
##	NP_001243469	NP_000337	NP_008872	NP_055402
##	0	0	6	26
##	NP_060889	NP_065154	NP_055763	NP_079103
##	55	39	17	11
##	NP_003146	NP_000230	NP_036371	NP_057056
##	10	0	0	0
##	NP_689918	NP 079416	NP_112236	NP_659480
##	3	0	0	6
##	NP 060091	NP 742034	NP 055184	NP_004573
##	0	8	0	0
##	NP_057142	NP_001070818	NP_001034565	NP 064517
##	0	8	19	0
##	NP_005002	NP_964011	NP_005613	NP_004022
##	0	49	44	0
##	NP_001036100	NP 062457	NP 001093097	NP_001161213
##	0	0	0	3
##	NP 001010891	NP_001036013	NP 005093	NP_077288
##	0	0	0	31
##	NP 000272	NP_001758	NP 001121077	NP_001153865
##	_ 0	12	0	0
##	NP 056166	NP 004580	NP_872584	NP_006085
##	- 0	_ 0	_ 19	- 24
##	NP_036576	NP_005862	NP 001004304	NP_003434
##	- 15	_ 0	_ 0	- 47
##	NP 057484	NP 001186814	NP 060123	NP_001735
##	_ 0	- 0	3	- 6
##	NP_002382	NP_775930	NP_001910	NP_001454
##	0	3	0	5
##	NP_004223	NP_001129491	NP_958437	NP_001819
##	12	0	0	0
##	NP_689490	NP_079230	NP_078865	NP_001034796
##	0	38	5	0
##	NP_003964	NP_073607	NP_060886	NP_001013685
##	0	6	6	0
##	NP_001929	NP_060822	NP_001035088	NP_001014446
##	0	0	20	0
##	NP_004081	NP_006223	NP_002687	NP_115569
##	0	0	0	0
##	NP_004046	NP_060959	NP_001634	NP_852615
##	29	12	0	0
##	NP_114129	NP_079185	NP_001026892	NP_003661
##	5	0	3	6
##	NP_110389	NP_002205	NP_060603	NP_848597
##	16	6	21	3
##	NP_073614	NP_054740	NP_003607	NP_055215
##	0	29	0	0
##	NP_002105	NP_006725	NP_002561	NP_071393
##	20	28	38	4
##	NP_002615	NP_001035926	NP_002414	NP_001311
##	0	0	0	0
##	NP_115664	NP_523353	NP_001437	NP_001243727
##	2	9	0	0
##	NP_003778	NP_064617	NP_061982	NP_116244

шш	0	^	•	0
##	2	0	U 000000	8
##	NP_059105	NP_061824	NP_008828	NP_002139
##	21	24	15	21
##	NP_076922	NP_872604	NP_872606	NP_001230655
##	18	0	0	0
##	NP_002938	NP_115708	NP_536355	NP_060298
##	_ 0	_ 0	_ 0	_ 0
##	XP_001718574	NP 055049	NP_001186136	NP_115976
##	24	N1 _000049	26	43
		•		
##	NP_919255	NP_057163	NP_056997	NP_057547
##	0	0	0	6
##	NP_005501	NP_004605	NP_001258864	NP_061964
##	16	0	0	0
##	NP_000392	NP_001121369	NP_056087	NP_004058
##	19	0	0	45
##	NP_689795	NP 001019907	NP 065747	NP_112181
##	47	0	0	21
##		NP 065390	•	
	NP_378669	-	NP_004039	NP_001138528
##	0	0	0	0
##	NP_065161	NP_443192	NP_078942	NP_071343
##	0	0	0	1
##	NP_006795	NP_009046	NP_071738	NP_005660
##	1	0	0	0
##	NP 612638	NP 006387	NP 006457	XP_003118566
##	_ 0	- 0	_ 0	- 6
##	XP 003403810	NP_005239	NP_003480	NP_002190
##	M _000100010	2	26	0
##	NP_002189	NP 001153620	NP 001153619	NP_001163873
	_	-	-	_
##	20	5	5	52
##	NP_338599	NP_653190	NP_002613	NP_055419
##	1	0	0	0
##	NP_060016	NP_057302	NP_056146	NP_003353
##	0	7	0	0
##	NP 550433	NP 001160165	NP 005926	NP 001526
##	_ 0	19	19	_ 14
##	NP 071929	NP 550438	NP 001012241	NP_078916
##	8	0	0	0
	NP_005244	NP_061974	NP_059965	NP_001137414
##				
##	0	0	0	1
##	NP_694565	NP_003383	NP_110402	NP_031390
##	4	41	45	0
##	NP_065878	NP_001229430	NP_443082	NP_056106
##	11	0	0	19
##	NP_001016	NP_477517	NP_003256	NP_003191
##	0	0	26	11
##	NP_996919	NP_001230161	NP_001230156	NP_001230162
##	0	0	0	0
##	NP_001129611			NP 872431
		NP_001124435	NP_996923	_
##	16	41	0	0
##	NP_002759	NP_443161	NP_115604	NP_060311
##	0	3	9	0
##	NP_000212	NP_078960	NP_061332	NP_079184
##	2	0	3	27
##	NP_116145	NP_036452	NP_001098	NP_004744
	<u> </u>	_	_	_

##	51	0	0	0
##	NP_009151	NP_071384	NP_689557	NP_002346
##	2	0	0	0
##	NP_002928	NP_001265172	NP_001229840	NP_060896
##	0	6	0	2
##	NP_061944	NP_775891	NP_068590	NP_054895
##	4	_ 11	_ 2	_ 0
##	NP 036332	NP_001035533	NP_001128573	NP_940863
##	8	4	0	46
##	NP 001658	NP_653219	NP_036360	NP_115909
##	0	11	0	0
##	NP 056327	NP 001003787	NP 001159441	NP_699166
##	NF_030327 37	NF_001003787	NF_001139441	NF_099100
		•	v	ND 001160E00
##	NP_057709	NP_001258763	NP_004417	NP_001162580
##	0	0	12	0
##	NP_057510	NP_054762	NP_076423	NP_055289
##	29	0	0	6
##	NP_620419	NP_689498	NP_055962	NP_006692
##	0	2	17	0
##	NP_002789	NP_005481	NP_001736	NP_002494
##	0	19	1	0
##	NP_056298	NP_001138368	NP_055072	NP_055082
##	8	2	4	0
##	NP_005202	NP_057259	NP_005639	NP_060212
##	0	15	0	28
##	NP_938080	NP 115649	NP_113618	NP_056525
##	13	_ 0	56	_ 12
##	NP 055622	NP 689529	NP 004278	NP_116019
##	- 23	- 0	- 0	- 0
##	NP 001092755	XP 001126659	XP 003119575	NP_079479
##	0	0	0	18
##	NP 056303	NP 072094	NP_001070867	NP_115712
##	3	3	6	6
##	NP 116242	NP 001026847	NP 776183	NP 115970
##	23	NF_001020047	NF_110103	NF_113970
##	NP_059136	NP 055537	NP 056948	NP_002198
	_	-	-	_
##	29	25	26	57
##	NP_660275	NP_055733	NP_787052	NP_001243750
##	1	25	35	0
##	NP_008843	NP_001243749	NP_055762	NP_004998
##	0	0	0	4
##	NP_689520	NP_060831	NP_057589	NP_001209
##	0	0	0	0
##	NP_060208	NP_054785	NP_001246	NP_689633
##	3	0	12	3
##	NP_078899	NP_036465	NP_775823	NP_001230261
##	9	0	35	43
##	NP_003343	NP_631917	NP_858057	NP_036549
##	0	5	0	0
##	NP_079541	NP_001165883	NP_001098065	NP_055104
##	35	15	- 0	16
##	NP_689506	NP_003943	NP_001181927	NP_775937
##	21	0	20	0
##	NP_001189477	NP_001191019	NP_060843	NP_003717
			000010	

##	0	11	0	5
##	NP_003920	NP_006825	NP_071732	NP_004991
##	- 0	_ 0	_ 0	- 0
##	NP_036588	NP_001005463	NP_076870	NP_001103984
##	0	47	42	47
##	NP_112570	NP_001002251	NP_061164	NP_001265309
##	3	0	0	0
##	NP_001138500	NP_001156787	NP_065880	NP_001034229
##	36	17	39	0
##	NP_005966	NP_061924	NP_066990	NP_000482
##	6	0	0	0
##	NP_005000	NP_996994	NP_149015	NP_055126
##	0	21	9	8
##	NP_071411	NP_000593	NP_110409	NP_006198
##	0	8	9	0
##	NP_060565	NP_002589	NP_056315	NP_115494
##	5 ND 660300	ND 001102001	38 ND 116194	ND 065702
## ##	NP_660309 0	NP_001123201 17	NP_116184 11	NP_065723
##	NP_055452	NP 001138773	NP_001036193	NP_055131
##	N1 _000402	N1_001130773	16	N1_000101 0
##	NP_002765	NP 536737	NP_004641	NP_004918
##	36	36	0	0
##	NP 612449	NP_003255	NP 115675	NP_060753
##	0	56	3	0
##	NP_001073946	NP 001092286	NP_055068	NP_001116151
##	- 22	9	_ 13	- 23
##	NP_060908	NP_000984	NP_001092047	NP_055619
##	0	0	0	10
##	NP_009118	NP_073746	NP_036526	NP_008990
##	0	17	0	0
##	NP_055688	NP_079163	NP_055029	NP_000228
##	0	58	2	43
##	NP_002349	NP_001220	NP_060315	NP_997716
##	0	0	0	0
##	NP_699205	NP_079081	NP_004879	NP_569730
##	ND 000046	0	0 ND 055000	0 ND 071740
## ##	NP_996846	XP_003960228	NP_055098 25	NP_071748
## ##	MD 006333	0 NP_000592		1 NP_570719
##	NP_006323 0	NF_000392 27	NP_004261 2	NF_5/0/19 0
##	NP_660322	NP_079037	NP_054900	NP_919268
##	11	M1_073007	M _004300	N1 _515200 21
##	NP_057105	NP_001035859	NP_714923	NP_110382
##	6	22	0	0
##	NP_060573	NP_116139	NP_612122	NP_055050
##	3	_ 0	16	_ 0
##	NP_543012	NP_001602	NP_001180242	NP_001180240
##	0	0	9	9
##	NP_078959	NP_001139746	NP_001185458	NP_003193
##	3	11	11	16
##	NP_057353	NP_001159591	NP_112573	NP_001128323
##	8	8	28	16
##	NP_067547	NP_061980	NP_003872	NP_079097

##	30	0	0	0
##	NP_001229658	NP_055702	NP_000068	NP_004927
##	27	0	0	4
##	NP_071921	NP_001167618	NP_055294	NP_001207423
##	0	21	44	6
##	NP_002372	NP_004195	NP_116250	NP_077027
##	35	31	15	0
				-
##	NP_006568	NP_001161937	NP_002846	NP_000974
##	8	48	5	0
##	NP_001093115	NP_077273	NP_001034792	XP_003960107
##	0	3	3	3
##	XP_003846583	XP_003960247	NP_899228	NP_060410
##	25	29	30	28
##	XP_003960241	XP 003846431	NP_065885	NP_060937
##	- 36	- 9	- 14	_ 0
##	NP_775323	NP 003125	NP 061914	XP_003960982
##	8	0	3	M _000000002
	_	•	_	_
##	NP_001087195	NP_071449	NP_056503	NP_078928
##	0	12	18	41
##	NP_079095	NP_079104	NP_001161047	NP_054907
##	0	0	0	0
##	NP_036453	NP_000193	NP_115676	NP_065978
##	0	4	3	53
##	NP_001092260	NP_004047	NP_055180	NP_878919
##	1	0	0	8
##	NP_056109	NP_036569	NP 000328	NP_000985
##	_ 20	_ 0	_ 0	_ 0
##	NP 660087	NP 003712	NP 001719	NP_940993
##	_ 0	_ 0	_ 0	0
##	NP_689616	NP 110417	NP 003102	NP_055572
##	17	MI_110117	0	40
		ND 000336	· ·	
##	NP_892010	NP_000336	NP_056307	NP_001017928
##	15	0	20	0
##	NP_002173	NP_001161403	NP_078886	NP_003870
##	13	13	18	14
##	NP_001189446	NP_057413	NP_006484	NP_644812
##	31	33	0	0
##	NP_001180353	NP_002527	NP_149977	NP_067544
##	35	28	6	58
##	NP_001172019	NP_001172020	NP 066934	NP_061910
##	- 58	58	56	_ 0
##	NP_057293	NP_689645	NP_110403	NP_001153890
##	0	28	26	26
##	NP_003656	NP_055316	NP_005629	NP_001172112
##	0	M1_000010	NI _003029	
		·		0 ND 0703E0
##	NP_055165	NP_054753	NP_060624	NP_079352
##	0	0	6	9
##	NP_001240804	NP_115715	NP_006604	NP_001035257
##	9	0	4	22
##	NP_004166	NP_060917	NP_061956	NP_057036
##	0	0	13	0
##	NP_443077	NP_056980	NP_000353	NP_006398
##	0	44	0	0
##	NP_689555	NP_001035972	NP_000979	NP_002903
	_	-		

##	8	0	0	15
##	NP_001053	NP_002886	NP_004699	NP_004811
##	28	45	0	50
##	NP 008854	NP_000534	NP_699196	NP_000233
##	- 0	- 14	- 26	- 7
##	NP_075383	NP_689866	NP_006348	NP_003332
##	19	M1_003000	M _000040	M1_000002
		•	•	-
##	NP_071736	NP_001002901	NP_694957	NP_660339
##	0	18	0	1
##	NP_009231	NP_004562	NP_071345	NP_569056
##	13	0	10	15
##	NP_653284	NP_612157	NP_620157	NP_789784
##	14	0	0	51
##	NP_003335	NP_006661	NP_071329	NP_071762
##	0	_ 0	48	5
##	NP_002650	NP 001005376	NP 653213	NP_001008528
##	N1_002000 0	0	30	M _001000020
	•	•		•
##	NP_940932	NP_001008529	NP_776158	NP_060774
##	0	0	15	59
##	NP_115495	NP_001245139	NP_065849	NP_001156946
##	36	10	50	26
##	NP_733787	NP_001182149	NP_036388	NP_001018121
##	46	43	43	3
##	NP_001265234	NP_006732	NP_620074	NP_001017536
##	0	15	44	37
##	NP_954587	NP_001120836	NP 060494	NP_001174
##	29	29	3	_ 0
##	NP 057204	NP 055223	NP 077270	NP_001092102
##	8	_ 0	_ 0	
##	NP 003245	NP 076933	NP 001243614	NP_055606
##	0	0	33	35
##	NP 005345	NP 060416	NP 003029	NP_004163
	-	-	- .	_
##	0	0	0	20
##	NP_001854	NP_057670	NP_001026873	NP_659489
##	0	5	6	0
##	NP_056998	NP_005970	NP_000794	NP_000795
##	21	0	0	22
##	NP_056349	NP_060065	NP_758869	NP_002380
##	13	0	0	0
##	NP_758860	NP_002614	NP_006638	NP_060209
##	0	0	19	27
##	NP 002389	NP 064545	NP_663622	NP_055589
##	5	12	49	30
##	NP_004356	NP_009140	NP_054902	NP_076945
##	0	0	0	16
##	NP 000446	NP_003103	NP 057161	NP_060910
##	M1_000440	M1_000100	M1_007101	M1_000310
##	NP_055613	NP_113657	NP_683685	NP_061833
##	8	0	0	50
##	NP_004350	NP_000912	NP_001231612	NP_036514
##	1	39	39	0
##	NP_067640	NP_942559	NP_005084	NP_001185608
##	7	31	23	45
##	NP_005178	NP_004877	NP_060687	NP_671722

	F.0	^	4.0	20
##	56	0	42	33
##	NP_036221	NP_003675	NP_116115	NP_060184
##	0	15	56	8
##	NP_476502	NP_001158220	NP_001008398	NP_056049
##	36	25	0	30
##	NP_079420	NP_542382	NP_003202	NP_006278
##	18	8	22	29
##	NP_001027452	NP_061969	NP_004036	NP_003163
##	32	21	0	0
##	NP 660304	NP 065189	NP 055701	NP_001258594
##	- 0	3	_ 0	- 6
##	NP 998814	NP_000308	NP 443168	NP_938023
##	11	3	1	0
		_	-	•
##	NP_001136060	NP_001136061	NP_057579	NP_001137438
##	0	0	18	18
##	NP_061871	NP_004144	NP_443076	NP_006225
##	21	44	11	0
##	NP_009199	NP_001244332	NP_055269	NP_653266
##	4	4	29	31
##	NP_001258523	NP_060678	NP_004835	NP_002510
##	47	27	9	15
##	NP_038286	NP_006701	NP_937832	NP_002406
##	- 33	_ 0	_ 0	_ 0
##	NP 000123	NP 000707	NP_002481	NP_066545
##	3	0	0	7
##	NP 060302	NP 003333	NP_064706	NP_001099043
##	NF_000302 3	NF_003333	NF_004700	NF_001099043
	-	· ·	-	_
##	NP_002927	NP_002996	NP_001017368	NP_000279
##	12	36	24	9
##	NP_653267	NP_001507	NP_001258796	NP_003556
##	8	0	0	45
##	NP_006396	NP_004891	NP_689639	NP_001180218
##	0	17	7	28
##	NP_878907	NP_665906	NP_665910	NP_653280
##	15	3	6	18
##	NP_057016	NP_000050	NP_068839	NP_055101
##	0	51	5	
##	NP_078857	NP_001164263	NP_653314	NP_060182
##	2	17	17	0
##	NP 891549	NP_001177915	NP_115686	NP_001026879
##	16	0	0	15
##	NP_057143	NP_067010	NP_001186137	NP_001073
##	0	49	51	51
##	NP_076433	NP_006562	NP_775846	NP_003327
##	54	0	15	0
##	NP_003328	NP_036412	NP_803183	NP_001025056
##	0	0	0	6
##	NP_001136113	NP_060411	NP_064577	NP_002940
##	8	17	20	0
##	NP_005787	NP_001268	NP_004543	NP_149105
##	_ 0	0	_ 0	_ 0
##	NP_757345	NP_003909	NP_061861	NP_054880
##	0	26	M _001001	0
##	NP_004137	NP_001119800	NP_612457	NP_073750
##	Nt _004191	Mt _001113000	ML_017491	MF_013150

##	0	43	0	0
##	NP_005776	NP_001230618	NP_001017979	NP_115682
##	0	6	0	4
##	NP_057075	NP_115990	NP_005698	NP_835465
##	0	0	26	46
##	NP_085143	NP 005229	NP_001137292	NP_001243224
##	18	3	16	55
##	NP 078796	NP 068762		NP_000090
	-	-	NP_005619	
##	0	0	0	0 ND 004700
##	NP_689648	NP_114105	NP_001138600	NP_001760
##	56	0	6	0
##	NP_570968	NP_001098719	NP_001107593	NP_004385
##	11	44	43	0
##	NP_060343	NP_821066	NP_443131	NP_542776
##	24	44	44	9
##	NP 997719	NP 689622	NP 116213	NP_031370
##	0	_ 0	_ 2	50
##	NP 001158101	NP 612639	NP 115883	NP_115946
##	19	N1_012039	N1_110003	43
##		•	· ·	
	NP_689592	NP_001153242	NP_631916	NP_002683
##	25	3	3	15
##	NP_872307	NP_036394	NP_036460	NP_055675
##	7	15	20	8
##	NP_001078868	NP_777591	NP_057529	NP_003640
##	4	3	20	20
##	NP_004480	NP_940820	NP_001889	NP_001258862
##	3	24	39	10
##	NP_932343	NP_612464	NP_937875	NP_001015
##	6	0	0	0
##	NP 055182	NP 003796	NP 997002	NP_067000
##	_ 0	_ 0	_ 0	_ 0
##	NP 054752	NP 795361	NP 604391	NP 005162
##	0	0	0	0
##	NP 898829	NP 874389	NP 001230587	NP_003098
##	M _030023	N1_014003	N1_001200001	13
##	NP_003099	ND 00074	ND OCEOUT	
	_	NP_008874	NP_005827	NP_001120729
##	20	28	0	5
##	NP_075568	NP_057141	NP_057149	NP_061865
##	3	0	0	6
##	NP_056168	NP_060943	NP_291027	NP_001074008
##	18	0	15	6
##	NP_068801	NP_077302	NP_078843	NP_060326
##	14	3	29	4
##	NP_060377	NP_037507	NP_065141	NP_001003803
##	0	0	0	0
##	NP_065184	NP_003466	NP_076955	NP_005322
##	6	44	0	22
##	NP 061945	NP 003801	NP_060316	NP_001028260
##	0	6	9	24
##	NP 064554	NP_002192	NP_059523	NP_079175
##	0	N1_002132 0	13	31
##	NP 078977	NP_653209	NP_056517	NP_006433
	-	NP_655209	_	_
##	24 ND 066057		27 ND 001005146	0 ND 001204
##	NP_066957	NP_115527	NP_001095146	NP_001324

##	0	0	2	6
##	NP_009044	NP_006488	NP_006667	NP_005180
##	3	18	19	21
##	NP_001017395	NP_060287	NP_258441	NP_001243804
##	7	25	21	27
##	NP 542381	NP 000018	NP_001135764	NP_005496
##	0	0	0	8
##	NP 008924	NP 776171	NP_061872	NP_001092871
	NF_000924 0	NF_//O1/1	_	_
##	•	· ·	16	19
##	NP_076972	NP_004519	NP_689531	NP_066273
##	5	0	6	0
##	NP_004756	NP_001698	NP_036590	NP_115685
##	0	0	0	0
##	NP_001963	NP_001017921	NP_689695	NP_060281
##	0	0	0	0
##	NP 001243725	NP_998760	NP 060836	NP_659473
##	_ 0	_ 0	_ 0	21
##	NP_001032402	NP_079341	NP 002844	NP_115613
##	N1_001032402 51	33	NI _002044	N1_113013 25
##			NP 002405	
	NP_008840	NP_002882	-	NP_116307
##	39	18	0	2
##	NP_001177655	NP_808880	NP_000981	NP_059973
##	2	3	0	0
##	NP_036593	NP_001077430	NP_085053	NP_073737
##	11	36	0	20
##	NP_060899	NP_001138	NP_004086	NP_004246
##	18	40	0	0
##	NP_733832	NP_064539	NP_115865	NP_945341
##	14	0	0	23
##	NP 036545	NP 065833	NP 891847	NP_001073986
##	20	16	31	17
##	NP 001121	NP 113612	NP 653229	NP_001166994
##	0	15	4	38
##	NP 009075	NP 005323	NP 006423	NP_004538
##	0	N1 _003323	NI _000425	N1 _004330
##	NP 079414	NP 665803	NP_001018082	NP 110446
	- ' ' ' '	-	_	-
##	27	25	34	0
##	NP_000223	NP_001183955	NP_001183956	NP_116028
##	35	26	33	33
##	NP_001001551	NP_001014	NP_001107573	NP_005204
##	0	0	0	0
##	NP_612431	NP_852127	NP_056177	NP_001185838
##	17	0	0	0
##	NP_073593	NP_060596	NP_991330	NP_002088
##	0	3	0	19
##	NP_055067	NP_001253	NP_291028	NP_057578
##	23	0	0	0
##	NP_001242954	NP_001230679	NP_116147	NP_001329
##	16	1	0	30
##	NP_001193992	NP 056003	NP_055115	NP_001258848
##	38	30	M1_000110	M _001200040
##	NP_112577	NP_689450	NP_004393	NP_000220
		_		
##	0 ND 005015	ND 001017363	7 ND 002024	ND 00000
##	NP_005215	NP_001017363	NP_003024	NP_008858

			_	
##	11	48	8	14
##	NP_001106878	NP_057158	NP_006224	NP_068805
##	0	0	0	0
##	NP_001019	NP_115548	NP_001243063	NP_002479
##	0	28	7	0
##	NP_079157	NP_005032	NP_057483	NP_055073
##	0	26	0	28
##	NP_004313	NP_064585	NP_071358	NP_001073863
##	0	0	0	26
##	NP_734465	NP 001193541	NP 775901	NP_054754
##	40	- 0	_ 13	- 0
##	NP_003769	NP 997239	NP_068765	NP_001177875
##	24	2	60	43
##	NP 001020128	NP 060945	NP_060134	NP_009216
##	8	6	25	0
##	NP_009209	NP 113600	NP 065827	NP_001020280
##	0	M1_110000	0	3
##	NP 001193570	NP_443158	NP 945187	NP_001035362
##	NF_001193370	NF_443138	NF_945167	NF_001035362 42
	·	NP 005195	NP 001302	
##	NP_005697	-	-	NP_115605
##	0	16	0 ND 0011300E6	50
##	NP_056382	NP_001905	NP_001139256	NP_001139255
##	14	0	0	0
##	NP_061108	NP_064628	NP_054887	NP_003185
##	24	0	3	9
##	NP_950248	NP_055365	NP_068834	NP_002153
##	14	25	28	21
##	NP_031376	NP_115982	NP_001139647	NP_115746
##	8	0	1	7
##	NP_002419	NP_060164	NP_612207	NP_001092287
##	30	11	9	6
##	NP_919257	NP_001186088	NP_068778	NP_525127
##	6	2	0	0
##	NP_001137291	NP_037504	NP_001178251	NP_071914
##	17	0	0	0
##	NP_002026	NP_542763	NP_116205	NP_060740
##	3	0	31	0
##	NP_620154	NP_055724	NP_057211	NP_005006
##	10	20	53	3
##	NP_001926	NP_699162	NP_060436	NP_079030
##	34	15	0	16
##	NP_775758	NP_004771	NP_005971	NP_938051
##	4	0	0	10
##	NP_938052	NP_777588	NP_001351	NP_076419
##	15	3	3	11
##	NP_115662	NP_001703	NP_004354	NP_001807
##	0	5	19	43
##	NP_057513	NP_647475	NP_064511	NP_001129245
##	0	28	35	36
##	NP 065764	NP_060567	NP_003789	NP_071398
##	36	5	44	26
##	NP_001228	NP_006214	NP_001008708	NP_056254
##	M _001220	0 000214	3	N1_000204 0
##	NP_001248760	NP_001366	NP_976033	NP_036527
πĦ	MI _001240100	MI _001200	ML _910033	MF_030321

##	20	0	c	0
##	32	•	6 ND 004004	ND 004400
##	NP_680544	NP_055022	NP_061331	NP_004498
##	17	48	35	0
##	NP_006840	NP_060441	NP_003722	NP_006005
##	3	0	0	0
##	NP_958834	NP_443125	NP_001691	NP_060062
##	8	34	2	0
##	NP_640335	NP_003124	NP_955452	NP_056052
##	7	0	10	22
##	NP_940929	NP_005309	NP_060479	NP_001167555
##	- 0	_ 0	- 6	16
##	NP 683515	NP 064538	NP_003984	NP_001022
##	_ 0	14	36	_ 0
##	NP_078913	NP 699170	NP_443096	NP_004252
##	43	4	M _ 1 10000	0
##	NP_079335	NP_938011	NP 055063	NP 644815
	-	-	-	-
##	32	47	0 ND 005050	0 ND 000474
##	NP_620145	NP_001274	NP_005253	NP_689474
##	3	0	0	0
##	NP_150281	NP_006467	NP_078929	NP_004855
##	0	0	0	31
##	NP_057415	NP_777582	NP_077274	NP_005685
##	40	24	0	0
##	NP_937862	NP_001147057	NP_006011	NP_001002755
##	0	0	27	0
##	NP_835361	NP_001035492	NP_055476	NP_054736
##	6	11	28	0
##	NP_056164	NP_079423	NP_653177	NP_000568
##	47	3	20	_ 0
##	NP 001098717	NP_079018	NP_694988	NP_001001330
##	2	2	20	4
##	NP_150628	NP_005493	NP 057210	NP_060795
##	20	24	0	39
##	NP 115716	NP 057137	NP 004076	NP_055558
##	N1_110710 0	M _007 107	M1_004010	M1_000000
##	NP_054737	NP 001191215	NP_001018016	NP 001191217
	- .	NF_001191213	0 00101001	-
##	0 ND 207467	•		0 ND 050000
##	NP_387467	NP_005612	NP_001014433	NP_059998
##	51	0	0	0
##	NP_219486	NP_775855	NP_001094288	NP_060403
##	0	11	35	38
##	NP_078851	NP_001120972	NP_001186802	NP_660205
##	0	34	25	4
##	NP_149101	NP_001158087	NP_060620	NP_003855
##	0	30	0	2
##	NP_004087	NP_075060	NP_059117	NP_064619
##	0	31	0	7
##	NP_000934	NP_009031	NP_073752	NP_001017964
##	0	0	20	32
##	NP_569735	NP_000646	NP_001241687	NP_060823
##	13	4	56	29
##	NP_997215	NP_787080	NP_940986	NP_079481
##	20	16	3	2
##	NP_004540	NP_115606	NP_660213	NP_112576
ππ	MI _004040	1/1 _112000	NI _000213	111 _112370

##	0	47	0	0
##	NP_001075223	NP_004441	NP_060691	NP_597998
##	18	0	0	2
##	NP_006601	NP_055569	NP_001001655	NP_068818
##	24	9	17	0
##	NP_057043	NP_001075955	NP_004813	NP_001161339
##	0	26	38	0
##	NP_859052	NP_060809	NP_001242959	NP_997251
##	35	0	5	0
##	NP_001122088	NP_001122087	NP_060129	NP_001011663
##	18	18	36	10
##	NP_065876	NP_060607	NP_149017	NP_061825
##	33	15	37	27 ND 070004
##	NP_005506	NP_073149	NP_055436	NP_076921
## ##	57 NP_062458	33 NP_076920	39 NP_002132	33 NP_004493
##	NF_002458 43	NF_070920 37	NF_002132 29	NF_004493 27
##	NP_055435	NP 002138	NP_061826	NP_004494
##	M _000400	N1 _002130 29	N1 _001020	N1 _004434 29
##	NP 115731	NP 003086	NP 778250	NP_001243042
##	30	0	14	16
##	NP 001478	NP 542161	NP 057525	NP_001093392
##	_ 0	_ 0	55	0
##	NP_060936	NP_004964	NP_001599	NP_478066
##	0	15	55	39
##	NP_006682	NP_115875	NP_001009996	NP_036323
##	2	0	14	30
##	NP_000557	NP_852100	NP_777572	NP_149074
##	26	0	26	0
##	NP_002399	NP_056955	NP_997717	NP_065690
##	39	0	0	12
##	NP_001245146	NP_004415	XP_003960943	NP_057206
##	2	42	4	6
##	NP_001127666	NP_612476	NP_783313	NP_004798
##	46	32 ND 073740	0 ND 001044100	36
## ##	XP_003960581 51	NP_073749 18	NP_001244109	NP_004844 4
##	NP_112202	NP_000103	NP_057506	NP_067646
##	NF_112202 0	NF_000103 59	NF_037300	33
##	NP 660294	NP_077269	NP_060288	NP_077283
##	27	6	W _000288	33
##	NP_036523	NP_079221	NP_001096034	NP_005867
##	3	16	0	26
##	NP_001166947	NP_001028741	NP_444269	NP_000135
##	37	7	0	0
##	NP_057197	NP_055739	NP_002351	NP_002350
##	24	8	6	3
##	NP_115904	NP_689895	NP_002644	NP_001191326
##	36	0	34	44
##	NP_001196	NP_116052	NP_009052	NP_001123573
##	12	28	33	40
##	NP_000607	NP_003302	NP_004919	NP_598395
##	11	26	16	0
##	NP_060189	NP_073728	NP_001137544	NP_115497

				•
##	39	36	12	0
##	NP_002862	NP_065746	NP_001068566	NP_775498
##	0	47	31	43
##	NP_055698	NP_542408	NP_001032583	NP_079495
##	24	0	0	4
##	NP 443087	NP_060142	NP_002484	NP_004827
##	8	11	0	44
			-	
##	NP_056943	NP_848934	NP_058636	NP_001230072
##	0	24	16	32
##	NP_057443	NP_060336	NP_059139	NP_660160
##	32	0	0	6
##	NP_002880	NP_110413	NP_001036249	NP_001135757
##	0	16	33	0
##	NP 001017	NP 060727	NP 079133	NP_055228
##	- 0	- 9	15	- 28
##	NP 113679	NP_065139	NP 689565	NP_919442
##	M _1100 / 6	54	3	M _010112
	-		~	
##	NP_003498	NP_003496	NP_001457	NP_001997
##	13	8	14	1
##	NP_127490	NP_444278	NP_002590	NP_005530
##	24	0	0	36
##	NP_002418	NP_663768	NP_116016	NP_001243661
##	36	0	9	2
##	NP_114428	NP_004099	NP_036560	NP_055767
##	32	12	0	28
##	NP 066925	NP_054896	NP 000992	NP_066357
##	_ 0	_ 12	_ 0	- 0
##	NP 113615	NP 000152	NP 036324	NP_001124396
##	0	48	0	5
##	NP 001229338	NP 001122064	NP 001247441	NP_001254527
##	16	N1_001122004 0	N1 _001247 441 8	NI _001204027
		•	_	-
##	NP_000073	NP_114131	NP_001243635	NP_057251
##	35	17	0	0
##	NP_001528	NP_612434	NP_005784	NP_951032
##	0	0	8	0
##	NP_951033	NP_003334	NP_001034933	NP_872357
##	0	0	19	0
##	NP_057284	NP_001154829	NP_001026893	NP_001265239
##	0	9	9	36
##	NP_110429	NP 079305	NP_001078834	NP_444292
##	51	0	0	_ 0
##	NP_002278	NP_689771	NP_110445	NP_068713
##	5	19	23	51
##	NP_036520			NP_001155879
		NP_115705	NP_001121054	
##	0	3	3	25
##	NP_065717	NP_071450	NP_060072	NP_115648
##	38	46	31	3
##	NP_079042	NP_001009944	NP_001001410	NP_000529
##	30	13	4	12
##	NP_001265227	NP_068602	NP_057121	NP_067007
##	- 27	28	9	- 11
##	NP 653237	NP_079016	NP_002609	NP_543026
##	0	23	17	40
##	NP_694983	NP_057437	NP_057146	NP_001156002
ππ	WI _094909	WI _001401	1/1 _001 140	111 _001100002

			•	50
##	11	51	0	53
##	NP_037416	NP_000989	NP_056045	NP_001325
##	42	0	12	0
##	NP_001438	NP_150596	NP_001073965	NP_005035
##	8	0	25	0
##	NP 116175	NP_001005291	NP_001424	NP_056276
##	- 0	51	54	54
##	NP_060375	NP_001013694	NP_203753	NP_694967
##	30	25	M _200700	
				2 ND 470070
##	NP_004737	NP_006318	NP_036451	NP_478070
##	55	3	18	3
##	NP_005457	NP_078867	NP_057130	NP_996769
##	0	0	0	9
##	NP_057151	NP_002280	NP_001231681	NP_671728
##	0	0	29	34
##	NP_116062	NP 006618	NP 114401	NP_001034658
##	56	3	3	13
##	NP 004544	NP_001107	NP_057649	NP_001122307
	0 10 10 10	_	NF_037049	
##	· ·	57	· ·	0
##	NP_000396	NP_037372	NP_004285	NP_061891
##	0	0	27	32
##	NP_001026864	NP_036368	NP_004266	NP_002764
##	10	10	0	0
##	NP_001257323	NP_065835	NP_036528	NP_001678
##	0	49	0	0
##	NP 758455	NP_000485	NP_001503	NP_871629
##	- 0	- 35	_ 10	_ 0
##	NP 077268	NP 003638	NP 976249	NP 057184
##	0	33	11	0
##	NP 997219	NP 008964	NP_001258867	NP_001012987
##	37	M1_000304	38	M _001012307
		•		_
##	NP_001152752	NP_061168	NP_115963	NP_002952
##	8	3	0	0
##	NP_116291	NP_976075	NP_059121	NP_000915
##	18	0	44	39
##	NP_001159447	NP_056190	XP_003960230	NP_001026862
##	39	52	40	38
##	NP_055443	NP_057144	NP_001107605	NP_057109
##	- 3	_ 10	_ 14	_ 13
##	NP_001070648	NP_061921	NP_115899	NP_113662
##	18	40	3	2
##	NP 057588	NP_775872	NP_057488	NP_078924
##	N1 _007300	14	0 - NI _037	24
##	NP_787098	NP_001098117	NP_001020	XP_003846586
##	24	5	0	0
##	NP_722541	NP_001314	NP_001436	NP_003000
##	32	22	50	0
##	NP_612480	NP_004724	NP_001177768	NP_057654
##	17	18	0	2
##	NP_078908	NP_004156	NP_005252	NP_054731
##	- 0	55	- 55	48
##	NP_002815	NP_115607	NP_000122	NP_065190
##	0	0	7	3
##	NP_660295	NP_003491	NP_073594	NP_036605
##	MF_000295	Mr_003491	NF_073394	ML_020002

##	13	49	35	0
##	NP_775919	NP_000205	NP_004995	NP_004442
##	26	46	0	5
##	NP_004443	NP_001191455	NP_001017420	NP_542399
##	49	0	58	12
##	NP 076997	NP 067033	NP_002742	NP_057315
##	- 0	_ 0	- 3	- 2
##	NP_005864	NP 068380	NP 114157	NP_006820
##	36	14	9	0
##	NP 057124	NP_060956	NP_001159894	NP_079375
##	NF_037124 0	NF_000930 7	NF_001139694 42	NF_079373 57
	-	•		
##	NP_060351	NP_004994	NP_612448	NP_004365
##	0	0	0	0
##	NP_060797	NP_001093148	NP_055508	NP_115707
##	60	21	10	15
##	NP_776169	NP_003092	NP_001071639	NP_001071640
##	32	0	0	3
##	NP_001071641	NP_003907	NP_001107648	NP_001182760
##	0	0	26	26
##	NP_699198	NP_001107595	NP_001091110	NP_002025
##	8	16	41	41
##	NP_000141	NP 004060	NP_113601	NP_653192
##	41	004000	M _110001	N1_000132
		· ·		-
##	NP_004657	NP_150592	NP_076957	NP_003603
##	34	9	26	32
##	NP_057179	NP_005759	NP_054890	NP_001017927
##	0	7	0	2
##	NP_001137530	XP_003403688	NP_057520	NP_003453
##	35	35	35	34
##	NP_065798	NP_064573	NP_003809	NP_620134
##	25	0	6	8
##	NP_115893	NP_057098	NP_849144	NP_079508
##	35	55	55	6
##	NP 073208	NP 851030	NP 001157915	NP_112174
##	25	25	42	53
##	NP_653198	NP 060109	NP_001136024	NP_059453
##	0	11	56	0
##	XP_001717815	NP_001157787	XP_003119912	XP_003960344
	_			
##	23	30 ND 500000	27 ND 001177757	23 ND 00100074
##	XP_003960355	NP_588609	NP_001177757	NP_001002274
##	28	49	0	0
##	NP_001257820	XP_003960691	NP_733745	YP_003024029
##	0	0	29	0
##	NP_006867	NP_057133	NP_001091081	NP_620133
##	27	0	1	29
##	NP_997264	NP_001161414	NP_071890	NP_001021
##	6	9	0	0
##	NP_057004	NP_004173	NP_997698	NP_065950
##	- 0	3	45	- 42
##	NP 079051	NP_113650	NP_065145	NP_004863
##	0	2	0	25
##	NP_055060	NP_689578	NP_115879	NP_001165419
##	NF_033000 0	14	NF_113079	001100419
				ND 001065250
##	NP_067652	NP_066979	NP_006173	NP_001265358

##	14	0	58	44
##	NP_001265356	NP_001075106	NP_006856	NP_005865
##	47	16	55	35
##	NP 001074919	NP_079203	NP_001004125	NP_065907
##	- 17	54	_ 18	34
##	NP 116130	NP 065944	NP_001240605	NP_789744
##	3	M _000544	M1_001240000	M1_703744 50
##	NP_060108	NP_112578	NP_006020	NP_001191375
##	0	22	38	0
##	NP_689985	NP_061817	NP_006510	NP_001018060
##	31	29	0	10
##	NP_006518	NP_653201	NP_071894	NP_004217
##	9	48	32	0
##	NP_072088	NP_003122	NP_057485	NP_821158
##	42	3	0	38
##	NP 004770	NP 036592	NP_005793	NP_036351
##	6	0	58	5
##	NP 996261	NP_055239	NP_002505	NP_005082
##	N1 _330201 0	NI _000209	N1 _002303	N1 _003002
	· ·			-
##	NP_057283	NP_064574	NP_114403	NP_009011
##	0	0	0	0
##	NP_060553	NP_004282	NP_001161707	NP_001124336
##	34	2	37	47
##	NP_001258448	NP_001129425	NP_001257965	NP_000560
##	53	2	0	0
##	NP_001158243	NP_003036	NP_003755	NP_114413
##	32	34	21	1
##	NP_057461	NP_061859	NP_065974	NP_061118
##	45	0	41	0
##	NP_079231	NP_079526	NP_777579	NP_001006611
##	54	53	_ 2	12
##	NP 005058	NP 001744	NP 115678	NP_001034761
##	19	0	0	13
##	NP 291035	NP 114124	NP 057550	NP_113640
##	N1 _231033	WI_II4124 4	N1 _037030	NI_113040
##		-	NP_001099033	
	NP_001244030	NP_660143	_	NP_067037
##	13	4	35	49
##	NP_112490	NP_001010989	NP_001628	NP_001170977
##	51	3	28	21
##	NP_000791	NP_078995	NP_116257	NP_057568
##	0	21	22	0
##	NP_899229	NP_055676	NP_004100	NP_060844
##	17	18	0	0
##	NP_940848	NP_055812	NP_066918	NP_004405
##	0	44	10	26
##	NP_001238907	NP_001238911	NP_001776	NP_115656
##	27	44	0	58
##	NP 001541	NP_065078	NP_001138505	NP_061993
##	25	8	0	37
##	NP 061758	NP_061760	NP 061759	NP_061763
##	M1_001760	M _001700	56	56
##	NP_061756	NP_061761	NP_061754	NP_056484
##	NF_001730 45	NF_001701 57	NF_001754 59	NF_030484 59
##	NP_001129691	NP_963857	NP_001555	NP_060520

##	3	9	29	23
##	NP_056203	NP_006262	NP_079412	NP_001017989
##	0	0	6	- 6
##	NP_002719	NP_071319	NP_001093140	NP_001241
##	3	24	3	22
##	NP_001074011	NP_001073379	NP_000900	NP_001128123
##	55	37	60	2
##	NP_001032755	NP_001138536	NP_001138535	NP_005054
##	0	11	12	7
##	NP_620153	NP_004692	NP_659498	NP_620158
##	35	51	34	6
##	NP_005946	NP_060320	NP_659449	NP_071927
##	43	26	3	40
##	NP_689920	NP_477511	NP_001153777	NP_056475
##	27	0	43	56
##	NP_000456	NP_002342	NP_001025035	NP_004772
##	14	1 ND 060204	6 ND 757305	ND 05755
## ##	NP_055047 0	NP_060294 21	NP_757385 5	NP_057556
##	NP 005573	NP 775889	NP_079411	NP_055300
##	NF_003373	NF_113009 4	NF_079411 36	NF_033300
##	NP 001008389	NP 742086	NP 067066	NP_001007468
##	0	M _ 7 42000	M _007000	25
##	NP 031367	NP 057131	NP_055668	NP_005688
##	22	0	20	0
##	NP_640339	NP 055580	NP_065727	NP_001015882
##	23	0	30	16
##	NP_001180583	NP_057145	NP_872329	NP_665805
##	3	0	0	24
##	NP_055217	NP_001472	NP_002603	NP_057339
##	0	31	0	25
##	NP_006275	NP_612427	NP_063949	NP_000990
##	0	22	0	0
##	NP_001001790	NP_036547	NP_001018864	NP_037431
##	0	60	31	25
##	NP_115680	NP_573438	NP_859056	NP_982293
##	32	37	0	45
##	NP_612446	NP_001171656	NP_055653	NP_061960
##	54	56	26	5 ND 070070
##	NP_542387	NP_004327	NP_997191	NP_076973
##	0 ND 001006006	58	ND 057501	ND 006949
## ##	NP_001026886	NP_859064 47	NP_057581	NP_006848
##	3 NP 443731	NP 060211	0 NP_060827	0 NP_776170
##	N1 _ 11 3731	N1_000211 8	35	N1_//01/0 0
##	NP_061899	NP_060935	NP_001001683	NP_005995
##	3	M1_000338	3	0
##	NP_001182225	NP 112494	NP_001252506	NP_085056
##	25	28	42	0
##	NP_733796	NP_068581	NP_001018024	NP_848602
##	3	5	0	5
##	NP_001073966	NP_620777	NP_060489	NP_060921
##	31	- 0	_ 0	7
##	NP_060915	NP_002560	NP_116195	NP_001093254

##	0	57	38	26
##	NP_818931	NP_612206	NP_653320	NP_002988
##	3	6	5	0
##	NP_037403	NP_001534	NP_001073921	NP_061977
##	- 46	29	_ 18	- 5
##	NP 001074260	NP 003159	NP_055616	NP_055976
##	0	0	16	M1_000310
	-	•		
##	NP_150597	NP_001167579	NP_009099	NP_525022
##	0	36	54	47
##	NP_001026909	NP_001162002	NP_115871	NP_001182412
##	21	21	49	54
##	NP_001229668	NP_114109	NP_003842	NP_112571
##	39	0	0	40
##	NP_001182317	NP_079540	NP_775746	NP_056978
##	40	32	24	23
##	NP 115866	NP_690002	NP_775969	NP_036477
##	N1_113000 0	N1 _030002	NI _775363	N1 _030477
	•	· ·		
##	NP_001138424	NP_001014765	NP_904320	NP_000234
##	48	3	11	11
##	NP_054701	NP_002285	NP_001229410	NP_001841
##	0	0	0	12
##	NP_722523	NP_078905	NP_620160	NP_067061
##	31	38	60	31
##	NP_001107561	NP_055087	NP_001028222	NP_115789
##	47	44	44	0
##	NP 995325	NP 001185903	NP 001129970	NP_005098
##	3	- 6	3	_ 0
##	NP_689733	NP 848026	NP 612395	NP_001159506
##	57	_ 0	3	44
##	NP 079047	NP_001182399	NP 115733	NP_055608
##	6	51	0	35
##	NP 997397	NP 001180265	NP 001124334	NP_004656
	-	-	-	_
##	40	51	41	48
##	NP_787077	NP_001842	NP_065157	NP_067032
##	46	38	2	45
##	NP_005405	NP_003944	NP_008868	NP_008867
##	52	5	0	0
##	NP_001002255	NP_056321	NP_872114	NP_954988
##	0	43	31	26
##	NP_079146	NP_001128690	NP_003792	NP_060639
##	35	19	8	48
##	NP 059129	NP_112567	NP_057155	NP_002013
##	31	38	_ 0	51
##	NP_872433	NP_937886	NP 000624	NP 001008749
##	35	0	20	14
##	NP_057046	NP_002384	NP_005515	NP_872309
##	4	49	14	46
				NP_001120680
##	NP_065068	NP_067673	NP_689947	_
##	40	56	8	3
##	NP_006306	NP_919431	NP_787954	NP_054799
##	42	17	3	49
##	NP_004709	NP_659495	NP_996816	NP_002816
##	0	5	33	55
##	NP_115506	NP_003961	NP_006062	NP_203526

##	19	47	25	0
##	NP_003551	NP_000031	NP_078933	NP_055714
##	16	0	9	52
##	NP_004366	NP_835260	NP_060179	NP_002866
##	16	M _000200	19	29
##	NP_003990	NP_078902	NP_005610	NP_689764
##	M _000350	3	14	16
##	NP_060875	NP_056180	NP_001186069	NP_079143
##	0	23	34	18
##	NP_036586	NP 000988	NP_871615	NP_003330
##	37	3	0	0
##	NP 862821	NP 003329	NP 057067	NP 001105
##	0	0	0	34
##	NP_899200	NP_056085	NP 001186571	NP_940878
##	M _033200 51	M _000000	55	33
##	NP 001074318	NP 004255	NP_001882	NP_059140
##	0	0	25	0
##	NP 874384	NP_036562	NP_057160	NP_057108
##	0	M _000002	M1_007100	0
##	NP 000986	NP_057079	NP_112589	NP_002592
##	M _000300	M _007079	53	0
##	NP_940910	NP_003167	NP 005828	NP_001138450
##	M _540510	33	0	M1_001100400
##	NP 001129580	NP 055514	NP_849163	NP_001035530
##	N1_001123000 8	31	19	M1_001000000
##	NP_001158090	NP 004475	NP_003449	NP_689620
##	48	57	15	M _000020
##	NP 115874	NP 002327	NP_002326	NP_001092257
##	7	36	29	5
##	NP_006567	NP 689948	NP 009188	NP 076931
##	37	37	2	0
##	NP 002990	NP 066011	NP_958802	XP_003960473
##	3	3	28	24
##	NP_001131082	NP 203751	NP 001013453	NP 002219
##	45	41	58	6
##	NP_057139	NP 115487	NP 008869	NP_003752
##	- 0	_ 0	- 0	- 0
##	NP_061178	NP_001137146	NP_066996	NP_003867
##	18	- 0	_ 11	_ 0
##	NP_002838	NP_570857	NP_789788	NP_703152
##	19	19	_ 0	21
##	NP_003611	NP_057646	XP_003960602	NP_874364
##	- 58	- 0	- 0	- 7
##	NP_689677	NP_009000	NP_003970	NP_703151
##	22	5	3	5
##	NP_110448	NP_057053	NP_057678	NP_612455
##	23	0	0	5
##	NP_004751	NP_006770	NP_003071	NP_005973
##	25	6	46	52
##	NP_775931	NP_001231693	NP_694566	NP_004590
##	51	11	12	10
##	NP_001953	NP_001008396	NP_001243084	NP_006511
##	50	0	46	0
##	NP_036455	NP_055725	NP_001243794	NP_002768

шш	00	47	40	^
##	23	47	13	0
##	NP_653191	NP_001248370	NP_003351	NP_057385
##	34	33	0	8
##	NP_001157844	NP_004698	NP_078858	NP_001177908
##	14	0	58	4
##	NP_001243333	NP_619527	NP_001191038	NP_001073893
##	13	37	54	36
##	NP_689699	NP_001245177	NP_000380	NP_115714
##	18	17	55	15
##	NP 005969	NP_003723	NP_620175	NP_001229476
##	- 2	- 3	- 17	_ 0
##	NP_002109	NP 001636	NP_001017998	NP_004116
##	0	0	0	0
##	NP_001171694	NP_001007023	NP_667338	NP_112190
	_	_	_	
##	48	48	43	58
##	NP_981942	NP_612402	NP_054889	NP_001004431
##	13	0	46	56
##	NP_060661	NP_001259	NP_001186048	NP_001186049
##	47	37	7	6
##	NP_067076	NP_071436	NP_057717	NP_689967
##	16	30	4	5
##	NP_001034842	NP_064570	XP_003403836	NP_001008910
##	27	36	41	39
##	NP_055293	NP_002064	NP_789786	NP 919224
##	_ 10	- 21	- 48	30
##	NP_079421	NP_001073918	NP 001091953	NP_733752
##	40	23	11	37
##	NP 006166	NP_919290	NP_060807	NP_006143
##	NF_000100 51	NF_919290 46	NF_000007	NF_000143
##	NP_060132	NP_001257313	NP_006263	NP_036577
##	38	60	0	16
##	NP_079001	NP_060934	NP_987091	NP_899243
##	59	0	0	27
##	NP_004066	NP_066940	NP_004551	NP_005003
##	27	46	45	48
##	NP_001103224	NP_001193901	NP_006260	NP_004914
##	24	24	42	31
##	NP_006250	NP_067680	NP_003590	NP_055632
##	5	40	48	43
##	NP_001124171	NP_001243055	NP_000423	NP_002468
##	59	25	7	7
##	NP_001106997	NP 068767	NP 003667	NP_001026788
##	59	45	16	0
##	NP_001002034	NP_036591	NP_060468	
	_		_	NP_061749
##	46	0 ND 064740	42	41
##	NP_061747	NP_061742	NP_061741	NP_061739
##	24	25	43	43
##	NP_003726	NP_061751	NP_061750	NP_061735
##	43	41	41	43
##	NP_061752	NP_002579	NP_061736	NP_061748
##	43	39	37	33
##	NP_061746	NP_061740	NP_061743	NP_061738
##	35	37	24	43
##	NP 061744	NP_006157	NP_005851	NP_149988
	_ : : = : = -	·		_= == = 30

##	41	0	28	27
##	NP_066189	NP_478056	NP_073622	NP_001164982
##	27	0	38	0
##	NP_644808	NP 787083	NP_064571	NP_056183
##	0 0	M1_707000	31	26
##	NP_620168	NP_001171473	NP_001007227	NP_001001664
##	NF_020108 10	10	NF_001007227 22	3
##		NP_005265		NP_699165
	NP_002926 31	NP_005265 0	NP_079034 51	NP_099105 47
## ##	XP 003960213	NP 872304	NP 036454	NP_001629
##	XF_003900213 22	NF_872304 6	NF_030434 0	NF_001029
		-	-	·
##	NP_002695	NP_001257365	NP_002113	NP_291032
##	0 ND 054760	13	0 ND 071030	0 ND 001130460
##	NP_054769	NP_065869	NP_071939	NP_001138462
##	0	31	33	12
##	NP_002596	NP_003710	NP_061929	NP_877435
##	56	54	0	40
##	NP_060600	NP_008844	NP_060053	NP_055917
##	39	27	48	33
##	NP_071355	NP_037525	NP_001257870	NP_689835
##	57	8	19	45
##	NP_060020	NP_659399	NP_000474	NP_006403
##	61	38	0	22
##	NP_001012745	NP_115554	NP_004633	NP_001229247
##	22	6	0	50
##	NP_066928	NP_077008	NP_060034	NP_003087
##	25	55	27	0
##	NP_057622	NP_036426	NP_006334	NP_653181
##	27	41	0	0
##	NP_002483	NP_954697	NP_055296	NP_066991
##	0	36	48	26
##	NP_001138502	NP_116267	NP_001034624	NP_057388
##	54	34	33	0
##	NP_036238	NP_114432	NP_001138341	NP_065699
##	0	52	37	9
##	NP_077289	NP_543136	NP_054726	NP_001321
##	10	34	40	45
##	NP_065698	NP_065166	NP_001035521	NP_061821
##	24	53	0	57
##	NP_001891	NP_060519	NP_003460	NP_848543
##	42	38	41	61
##	NP_055154	NP_065749	NP_065853	NP_003206
##	28	57	35	21
##	NP_789776	NP_065148	NP_064633	NP_001771
##	0	2	0	0
##	NP_110442	NP_005249	NP_612213	NP_254274
##	0	0	33	40
##	NP_001073910	NP_001023	NP_060771	NP_569082
##	59	0	53	59
##	NP_006849	NP_001231121	NP_001231118	NP_001231122
##	3	33	33	33
##	NP_001121170	NP_001255973	NP_057183	NP_001008392
##	16	3	14	46
##	NP_000983	NP_036469	NP_055551	NP_001025160
	_	- -	_	_

шш	0	00	2	10
##	0 ND 540270	20 ND 004440	3 ND 444407	10
##	NP_542378	NP_004419	NP_114107	NP_002169
##	48	9	0	4
##	NP_000945	NP_057232	NP_951009	NP_057538
##	0	44	53	0
##	NP_001092255	NP_001817	NP_001034779	NP_036471
##	5	0	31	0
##	NP_001171506	NP_004677	NP_001094284	NP_061180
##	0	31	12	45
##	NP 777555	NP_001139345	NP 068747	NP_000724
##	- 44	31	_ 0	- 42
##	NP 002480	NP_689830	NP_061039	NP_078828
##	0	29	25	9
##	NP 919267	NP 001136156	NP_001034739	NP_872302
##	N1 _515207	N1_001130130	52	33
	-			
##	NP_061329	NP_061092	NP_077275	NP_071390
##	0	39	3	0
##	NP_085150	NP_006496	NP_005634	NP_001159968
##	45	16	29	11
##	NP_110416	NP_848697	NP_543010	NP_000118
##	3	6	3	37
##	NP_006755	NP_009192	NP_001034786	NP_001073905
##	56	17	41	19
##	NP_114427	NP 001015038	NP_997222	NP_003471
##	32	0	_ 0	_ 0
##	NP_991111	NP_694570	NP 001258837	NP_443714
##	15	29	29	18
##	NP 001138845	NP 060328	NP 005825	NP_001161419
##	53	26	9	9
##	NP 057223	NP_076416	NP 001137453	NP 116295
	NF_037223 0	_	-	-
##	· ·	58	58	0
##	NP_001073988	NP_079193	NP_000352	NP_001092007
##	51	58	52	22
##	NP_942094	NP_001010974	NP_055974	NP_003085
##	46	46	0	0
##	NP_067041	NP_001001481	NP_065877	NP_002440
##	0	15	13	47
##	NP_006839	NP_037380	NP_835225	NP_001093859
##	28	0	8	6
##	NP_001028774	NP_932342	NP_000341	NP_003298
##	24	_ 0	49	25
##	NP_001241667	NP 078907	NP_001034864	NP_078792
##	10	31	58	35
##	XP 003960361	XP_291007	NP_653185	NP_064603
##	23	22	29	38
##	NP 001253994	NP_001253996	NP 001988	
	-	_	-	NP_000291
##	34	59	0	12
##	NP_003135	NP_055132	NP_001092086	NP_710142
##	0	58	33	14
##	NP_783640	NP_055232	NP_055094	NP_001170858
##	44	12	10	44
##	NP_076422	NP_055519	NP_065868	NP_008955
##	44	18	3	0
##	NP_112561	NP_057701	NP_003477	NP_078983

		_		_
##	14	0	6	3
##	NP_060972	NP_689755	NP_775816	NP_060173
##	33	54	45	43
##	NP_061189	NP_612635	NP_443197	NP_009144
##	56	52	0	0
##	NP_620128	NP_116221	NP_055278	NP_078970
##	57	50	0	52
##	NP_055722	NP_112592	NP_640338	NP_001008723
	_	_		NF_001008723
##	1	54	34	
##	NP_112200	NP_115599	NP_619542	NP_001768
##	7	55	43	0
##	NP_536724	NP_005181	NP_002935	NP_001136235
##	30	0	9	24
##	NP_001136243	NP_071937	NP_079452	NP_055749
##	54	55	34	34
##	NP_997186	NP_060169	NP 002428	NP_115740
##	12	55	- 8	41
##	NP_001186	NP_001265536	NP 001637	NP_115830
##	17	17	27	43
##	NP 004170	NP 705838	NP 002496	NP_000302
##	WI_004170 46	M _ 7 00000	N1_002430	N1 _000002
##	NP_006349	NP_001010870	NP 005606	
	_	_	-	NP_003794
##	40	11	18	53
##	XP_003960229	NP_653252	NP_001092253	NP_000766
##	4	35	18	50
##	NP_002925	NP_075053	NP_001001660	NP_066921
##	34	17	17	46
##	NP_066919	NP_694941	NP_001078841	NP_002489
##	61	6	0	44
##	NP_057389	NP_001994	NP_001035887	NP_065695
##	18	30	3	14
##	NP_000542	NP_937799	NP_931045	NP_001265360
##	8	11	45	22
##	NP_000558	NP 065155	NP 006542	XP_003960392
##	7	47	3	39
##	NP 001028174	NP 665698	NP_001193884	NP_783302
##	12	7	10	10
	NP_001010852	NP_001028284	NP_079119	NP_937983
##	_		NF_079119	_
##	45	0 ND 440433		16
##	NP_001157413	NP_110433	NP_001098549	NP_060331
##	28	5	41	52
##	NP_071428	NP_036248	NP_005208	NP_001245249
##	41	58	0	9
##	NP_075051	NP_000012	NP_000438	NP_001121070
##	56	15	36	0
##	NP_478061	NP_065111	NP_060394	NP_612417
##	0	13	3	0
##	NP_056207	NP_065099	NP_945173	NP_002951
##	30	44	_ 0	54
##	NP 079357	NP_689560	NP_001127835	NP_001984
##	43	46	_ 6	41
##	NP 065744	NP_001025051	NP 115696	NP_976043
##	50	39	W _ 110030	0
##	NP_003753	NP_068809	NP_001002759	NP_001004316
ππ	MI _000100	1/1 _000003	MI _001002139	W1 _00100#310

##	37	5	15	18
##	NP_060323	NP_005186	NP_000407	NP_525021
##	12	44	45	51
##	NP_004279	NP_079218	NP_001243342	NP_006071
##	28	40	25	20
##	NP_054796	NP_542172	NP_659494	NP_543011
##	- 2	- 28	3	- 6
##	NP_065186	NP_997246	NP_003587	NP_004420
##	- 6	20	53	53
##	NP_001856	NP_060229	NP_057725	NP_058649
##	0	37	42	39
##	NP_001070665	NP_001070826	NP_061907	NP_001002292
##	42	45	56	27
##	NP_057150	NP_001257513	NP_009186	NP_005075
##	19	21	40	44
##	NP_660310	NP_001154970	NP_008862	NP_703150
##	25	41	5	8
##	NP_689521	NP_000431	NP_444513	NP_619639
##	43	43	11	3
##	NP_001099129	NP_056343	NP_064608	NP_060035
##	35	45	54	47
##	NP_071383	NP_778230	NP_001035755	NP_775741
##	0	32	58	12
##	NP_112211	NP_004749	NP_861452	NP_861520
##	25	42	56	0
##	NP_004326	NP_660299	NP_660327	NP_001160072
##	0	39	36	36
##	NP_061908	NP_002333	NP_060174	NP_116191
##	2	56	20	2
##	NP_001030013	NP_001070148	NP_113634	NP_060420
##	4	6	52	47
##	NP_001230271	NP_006310	NP_002482	NP_006094
##	54	12	0	0
##	NP_002989	NP_057425	NP_006703	NP_001265319
##	39	61	51	47
##	NP_057400	NP_001186018	NP_001186021	NP_079512
##	17	21	29	51
##	NP_060665	NP_055510	NP_078827	NP_510870
##	27	41	55	43
##	NP_001008216	NP_001157871	NP_859065	NP_003623
##	0	12	55 ND 107101	56
##	NP_690850	NP_066951	NP_127491	NP_001002261
##	44	0	50	40
##	NP_848930	NP_036229	NP_003084	NP_065983
##	11 ND 006202	ND 057367	0	0 ND 00000
##	NP_006393	NP_057367	NP_859060	NP_060662
##	0 ND 004064	5 ND 00100000	34 ND 057074	0 ND 004704
##	NP_004861	NP_001098668	NP_057074	NP_001781
##	MD 0610E0	36	52 ND 055006	25 ND 001035545
##	NP_061952	NP_065986	NP_055906	NP_001035545
## ##	33 ND 116000	31 NP_036397	38 NP_001158137	28 NP_861449
##	NP_116000 43	NP_036397 29	NP_001158137 48	NP_861449 54
##	NP_689807		NP_078861	NP_001036096
##	ML_003001	NP_006625	ML_010001	ML_001020030

##	54	5	13	49
##	NP_775798	NP_653249	NP_065824	NP_001188472
##	N1 _773730 47	N1 _033249 42	M _003024 55	26
##	NP_008980	NP_001184162	NP_057017	NP_077296
##	36	36	0 0	NF_077290 15
##		NP_006197	NP_055455	NP_057394
	NP_631957 0	NP_000197 2	NP_055455	NP_057594
##	-	_		
##	NP_065199	NP_001011668	NP_001010922	NP_478072
##	9 ND 055004	9 VD 002046607	ND 000044	57
##	NP_055804	XP_003846627	NP_002244	NP_008911
##	56	19	2	43
##	YP_003024036	NP_620482	NP_065706	NP_001182555
##	31	28	29	31
##	NP_076946	NP_438169	NP_001073159	NP_001123990
##	59	29	31	0
##	NP_061060	NP_036213	NP_002498	NP_937756
##	3	33	40	14
##	NP_001135419	NP_001017975	NP_056305	NP_001818
##	12	59	3	6
##	NP_003349	NP_001004298	NP_653269	NP_054778
##	38	35	0	5
##	NP_001376	NP_056505	NP_690865	NP_075601
##	0	36	5	56
##	NP_001098991	NP_055964	NP_001092903	NP_653321
##	24	18	6	8
##	NP_001018118	NP_066010	NP_006760	NP_004620
##	6	40	26	3
##	NP_001243559	NP_115806	NP_954981	NP_001128476
##	28	25	47	49
##	NP_001034451	NP_005642	NP_006724	NP_612362
##	54	55	0	49
##	NP_055543	NP_008875	NP_031363	NP_004474
##	25	5	13	0
##	NP_001171938	NP_001153695	NP_919307	NP_112177
##	32	46	26	0
##	NP_004483	NP_478123	NP_542390	NP_110422
##	6	53	15	0
##	NP_000484	NP_001264262	NP_001121903	NP_006332
##	22	26	22	18
##	NP_004934	NP_057209	NP_004322	NP_004533
##	54	17	34	_ 0
##	NP 005731	NP_073721	NP_001010875	NP_055240
##	26	58	57	48
##	NP_001033729	NP_004348	NP 061946	NP_002610
##	29	6	53	_ 6
##	NP 002611	NP_775760	NP_001179	NP_057162
##	6	8	13	16
##	NP 660344	NP_001159603	NP_001257935	NP_116166
##	N1 _000344 19	35	17	NI_110100 40
##	NP_003697	NP 002546	NP_714542	NP_060103
##	NF_003097 55	NF_002546	NF_714542 43	NF_000103 32
##	NP_001258769	NP_112192	NP_006809	NP_859075
##	NP_001256769 43	NP_112192 14	NP_006809 48	NP_059075
##	NP_775107	NP_055438	NP_001229735	NP_001770

##	24	0	9	16
##	NP_036416	NP_004206	NP_277048	NP_055244
##	55	8	40	10
##	NP_060347	NP_803237	NP_006084	NP_001096
##	4	3	20	44
##	NP_004603	NP_005944	NP_783316	NP_005943
##	35	_ 0	_ 0	_ 0
##	NP 005941	NP_789846	NP_005937	NP_065192
##	_ 0	- 0	- 3	- 32
##	NP_004536	NP_277055	NP 001092688	NP_443099
##	_ 1	_ 0	32	- 44
##	NP_758872	NP_647479	NP_057608	NP_085128
##	52	53	50	- 48
##	NP_689912	NP_001138483	NP_065874	NP_714922
##	- 46	- 27	_ 10	- 39
##	NP_001878	NP 001106853	NP_055273	NP_699202
##	42	27	35	56
##	NP_078957	NP_689613	NP_008853	NP_001229797
##	53	41	12	19
##	NP_071417	NP_852091	NP_054886	NP_001122084
##	17	15	50	33
##	NP_644809	NP_057551	NP 612437	NP_055324
##	36	38	6	60
##	NP_001072	NP 001005920	NP_055109	NP_001034201
##	56	46	4	31
##	NP_001128522	XP 003846820	NP 477515	NP_689943
##	54	52	13	36
##	NP 001070966	NP_004364	NP 116248	NP_057706
##	49	0	56	18
##	NP_659482	NP 001007795	NP_776190	NP_525023
##	13	37	29	33
##	NP 001243624	NP 115800	NP 740753	NP_001120936
##	32	9	_ 0	30
##	NP 542383	NP 003217	NP_004110	NP_775836
##	42	12	_ 0	59
##	NP 001229757	NP_061725	NP_001135772	NP_689736
##	- 0	9	_ 10	_ 10
##	XP_001717260	NP_036211	NP_054775	NP_001103089
##	- 8	39	_ 0	- 8
##	NP_612815	NP_004265	NP_005117	NP_001129595
##	0	54	54	30
##	NP_054766	NP_005224	NP_004430	NP_078962
##	3	_ 20	- 28	30
##	NP_002222	NP_079022	NP_001120730	NP_149108
##	35	29	48	22
##	NP_116171	NP_443179	NP_001257589	NP_001171808
##	32	12	52	35
##	NP_663761	NP_060536	NP_001073904	NP_037433
##	- 56	_ 26	59	50
##	NP_002980	NP_057083	NP_002702	NP_001129516
##	37	16	52	40
##	NP_085134	NP_006340	NP_000725	NP_619731
##	51	43	47	54
##	NP_932766	NP_715638	NP_071402	NP_078883
	<u> </u>	_	_	_

##	40	32	56	45
##	NP_065988	NP_065920	NP_065138	NP_001003940
##	32	19	18	39
##	NP_001129419	NP_620411	NP_000825	XP_001724434
##	1	38	50	32
##	XP_003959988	NP_001264237	NP_116129	NP_694854
##	22	27	12	47
##	NP_057355	NP_937761	NP_065089	NP_940891
##	36	49	13	42
##	NP_001121143	NP_690878	NP_009038	NP_005728
##	56	43	3	0
##	NP_078968	NP_073729	NP_055393	NP_071378
##	52	1	56	58
##	NP_067082	NP_001010906	NP_001171467	NP_002976
##	55	39	53	16
##	NP_001265665	NP_009098	NP_001041637	NP_001121897
##	24	57	60	60
##	NP_001254514	NP_054772	NP_006486	NP_150375
##	47	53	15	59
##	NP_065203	NP_932348	NP_006414	NP_001230696
##	3	0	0	4
##	NP_065806	NP_056496	NP_056979	NP_004330
##	10	50	49	3
##	NP_443155	NP_077010	NP_060790	NP_006842
##	52	45	25	56
##	NP_002553	NP_001159	NP_690909	NP_775735
##	53	18	56	41
##	NP_001532	NP_689597	NP_079400	NP_001129737
##	49	36	34	14
##	NP_001180386	NP_361012	NP_001241658	NP_036604
##	16	49	19	34
##	NP_065766	NP_071442	NP_004878	NP_055835
##	26	41	32	13
##	NP_000354	NP_001073949	NP_079503	NP_001139541
##	14	53	43	22
##	NP_055187	NP_001136272	NP_001017969	NP_775788
##	46	9	49	4
##	NP_116738	NP_542159	NP_149351	NP_001074020
##	21	41	0	60
##	NP_060887	NP_001017372	NP_076934	NP_001092275
##	31	50	27	45
##	NP_001167596	NP_001858	NP_006418	NP_001065243
##	30 ND 057576	0 ND 001171073	31 ND 070034	ND OFFOCE
##	NP_057576	NP_001171273	NP_078834	NP_055006
##	14	29 ND 006473	30 ND 040851	15 ND 001000E30
##	NP_066287 15	NP_006473 22	NP_940851 55	NP_001020539
##				45 ND 775020
## ##	NP_001020537 45	NP_870998 53	NP_689585 28	NP_775838
				58 ND 690933
## ##	NP_112189 13	NP_110410 53	NP_004754	NP_689823
##			18 ND 060636	NP 00067
## ##	NP_065433 37	NP_001326 50	NP_060636 30	NP_000067 52
##	NP_872352	NP_000864	NP_998768	
##	NF_0/2302	NF_00004	ML_330100	NP_036423

##	48	6	13	60
##	NP_862830	NP_001034485	NP_004526	NP_001128521
##	NF_002000 53	NF_001034465 56	NF_004320 40	60
##		NP_056400	NP_112210	NP_004701
	NP_775259 53	NP_056400 2	NP_112210 16	NP_004701
##				
##	NP_008910	NP_037489	NP_001015508	NP_006279
##	9	0	0	0 ND 004064456
##	NP_085125	NP_001035897	NP_001229303	NP_001264156
##	44	20 ND 055700	39 ND 070004	ND 054747
##	NP_002199	NP_055782	NP_078991	NP_054747
##	25	50	42 ND 056467	15
##	NP_659469	NP_060280	NP_056467	NP_001177693
##	36	37	42	33
##	NP_061166	NP_060694	NP_443079	NP_001373
##	54	48	12	36
##	NP_079057	NP_002761	NP_002760	XP_003960992
##	22	42	23	49
##	NP_058515	NP_057286	NP_003261	NP_001104753
##	57	58	12	13
##	NP_000609	NP_001129396	NP_001706	NP_207646
##	19	27	11	33
##	NP_653309	NP_056519	NP_149132	NP_001242914
##	28	7	29	45
##	NP_997255	NP_065743	NP_000822	NP_000821
##	14	51	30	32
##	NP_002011	NP_689624	NP_078868	NP_004262
##	2	42	32	51
##	NP_004117	NP_068774	NP_001185683	NP_002718
##	32	39	49	43
##	NP_660349	XP_003118897	XP_003119818	NP_150638
##	42	12	12	0
##	NP_006395	NP_003216	XP_003960722	NP_057619
##	16	5	12	33
##	NP_001185947	NP_689663	NP_776195	NP_619646
##	25	7	2	25
##	NP_001157196	NP_004870	NP_005661	NP_003311
##	55	9	18	48
##	NP_620126	NP_055377	NP_653188	NP_848595
##	- 59	50	46	3
##	NP 071400	NP_803875	NP_001074294	NP_060021
##	17	42	46	- 6
##	NP_001129734	NP_006326	NP_001032852	NP_065185
##	0	40	24	36
##	NP_115666	NP_443090	NP_002079	NP_001127965
##	2	54	29	0
##	NP 004966	NP_997194	XP_003960637	NP_008957
##	42	55	27	32
##	NP 996879	NP_683513	NP 073619	NP_001104740
##	52	33	48	43
##	NP_001074016	NP_659475		
##		NP_659475 34	NP_071373 23	NP_112583 22
##	14 NP 061932		XP_001719673	
	-	NP_078953	_	YP_003024030
##	25 ND 979001	50 ND 006105	39 ND 004764	ND 001195970
##	NP_878901	NP_006195	NP_004764	NP_001185879

	_	_		
##	3	9	22	36
##	NP_653170	NP_001264301	NP_981948	NP_004084
##	49	42	24	30
##	NP_004071	NP_938012	NP_116251	NP_001161693
##	12	29	6	54
##	NP_001019845	NP_079040	NP_115512	NP_001165347
##	- 22	- 36	- 0	- 36
##	NP_005468	NP_066550	NP_071379	NP_001027535
##	WI _000400 48	53	11	M _001027000
##	NP_001035546	NP_000187	NP_789790	NP_689630
##	34	51	21	50
##	NP_000918	NP_689986	NP_001015072	NP_848661
##	33	43	38	53
##	NP_001556	NP_060470	NP_001019978	NP_006276
##	16	55	11	53
##	NP_003586	NP_057032	NP_037454	NP_055370
##	- 58	- 56	- 55	35
##	NP_060792	NP 857596	NP_005235	NP_067021
##	34	1	M _000200 52	43
##	NP 940969			NP_001013671
	-	NP_056074	NP_003398	
##	16	57	21	16
##	NP_001030177	NP_001034589	NP_060247	NP_056323
##	6	6	13	36
##	NP_037469	NP_689480	NP_689801	NP_002821
##	20	42	29	31
##	NP_001012986	NP_115536	NP_001029058	NP_078963
##	12	39	2	19
##	NP_945346	NP_005940	NP_110404	NP_001079990
##	52	6	56	17
##	NP 055489	NP 705841	NP_001005404	NP_115732
##	53	41	54	5
##	NP 077271	NP 056487	NP_001025125	NP_001139334
##	3	51	57	34
##	NP_001264993	NP 064455	NP_060166	NP_008969
		-	_	NF_000909 41
##	32	7	27	
##	NP_001073885	NP_476515	NP_004347	NP_001001436
##	22	61	0	53
##	NP_001244220	NP_001139020	NP_001036075	NP_997001
##	49	62	25	8
##	NP_835463	NP_066362	NP_006426	NP_003632
##	47	0	9	9
##	NP_001073970	NP_004885	NP_001120865	XP_003846622
##	48	4	4	33
##	NP_757351	NP_689656	NP_115693	NP_001479
##	45	10	39	46
##	NP_001034636	NP_001129735	XP_003960882	XP_003846818
##	40	37	44	46
##	NP_848594	NP_006666	NP_001073954	NP_001243276
	N1 _040034 25	_	_	_
##		32 ND 210500	12 ND 067639	WD 079003
##	NP_001033043	NP_219500	NP_067638	NP_078993
##	58	44 ND 115000	59 ND 057540	ND 600720
##	NP_001071174	NP_115809	NP_057540	NP_689732
##	47	14	16	45
##	NP_001136155	NP_001026904	NP_085127	NP_733828

##	4	3	54	47
##	NP_660312	NP_002537	NP_005447	XP_003960780
##	N1 _000012	39	13	14
##	NP_524558	NP_689540	NP_001009923	NP_003924
##	0	22	9	31
##	NP_001104768	NP_076426	NP_852610	NP_006040
##	M _001104700	M _070420	M _002010	M _000040
##	NP_001033796	NP_004043	NP_872434	NP_064629
##	33	M1_004048	36	24
##	NP_000824	NP_060689	NP_001180442	NP_598004
##	41	13	50	0
##	NP_001096036	NP_619636	NP 207837	NP_940938
##	33	37	0	4
##	NP_002293	NP 001007560	NP_976054	NP_848642
##	48	0	25	4
##	NP_001239225	NP_443142	NP_006167	NP_775886
##	27	22	9	5
##	NP 055214	NP_005025	NP_001164163	NP_995586
##	6	0	22	21
##	NP_954983	NP_149985	NP_203528	NP 054728
##	- 46	45	54	52
##	NP_055598	NP_000404	NP_004561	NP_055179
##	- 28	46	51	- 0
##	NP_002514	NP_004107	NP_001289	NP_004622
##	57	24	50	14
##	NP_689879	NP_005105	NP_689672	NP_001916
##	20	0	31	56
##	NP_003713	NP_001108451	NP_001012267	NP_064450
##	3	56	28	56
##	NP_005067	NP_001011657	NP_076985	NP_037490
##	44	50	41	34
##	NP_001189	NP_005406	NP_004532	NP_037519
##	20	19	6	0
##	NP_001195	NP_699192	NP_001156896	NP_079333
##	51	41	41	34
##	NP_612439	NP_001027006	NP_003838	NP_599023
##	18	45	46	38
##	NP_005051	NP_065198	XP_370934	NP_001034795
##	10	55	53	56
##	NP_002142	NP_852662	NP_001129467	NP_477521
##	12	50	7	0
##	NP_060804	NP_001025459	NP_689692	NP_001009993
##	39	31	20	2
##	NP_001092302	NP_001177695	NP_002437	NP_001229
##	39	38	13	50
##	NP_001025034	NP_001138904	NP_115688	NP_690008
##	47	0	0	23
##	NP_002968	NP_060838	NP_001025031	NP_115811
##	15	8	22	34
##	NP_000411	NP_037474	NP_057423	NP_056297
##	57	46	17	22
##	NP_872381	NP_001186978	NP_742105	NP_001120838
##	47	60	56	9
##	NP_852124	NP_002575	NP_002407	NP_000325

##	9	9	47	15
##	NP_009206	NP_005041	NP_663783	NP_004702
	_	_	_	_
##	9	46	26	26
##	NP_116190	NP_001138776	NP_001012991	NP_859057
##	49	57	7	43
##	NP_079195	NP_653242	NP_001265374	NP_112566
##	29	55	47	37
##	NP_001026915	NP_115799	XP_003960648	NP_653226
##	54	3	5	37
##	NP_996662	NP_872299	NP_066307	NP_714925
##	10	3	50	28
##	NP_000762	NP_000760	NP_003276	NP_001073895
##	33	49	54	34
##	NP_037451	NP_997329	NP_057263	NP_001254637
##	33	21	16	0
##	NP_004882	XP_003846609	NP_001106873	NP_037487
##	0	12	57	6
##	NP_660323	NP_031374	NP_689658	NP_722576
##	17	60	37	27
##	NP_055579	NP_002891	NP_001229243	NP_037471
##	2	37	17	60
##	NP_612641	NP_060534	NP_057560	NP_001258468
##	12	60	14	58
##	NP_976046	NP_001073929	NP_001167561	NP_115827
##	55	26	6	32
##	NP_002607	NP_006508	NP_000092	NP_001258745
##	57	46	11	26
##	NP_001035197	NP_001073862	XP_003959997	XP_003403509
##	60	55	28	26
##	NP_775933	NP_001020118	NP_071768	YP_003024026
##	25	9	40	6
##	NP_001297	NP_689810	NP_115513	NP_000093
##	8	59	6	36
##	NP_057687	NP_114147	NP_115981	NP_955374
##	46	14	24	0
##	NP_776252	NP_071920	NP_976328	NP_056162
##	33	1	36	35
##	NP_061963	NP_055827	NP_001158502	NP_004097
##	54	18	18	1
##	NP_077817	NP_055213	NP_005859	NP_689829
##	53	49	2	36
##	NP_001258781	NP_039258	NP_006324	NP_542403
##	48	50	46	47
##	NP_001224	NP_000833	NP_001002860	NP_001191335
##	58	32	8	5
##	NP_115501	NP_056179	NP_001154848	NP_001012276
##	44	43	53	48
##	NP_001074299	NP_001035037	NP_848605	NP_064618
##	57	59	10	44
##	NP_955387	NP_003208	NP_001335	NP_005945
##	18	25	16	59
##	NP_055959	NP_055916	NP_000432	NP_775859
##	51	34	31	56
##	NP_694962	NP_001034450	NP_001038945	NP_001035100

##	57	26	55	58
##	NP_036396	NP_004254	NP_060385	NP_001191015
##	46	2	35	26
##	NP_001191017	NP_005636	NP_001012338	NP_054760
##	26	18	54	- 6
##	NP 852607	NP_061909	NP_002509	NP_071742
##	- 51	- 11	49	- 24
##	NP_001940	NP_001071643	NP_001010868	XP_003960127
##	46	53	_ 18	15
##	NP_848611	NP 060321	NP 982278	NP_005672
##	- 59	- 55	38	- 41
##	NP 068368	NP_001380	NP_000483	NP_001094827
##	54	52	42	_ 11
##	NP_055893	NP_079533	NP_001010971	NP_004853
##	45	- 57	- 37	- 31
##	NP_001026896	NP 001243210	NP_001106795	NP_075462
##	38	- 60	- 37	- 37
##	NP_620590	NP_002967	NP_000390	NP_003073
##	0	54	48	44
##	NP_065180	NP_079490	NP 277045	NP_001138466
##	60	25	33	57
##	NP 078919	NP_001171915	NP_940907	NP_004728
##	6	18	29	40
##	NP_689525	NP 055209	NP 660156	NP_001135389
##	59	13	10	52
##	NP_001070731	NP 001139113	NP 683696	NP_705691
##	36	50	49	51
##	NP_872426	NP 004846	NP 060888	NP_998821
##	53	3	42	42
##	NP_001244953	XP 003118768	NP_940912	NP_061157
##	39	13	48	29
##	NP_005718	NP 001265486	NP 064584	NP_003264
##	38	56	46	47
##	NP 001254747	NP_001078940	NP 002149	NP_940684
##	2	- 45	- 43	- 40
##	NP_006064	NP_060019	NP_001027549	NP_060768
##	16	- 56	52	- 38
##	NP_000060	NP_001185521	NP_003467	NP_057359
##	18	56	_ 0	- 40
##	NP_114113	NP_001258074	NP_060399	NP_001238774
##	56	27	- 60	29
##	NP_277040	NP_631902	NP_001092874	NP_001186728
##	10	20	15	29
##	NP_005537	NP_054858	NP_064562	NP_005439
##	51	15	59	_ 0
##	NP_001177658	NP_079534	NP_003629	NP_612369
##	20	58	7	52
##	NP_001012520	NP_004869	NP_001091999	NP_001107047
##	46	16	- 51	32
##	NP_001230678	NP_076938	NP_005972	NP_057617
##	43	43	6	57
##	NP_001128627	NP_004638	NP_060835	NP_201575
##	- 5	- 5	48	- 57
##	NP_006505	NP_073623	NP_077272	NP_001501
	<u> </u>	_	_	=

##	15	36	53	58
##	NP_001838	NP_001035948	NP_001012301	NP_060251
##	21	20	28	4
##	NP_783866	NP_071445	NP_001123552	NP_036603
##	56	57	20	21
##	NP_002356	NP_004139	NP_775970	NP_060352
##	- 54	- 28	36	54
##	NP_001071062	NP_008817	NP_006799	NP_001229434
##	10	3	M1_000733	46
##	NP_113623	NP_001092690	NP_006111	YP_003024028
##	57	27	27	10
##	NP_003027	NP_922932	NP_004935	NP_001159694
##	53	43	21	27
##	NP_003939	NP_001034976	NP_001197	NP_848658
##	40	55	17	50
##	NP_001137236	NP_000705	NP_001017437	NP_056968
##	28	24	10	54
##	NP_699164	NP_071881	NP_000021	NP_001193602
##	21	54	52	48
##	NP_060271	NP_001123482	NP_001264282	XP_003960973
##	47	25	35	54
##			NP_001128504	NP_114106
	NP_689735	NP_002402	_	_
##	58	45	30	35
##	NP_073732	NP_060135	NP_001171925	NP_004723
##	2	47	19	36
##	NP_073599	NP_001298	NP_940857	NP_775924
##	38	3	55	56
##	NP_008959	NP_005942	NP_115507	NP_001129624
##	27	3	53	22
##	NP_064610	NP_542119	NP_061927	NP_076947
##	61	13	46	23
##	NP 001020775	NP 002092	NP 542786	NP_001159432
##	- 8	49	32	9
##	NP_115540	NP 065785	NP_997201	NP_997297
##	54	63	56	56
##	NP_660296	NP_001012418	NP_997210	NP_056963
##	20	3	M _337210	M _000360
##	NP_004618	NP_003250	NP_004787	NP_001258949
##	55	24	41	42
##	NP_001128320	XP_001132754	NP_001131141	NP_001034463
##	44	28	58	39
##	YP_003024031	NP_001157903	NP_689932	NP_001153505
##	7	54	42	53
##	NP_689737	NP_689482	NP_689598	NP_060656
##	53	57	44	57
##	NP_001245344	NP_060363	NP_008940	NP_776155
##	58	55	21	27
##	NP 002137	NP_620129	NP_001158283	NP_001167594
##	44	58	46	36
##	NP_116174	NP_001011552	NP 998885	NP_115881
##	55	58	2	20
##	NP 695002	NP_001159349	NP_036321	NP_057682
##	NF_093002 47	NF_001139349 15	NF_030321	NF_037002 58
##	NP_015564	NP_899230	NP_001034931	NP_005932

	10	47	4.0	•
##	18	47	10	9
##	NP_001258635	NP_060047	NP_201591	NP_001013716
##	24	0	42	29
##	NP_000360	NP_055507	XP_001716463	NP_775966
##	20	27	0	56
##	NP_001008777	NP_064625	NP_001070142	NP_004824
##	51	20	39	10
##	NP_037412	NP_076869	NP_001170787	NP_002033
##	36	9	34	21
##	NP 777550	NP_149109	NP_056219	NP_954870
##	35	_ 0	52	- 3
##	NP 954863	NP_689627	NP_060749	NP_079497
##	3	41	46	55
##	NP_114095	NP_059112	NP 062815	NP_001073887
##	N1_114035	N1_033112 29	N1 _002013	N1 _001073087
			-	
##	NP_002771	NP_002775	NP_115753	NP_001103373
##	39	39	0	24
##	NP_078853	NP_115510	NP_694573	NP_542193
##	49	15	15	12
##	NP_005651	NP_612459	NP_004211	NP_061962
##	21	0	48	24
##	NP_001774	NP_003018	NP_659402	NP_000209
##	55	6	51	37
##	XP_003960863	NP_114121	NP_004818	NP_001127846
##	59	52	8	22
##	NP_788954	NP_003323	NP_001092878	NP_001007793
##	29	40	19	57
##	NP 001248364	NP 751897	NP 001130	NP_061846
##	_ 14	54	- 11	16
##	NP 078848	NP 006681	NP 001445	NP_940918
##	34	40	44	30
##	NP 001123470	NP 071434	NP 115695	NP_000740
##	M1_001120470 50	26	46	M1 _0007 40
##	NP 061722			NP_060512
	-	NP_776245	NP_006821	_
##	ND 001100710	12 ND 604571	ND 054076	28 ND 0049E9
##	NP_001182719	NP_694571	NP_054876	NP_004852
##	21	49	37	27
##	NP_001013049	NP_055989	NP_001003897	NP_060766
##	56	44	34	57
##	NP_075555	XP_003960164	NP_001153047	NP_002337
##	45	53	37	60
##	NP_543013	NP_005997	NP_775844	NP_065083
##	36	26	32	0
##	NP_006762	NP_003761	NP_056483	NP_114104
##	6	8	10	55
##	NP_001171672	NP_001091997	NP_003836	NP_789791
##	22	56	62	23
##	NP_001264129	NP 612412	NP 742093	NP_776194
##	13	7	57	59
##	NP 001138482	NP_570124	NP 079321	NP_060946
##	32	60	39	52
##	NP 004874	NP 001035706	NP 001129131	NP_620072
##	M1 _004074 47	17	58	58
##	NP_849158	NP_892030	NP_078936	NP_001034930
##	ML 043100	ME _032030	ML_010930	ML _00109#320

##	13	56	58	30
##	XP_003846812	NP_001034892	NP_001243528	NP_116136
##	56	14	34	24
##	NP_955777	NP_061831	NP_001180247	NP_004056
##	26	56	56	38
##	NP_005459	NP_067650	NP_109377	NP_653273
##	26	32	44	34
##	NP_775908	NP_659430	NP_775815	NP_803251
##	47	36	27	54
##	NP_005059	NP_001007091	NP_065394	NP_085076
##	- 44	- 47	- 24	- 47
##	NP_001159365	NP 001652	NP_443196	NP_071753
##	23	0	42	0
##	NP_060743	NP_443071	NP_001294	NP_057051
##	16	N1 _445071 50	N1_001234	34
##			·	
	NP_002458	NP_005238	NP_060147	NP_835231
##	58	14	56	51
##	NP_001004432	NP_054729	NP_003314	NP_060926
##	58	26	54	31
##	NP_009157	NP_001421	NP_001035267	NP_006033
##	39	15	6	50
##	NP_473365	NP_003826	NP_076964	NP_002404
##	13	41	59	27
##	NP_057181	NP_001171532	NP_001071121	NP_002633
##	12	53	60	29
##	NP_000470	NP_001004489	NP_001138671	NP_036537
##	40	9	40	46
##	NP_060805	NP_057324	NP_067025	NP_001129042
##	- 47	- 61	- 15	- 49
##	NP 976223	NP 001137150	NP_003136	NP_000071
##	57	28	25	6
##	NP 963925	NP 689603	NP_001185839	NP_060330
##	26	57	23	M _000000
##	NP 683691	NP 659411	XP_002344489	NP_695001
	-	-	_	_
##	57	46	16	31 ND 004761
##	NP_002763	NP_116317	NP_055794	NP_004761
##	31	29	50	49
##	NP_001185837	NP_001027004	NP_001186907	NP_937790
##	23	48	53	13
##	NP_115923	NP_065775	NP_001129575	NP_001010903
##	36	0	24	37
##	NP_004378	NP_683684	NP_001159439	NP_057219
##	59	31	40	56
##	NP_001094348	NP_997254	NP_003386	NP_742055
##	0	23	26	53
##	NP_001181887	NP_055117	XP_003403796	NP_001514
##	62	4	45	55
##	NP_006059	NP_612424	NP_001177945	XP_001716411
##	58	19	8	- 57
##	NP_776159	NP_003562	NP_653303	NP_001245240
##	17	52	51	61
##	NP 001128130	NP_077017	NP_000184	NP_006570
##	M _001120100	14	M1_000184 47	35
##	XP_003961041	NP_001191125	NP_872333	NP_060307
##	VL _002201041	NF_001191125	ME _01 Z333	ML_000201

##	19	51	25	12
##	NP_443118	NP_443724	NP_079273	NP_068775
##	43	3	50	27
##	NP_663696	NP_001139313	NP_778232	NP_699207
##	50	12	29	56
##	XP_003960847	NP_001008536	NP_005418	NP_067003
##	- 54	- 48	- 15	56
##	NP_058644	NP_005240	NP_060370	NP_001177370
##	57	51	48	31
##	NP_001035194	NP_671723	NP_001161362	NP_004952
##	M1_001000154 9	39	34	15
##	NP_001139108			NP_001017403
	_	NP_036386	NP_660329	NP_001017403
##	47	4 ND 070044	47	
##	NP_001017404	NP_076941	XP_373277	NP_116053
##	41	45	14	54
##	XP_003846524	NP_001123885	NP_569157	NP_443175
##	0	43	51	29
##	NP_955369	NP_056002	NP_001857	NP_689975
##	13	11	15	37
##	NP_002176	NP_001164626	NP_443073	NP_001193949
##	36	44	0	50
##	NP_660352	NP_001138779	NP_037464	NP_443165
##	37	39	60	57
##	NP 001121687	NP_115950	NP_001026907	NP_620169
##	12	24	39	42
##	NP 945193	NP_055281	NP_115743	NP_777569
##	5	49	55	27
##	NP 653326	NP 710141	NP 065954	NP_775806
##	1	47	45	M1_770000
##	NP_001264236	NP 001243586		NP_004786
	_	-	XP_001129515	_
##	56	4	41 ND 000040	59
##	NP_000991	NP_001093391	NP_002243	NP_542787
##	10	56	51	35
##	NP_001164102	NP_689846	NP_057700	NP_001288
##	11	56	19	61
##	NP_000551	NP_742099	NP_002659	NP_071765
##	49	50	34	53
##	NP_078945	NP_940893	NP_055716	NP_115995
##	33	51	22	29
##	XP_003960169	NP_006813	NP_777552	NP_542778
##	38	48	60	26
##	NP_115996	NP_849150	XP_003960895	NP_001124183
##	15	55	59	58
##	NP 699167	NP_001004456	NP_002034	NP_057156
##	15	0	23	53
##	NP 001005234	NP_001017980	NP_001087199	NP_068575
##	47	53	17	43
##	NP 653322	NP_808211	NP_001093260	NP_005622
##	NF_033322 53	NF_000211 53	NF_001093200 48	NF_003022 59
##	NP_001138640	NP_006845	XP_003960233	NP_003262
##	55 ND 001004330	57	24 ND 476500	57 ND 040640
##	NP_001004339	NP_443116	NP_476509	NP_848648
##	52	54	42	58
##	NP_001792	NP_115928	NP_149040	NP_775928

##	17	23	27	24
## ##	NP_001185982	23 NP_001001850	NP_005369	NP_001013642
##	15	NF_001001630 44	38	NF_001013042 53
##	NP_073618	NP_848578	NP_996805	NP_001161940
##	NF_073018 22	NF_040378	33	NF_001101940 22
##	NP_997625	NP_003270	NP_078879	NP_060152
##	NF_997025	NF_003270 0	NF_070079	23
##	NP_036314	NP_849155	NP_001012414	NP_612396
##	NF_030314 47	NF_049155 55	NF_001012414 56	NF_012390 54
##	NP_001073947	NP_001486	NP_001013757	NP_001158162
##	NF_001073947 58	NF_001480 22	NF_001013737 47	NF_001136102 42
##	NP_057655	NP_937858	NP_005363	XP_003118572
##	NF_037033	NF _937638 58	NF_003303	29
##	NP_001009606	NP_057125	NP_036318	NP_066009
##	NF_001009000 59	NF_037123 28	NF_030318 47	25
##	NP_653179	NP_068741	NP_112590	NP_872416
##	NF_033179 5	NF_000741 21	WF_112590 40	NF_872410 47
##	NP_001007596	NP_002460	NP_115956	NP_006685
##	NF_001007390 40	NF_002400 15	NF_113 <i>9</i> 30	NF_000083
##	NP 004464	NP_036224	NP 001092808	NP_694948
##	M _004404 47	37	13	M _034340
##	NP_000610	NP_699178	NP_955523	NP_006140
##	10	3	46	M _000140
##	NP_001138229	NP_068835	NP_001906	NP_112509
##	53	3	57	48
##	NP_065933	NP_872308	NP_055176	NP_997279
##	M _000388	30	49	50
##	NP_060529	NP 001129479	NP_004051	NP_114404
##	50	37	40	55
##	NP_612382	NP_003927	NP_061841	NP_001014979
##	11	57	31	41
##	NP_848561	NP_000331	NP_690591	XP_003960234
##	11	52	52	45
##	NP_000237	NP_057387	NP_004187	NP_065087
##	59	28	47	46
##	NP_004953	NP_001034806	NP_065119	NP_110392
##	- 38	53	38	- 44
##	NP_620138	NP_062558	NP_001005163	NP_665813
##	- 63	- 27	52	_ 10
##	NP_005279	NP_998761	YP_003024032	NP_001165887
##	36	51	53	_ 22
##	NP_001010893	NP_872303	NP_000871	NP_690872
##	32	50	34	45
##	NP_683762	NP_060064	NP_699174	NP_001164045
##	26	47	27	44
##	NP_001193600	NP_061134	NP_932347	NP_003593
##	15	39	47	31
##	NP_997203	NP_001191293	NP_775791	NP_004065
##	54	34	58	- 60
##	NP_068752	NP_219494		
##	55	57		

I have already used the sumarry() in the previous section

According to the plot we see that we have 10% of missing values in the dataset. Without suffering too much of a loss, we can exclude all variables with >25 percent missing data. Using the mean, the remaining missing data can be imputed (a more sophisticated form of imputation would be preferable but is quite computationally expensive and we dont have a huge amount of missing data, so I stuck with means in my analysis).

- 1. Discarded variables with more than 25% of the data missing because they wouldn't be significant for further research.
- 2. Used the mean imputation method, we impute the NA values with the mean of a specific column for the remaining variables with missing data. (for which i have implemented the for-loop which did the job)

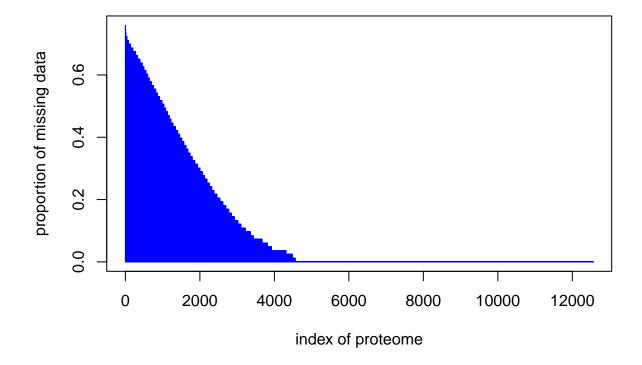
Step 1 - Missing data : The code below counts missing data by column and plots a graph to show the proportion of missing data for each variable.

```
#looking for proteomes with many NAs
naCounts <- colSums(is.na(proteomes)) / nrow(proteomes)

#plotting missing data proportions

plot(sort(naCounts, decreasing = TRUE), col = "blue", type = 'h', xlab = "index of proteome", ylab="prop"</pre>
```

Propotion of missing data for each proteome



```
#how many have more than 25% missing data
length(naCounts[naCounts>0.25])
```

[1] 2251

Without suffering too much loss, we can exclude all variables with >25% missing data. Using the mean, the remaining missing data can be imputed (a more sophisticated form of imputation would be preferable but is quite computationally expensive and we dont have a huge amount of missing data, so I stuck with means in my analysis).

Step 2:

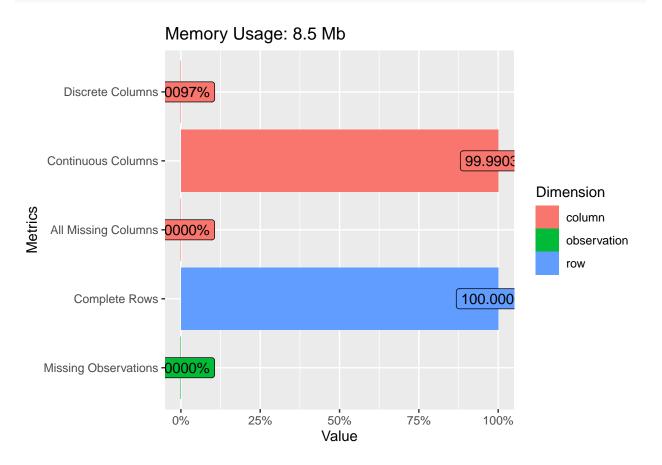
```
#remove variables with >25% missing data
proteomes <- proteomes[ , colSums(is.na(proteomes)) / nrow(proteomes) < 0.25] #removing variables with
#loop to impute means for remaining missing data
for (i in which(sapply(proteomes, is.numeric))) {
    proteomes[is.na(proteomes[, i]), i] <- mean(proteomes[, i], na.rm = TRUE)
}</pre>
```

Now the dataset is clean, Lets explore if there are any missing values

```
dim(proteomes) # a total of 2251 variables are removed
```

[1] 83 10303

plot_intro(proteomes)



The proteome dataset is now clean, Therefore, I have now joined the proteome dataset and clinical datset using inner_join() from dplyr package.

```
#inner join on data to create full data set
data <- inner_join(clinical, proteomes, by = "Complete.TCGA.ID")
#replacing lengthy col name
colnames(data)[3] <- "diag_age"</pre>
```

Exploring the final datset

generated.

```
dim(data)

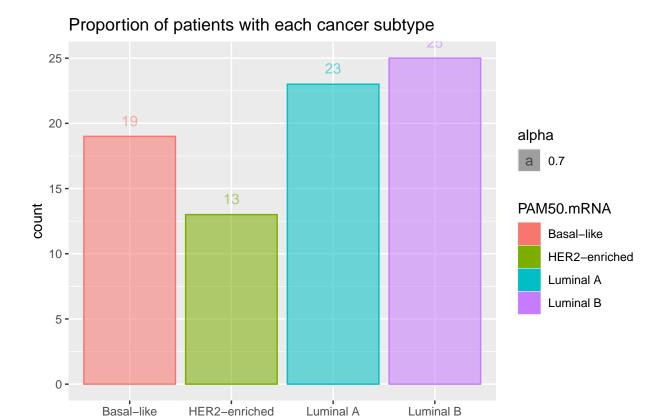
## [1] 80 10332

#head(data)
#tail(data)
#str(data)
```

Evaluation of distribution

The main idea of the project is to check if the proteomic datset can classify the subtype of breast cancer. Therefore, it is important to check the number of observations in each subtype The plot below shows how many patients have each subtype of breast cancer.

```
#Barplot of subtypes
ggplot(data, aes(PAM50.mRNA, col = PAM50.mRNA, fill = PAM50.mRNA, alpha=0.7)) + geom_bar() + ggtitle("PAM50.mRNA, alpha=0.7)) + geom_bar() + ggtitle
```



Therefore, They are reasonably well balanced, although HER2 is slightly underrepresented.

PAM50.mRNA

3. Data Cleaning & Shaping

Data Imputation

- Data imputation is already done in previous chunks
- Mean imputation for proteome dataset is done

Proper Encoding of Data

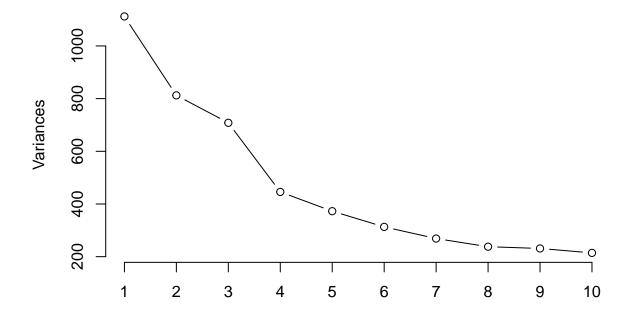
- Encoding was done for only PAM50.mRNA column
- PAM50.mRNA is categorized into four types "Basal.like", "HER2-enriched", "Luminal.A", "Luminal.B"

Normalization/Standardization

- Normalizing the data does not make any difference in the predictions as the protein expression data already ranges on -1 0 1 scale ### Feature engineering PCA
- Principal component analysis is also done using prComp function
- \bullet Principal components are taken into consideration only when the cumulative variance is greater than 85%
- To get the cumulative variance of 85 or greater, I was forced to select 48 components

- But i want to know the list of features (proteins) important of the classification of breast cancer subtype, Because of this I haven't used Principal components for my models ### Feature selection - repeated lasso regression
- I selected variables using repeated lasso regression as my method. -A total of 30 proteins were selected over more than 20 times.
- These proteins are taken into consideration for further analysis
- Checked the distribution of cancer subtype using these proteins

Plot of the Principal Components



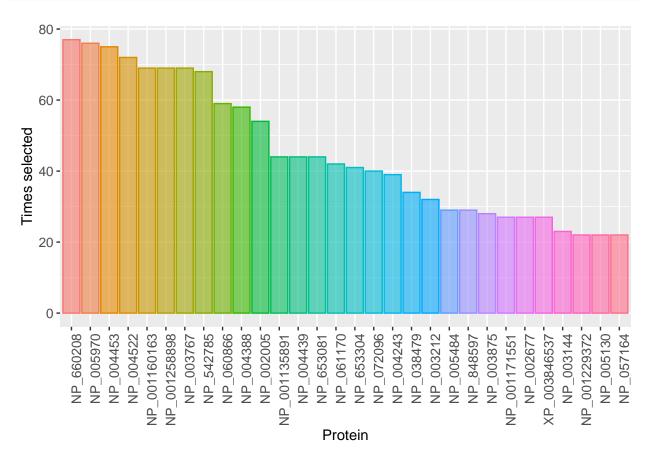
Feature selection: repeated lasso regression

The PAM50 genetic test is used to identify the subtype of breast cancer. A list of proteins linked to the PAM50 genes is included in this data collection. Therefore, it would appear likely that these are the best factors to utilize when categorizing breast cancer subtypes. To test if machine learning techniques might be used to find a set of proteins with as good or higher prediction power at classifying cancer subtypes, I chose an approach that was independent of biology. I'm selecting variables using repeated lasso regression as my method. The data set is reduced via lasso regression, which also creates a sparse set of predictor variables. However, because of the stochastic nature of the reduction, the findings are not always reliable. To get around this, I conducted 100 iterations of the lasso regression and prioritized the variables according to how frequently they were used in the final model.

```
# Defining a function that performs lasso regression again and returns the chosen model variables
LassoSub=function(k=1, Xdata, Ydata){
  set.seed(k)
  s=sample(nrow(data), size=0.8*nrow(data))
  Xsub=Xdata[s, ]
  Ysub=Ydata[s]
  model.sub=cv.glmnet(x=Xsub, y=Ysub, alpha=1, family="multinomial") #cross validated lasso
  coef.sub=coef(model.sub, s='lambda.1se')[-1] #using lambda +1se hyperparameter value for parsimony
  return(coef.sub)
}
options(warn = -1) #turn off warnings
#Run model 100 times and save results
niter=100
lasso.stab=sapply(1:niter, FUN=LassoSub, Xdata=as.matrix(data[,31:ncol(data)]), Ydata=as.matrix(data[,2
#create a matrix of all predictor variables
stability_matrix <- matrix(nrow=length(lasso.stab[[1]]),ncol=length(lasso.stab))</pre>
rownames(stability_matrix) <- rownames(lasso.stab[[1]])</pre>
#loop through to put list contents into matrix
for (i in 1:300){
  temp.data.frame <- as.matrix(lasso.stab[[i]])</pre>
  stability_matrix[,i] <- temp.data.frame</pre>
}
stability_matrix <- ifelse(stability_matrix != 0, 1, 0) #Replacing beta values with binary 1/0 (selecte
stability_matrix <- stability_matrix[2:nrow(stability_matrix),] #remove intercept value
stable_variables <- as.data.frame(rowSums(stability_matrix)) #create data frame with count of how many
stable_variables$protein <- rownames(stable_variables) #create column of variable names
colnames(stable_variables)[1] <- "times_selected" #assign appropriate column name</pre>
stable_variables <- stable_variables[!is.na(stable_variables$times_selected),] #remove NAs
stable_variables <- stable_variables[stable_variables$times_selected != 0,] #remove all variables that
stable_variables <- stable_variables[order(-stable_variables$times_selected),] #ordering by number of t
```

visualizing the selected features

#plotting stable variables
ggplot(stable_variables[1:30,], aes(x=reorder(as.factor(protein),-abs(times_selected),mean), y=times_se



```
STABVARS <- stable_variables$protein[1:30]

STABVARS.ind <- which(colnames(data) %in% STABVARS)
```

We now have a collection of variables that the lasso regression repeatedly chose. Due to the size of the data set, instability still exists after 100 iterations, and only roughly 30 variables were chosen more frequently than 20% of the time. These are the factors that we will classify using.

An indication of how well the chosen protein variables will be able to categorize the subtypes will be provided by visualizing the relative amounts of the most-selected protein in patients with each subtype of cancer:

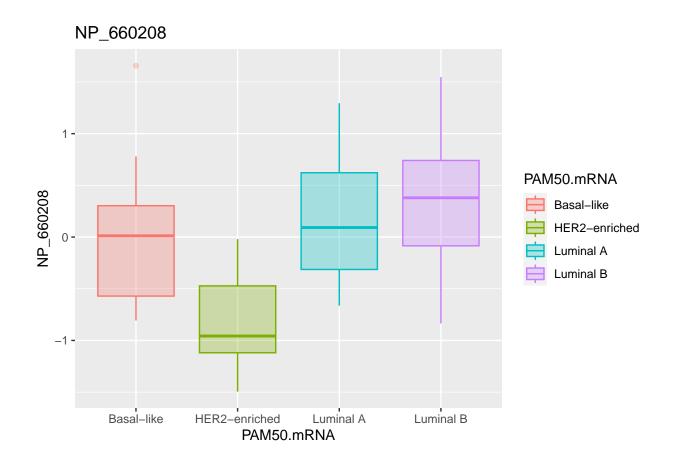
```
library(gridExtra)
```

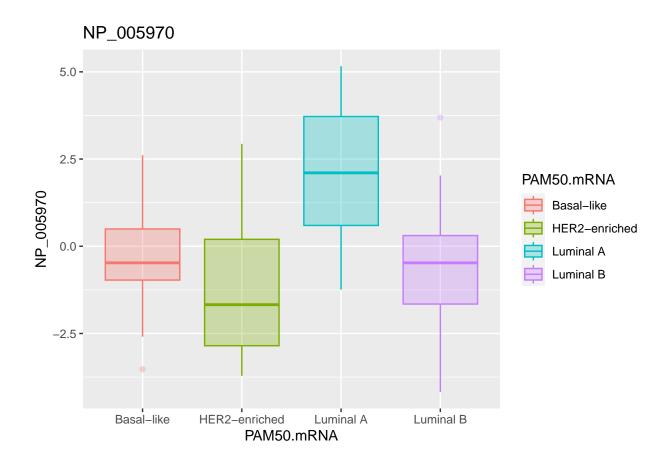
```
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
## combine
```

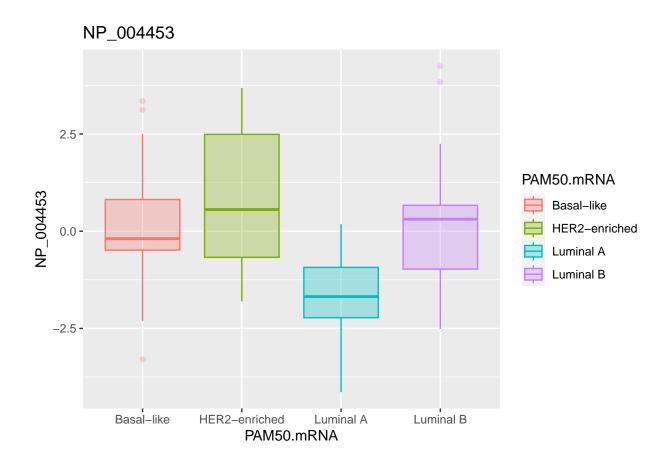
```
##
## combine

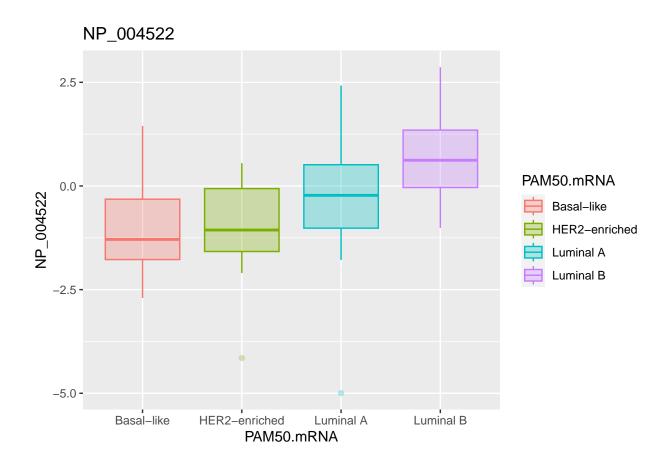
for (i in 1:length(STABVARS[1:5])){
print(ggplot(data, aes_string("PAM50.mRNA", STABVARS[i], col="PAM50.mRNA", fill="PAM50.mRNA")) + geom_b
```

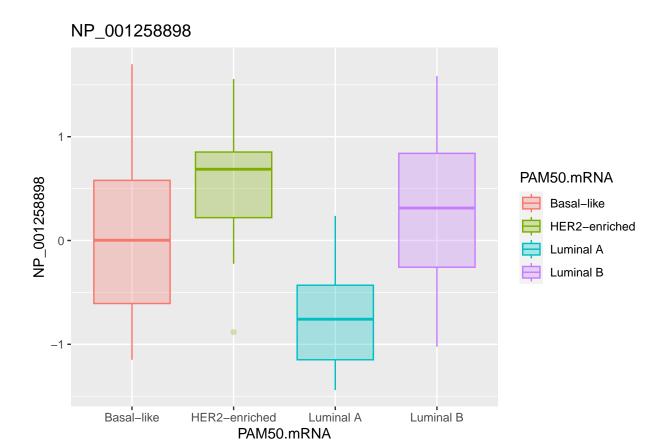
The following object is masked from 'package:randomForest':











This is promising. There are clear differences in the levels of each of the selected proteins.

Now that i have slected the proteins, I am going to create a subset of data with the selected features(predictor variables) and response variable only and stored as final_data

```
final_data <- data[,c(21, STABVARS.ind)]
head(final_data)</pre>
```

```
##
    PAM50.mRNA
               NP 038479 NP 001258898
                                        NP 542785 NP 002005 XP 003846537
## 1 Basal-like -0.3708959
                          -0.01456473
                                       0.05890562 -2.1305106
                                                               -0.4957955
## 2 Basal-like -1.6044748
                                       3.17273051 -0.9688522
                           -0.51053492
                                                               -0.6711132
## 3 Basal-like -3.0516640
                                       3.71857004 -0.3435704
                            0.53387720
                                                               -0.3498830
## 4 Basal-like -1.0663490
                            1.69892745
                                       2.96770134 -3.2821848
                                                               -0.4373841
## 5 Basal-like -0.3009396
                           -1.09705962
                                       0.03501174 -1.2946781
                                                               -0.3376402
  6 Basal-like 0.6787788
                           -0.70384668 -1.91679672 -2.7445377
##
                                                               -0.4103540
##
      NP_660208 NP_004439
                           NP_004388 NP_001229372 NP_005130 NP_004243 NP_003144
  1 -0.09170859 -4.419112
                           0.1691111
                                       -3.004808 2.2373013 -3.717470 0.1948258
## 2 -0.51388030 -5.187379
                           0.4863889
                                       -3.273820 -2.2066435 -2.905828 0.9748146
## 3
     1.65751511 -3.099008 -0.7349499
                                       -2.969601 -0.4698218 -3.076914 0.3539689
## 4 -0.70126018 -3.130366
                                       -2.859260 3.8894601 -3.180972 0.3397851
                           0.7843981
     0.78031558 -2.062567 -0.1823685
                                       -1.774609 -0.2388310 -1.785901 1.5453813
##
     0.41280109 -3.322351 -0.2659006
                                       -2.127745 -0.6167161 -1.937433 1.4835906
##
    NP_001171551 NP_072096
                              NP_005484
                                         NP_061170
                                                    NP_653304 NP_653081
## 1
       0.1948258 - 0.5215101 - 0.96600568 - 1.41417478 \ 0.5548304 - 0.2753844
## 2
       0.9547423 -0.5506795 -0.29643048 -2.19660738 -0.8751815 0.2890113
## 3
```

```
0.3470145 -1.2145533 -0.11205750 -1.51096203 0.1301301 0.3723177
## 5
       ## 6
       1.5432063 0.3623570 0.07115731 -0.06412445 -2.9967579 -0.3690816
     NP_057164 NP_004453
                           NP_003212
                                       NP_060866 NP_001160163
##
                                                              NP_003875
## 1 -0.6243686 -2.3105130 0.007476375 -1.399480707 1.340963113 0.04810062
## 2 0.6034772 -0.3499566 -0.376719645 -1.457277986 0.740637896 1.71079864
## 3 1.8279546 -3.2915417 -0.476134402 0.335031167 2.493930969 -3.12110229
## 4 -0.1771228 1.3374534 0.159048035 -0.451843095 1.055503657 -0.82416136
    1.4155177 -0.2021304 -1.819778494 -0.600190397 1.886978849 0.83677799
## 6  0.6375064  3.3454346  -2.334106568  0.002369978  0.002369978  -0.66257428
    NP_002677 NP_001135891 NP_004522 NP_003767 NP_848597 NP_005970
## 1 -7.684869
               0.9038146 -2.3839833 -0.3892635 0.3819222 0.8193237
## 2 -8.877335
               -1.2933543 -0.7514024 -1.0558322 -0.7346755 -0.9655069
## 3 -7.479934
              -0.4603530 -0.7412624 -1.0095468 -0.1952249 2.6107136
               -1.7784528 -0.7771697 -0.7446371 0.2277281 -0.2638766
## 4 -7.963274
## 5 -4.388818
                0.6306902 1.4493952 -0.5521973 1.9688493
                                                        1.8389858
## 6 -3.342988
               -1.0913487 1.2864003 0.2889839 -0.5502216 1.5019339
```

dim(final_data)

[1] 80 31

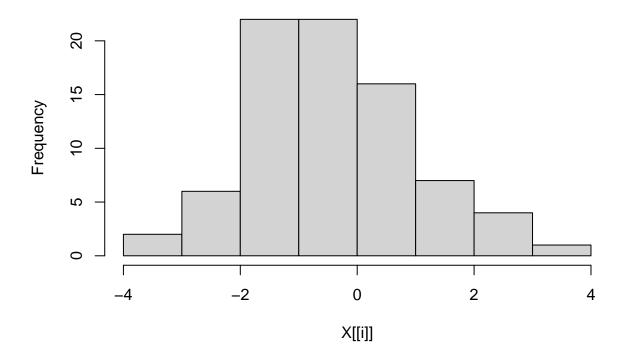
tail(final data)

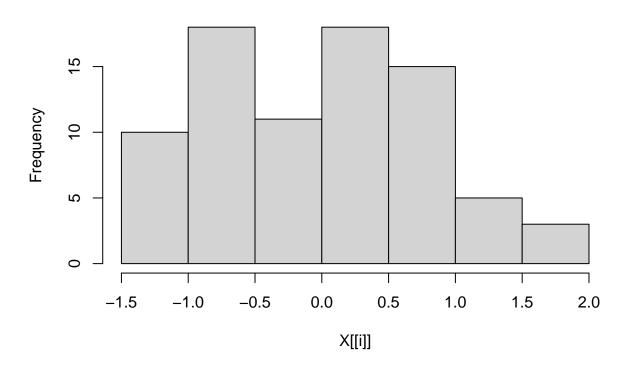
```
PAM50.mRNA NP_038479 NP_001258898 NP_542785 NP_002005 XP_003846537
##
## 75
     Luminal B -0.6318314
                          0.4327078 -5.5679315 0.8108994 -0.060341913
                          -0.7009215 -2.6553809 -0.8374660 0.624362546
## 76
      Luminal B -0.3073523
## 77
      Luminal B -2.5001368
                          -1.0230163 -2.3908031 -1.5362152 -0.001080874
      Luminal B -1.5815337
                          0.8397669 -4.2961120 -0.4483087 -1.344565342
                          0.3127390 -3.2926750 3.4240101 -0.115518269
## 79
      Luminal B -1.2355758
      Luminal B 0.4533356
                           1.4196209 0.3012351 -0.1103309 -0.696365039
                          NP_004388 NP_001229372 NP_005130 NP_004243
##
      NP_660208 NP_004439
     0.3346582 -2.0577536 0.22820424
## 75
                                     -1.704775 -5.100095 -1.6375408
                                     -1.388998 -4.925767 0.7448429
## 76
     1.0580919 -1.0382666 -0.41980067
                                     -2.292626 -2.281469 1.8062721
## 77
      0.5255059 -2.7790493
                          0.08370852
     0.7435718 0.9922713 0.34940658
                                       0.935962 -1.445453 -1.3633351
## 79 -0.1704231 -4.5628057 -0.77071534
                                       -5.997651 -2.066468 5.6714458
      0.4533356 -4.6822921 0.09097858
                                       -6.543286 -4.404932 1.9922345
##
       NP_003144 NP_001171551 NP_072096 NP_005484 NP_061170 NP_653304
## 75 -0.92317894 -0.92037752 0.1665730 0.6316086 -0.9427889 0.6456157
## 76 -0.11190639
                 ## 77 -0.11487717 -0.11487717 -0.8869070 0.7419420 3.6292442 -0.6236136
                  ## 78 -0.01660397
## 79 -0.30219453
                 -0.30219453 -0.6389439
                                      0.2688152 5.6897474 -0.6096613
                 0.15360819 -0.2803255 -0.6874180 -2.5439384 -0.9066216
     0.17597590
       NP_653081
                  NP_057164 NP_004453
                                       NP_003212 NP_060866 NP_001160163
## 75  0.53355890  -0.015519210  0.3150482  -3.45005881  0.2786298
                                                           1.29274343
      0.16921448 -1.466641235 -0.1841946 0.01392866 0.4958501
                                                            0.06316141
## 77 0.03908252 1.083330908 1.3198487 -0.25544907 0.4451791
                                                            0.72632291
## 78 -0.19726303 -0.664161107 0.8045736 0.03970534 0.9993099
                                                            0.90546108
## 79 -1.01961703 -0.569397806 0.4591518 -1.41493146 2.9262067
                                                            0.05285640
## 80 -0.24901071 -0.007439371 -2.1413195 -1.23766378 0.1088728
                                                           -0.69636504
       NP_003875 NP_002677 NP_001135891 NP_004522 NP_003767 NP_848597
##
```

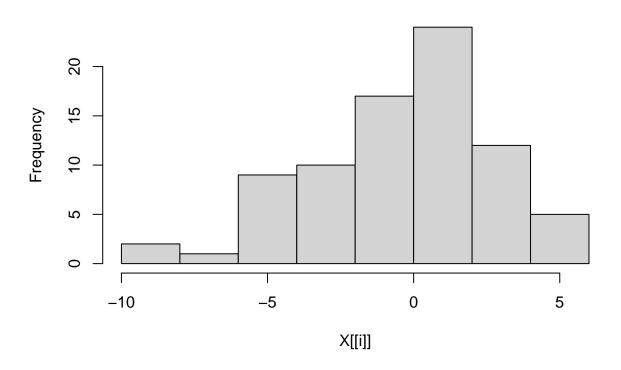
```
## 75 0.80809795 -4.1644206
                                0.9957930 1.9650840 1.4468215 0.9565732
## 76  0.04810062  -0.7973058  -1.0409439  0.2950495  2.6778835  3.2187065
## 77 -4.15353011
                  -5.7756852
                                0.3492332 1.6790880 0.4741860 -1.2930036
## 78 -4.17645469
                  -4.9108220
                               -0.7275091 0.8303820 1.1658917 0.2602502
## 79 -0.78535662 -10.4852016
                                0.2871168 -0.9720329 0.3127390 -0.7377725
      0.78885134
                  -9.2050444
                                1.5493537 1.0796316 0.6054361 -0.9513570
##
      NP_005970
## 75 -0.4749519
## 76 -1.2470992
      0.2711377
## 77
## 78
      1.2573943
## 79 -4.1821325
## 80 0.5696477
```

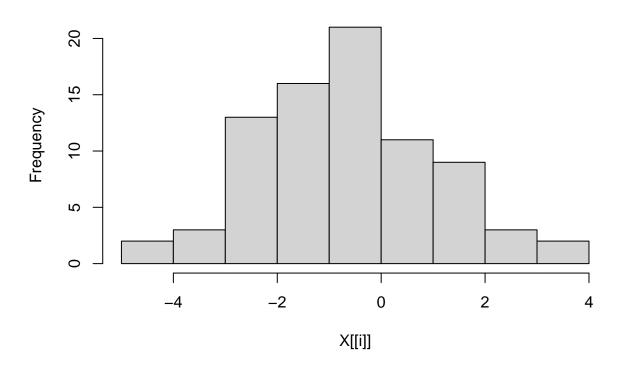
Checking the distribution of 30 proteins - built histogram using lapply()

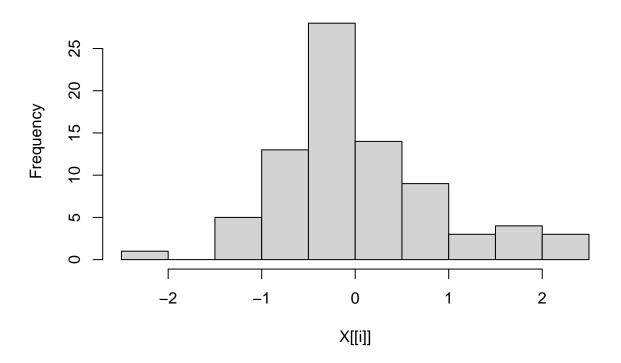
```
lapply(final_data[, 2:ncol(final_data)], hist)
```

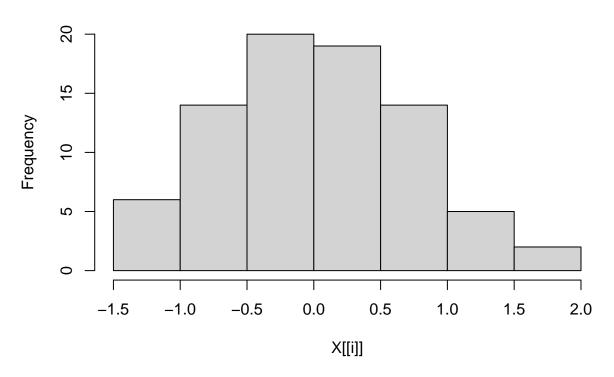


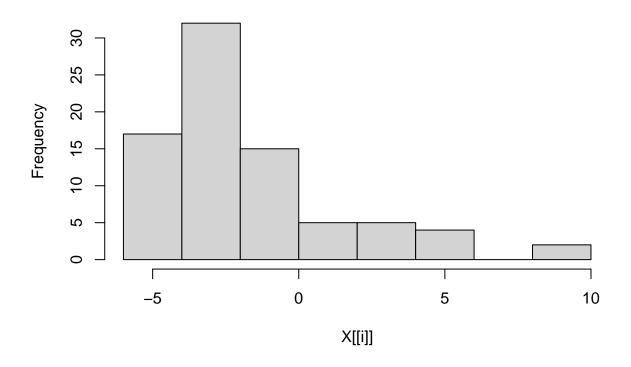


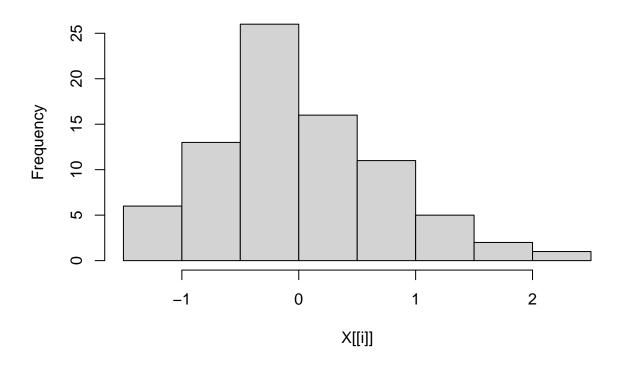


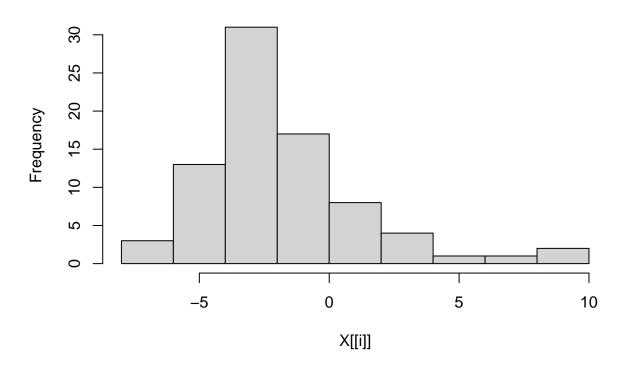


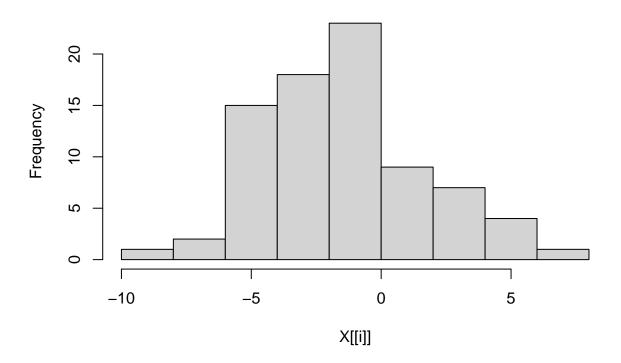


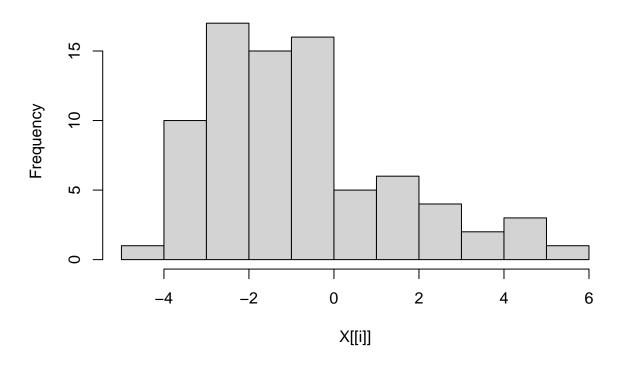


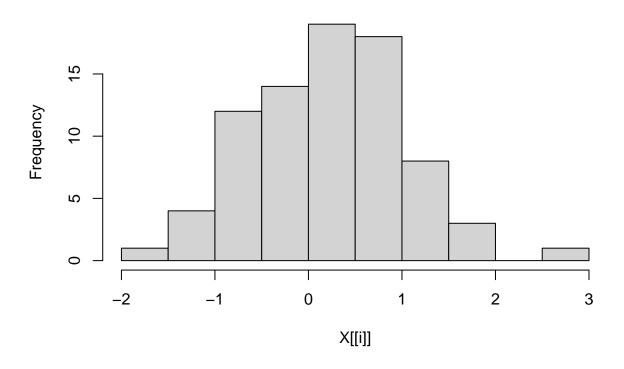


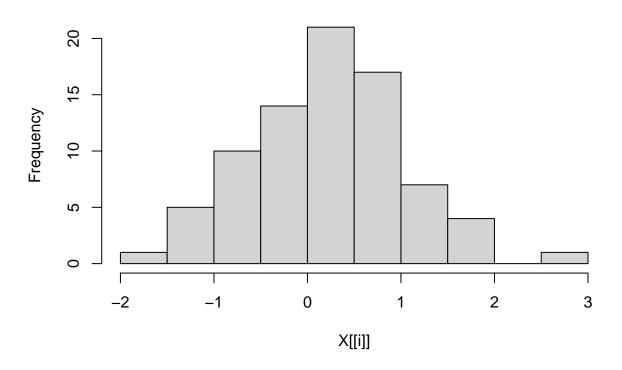


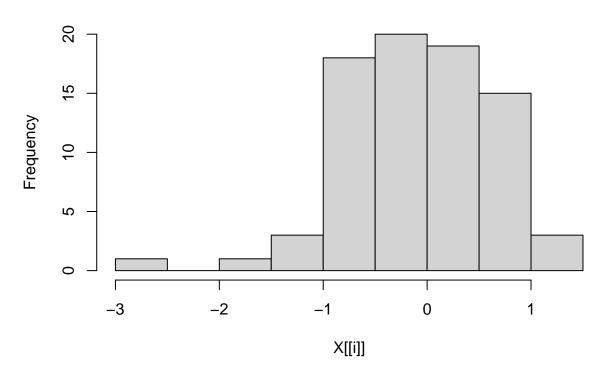


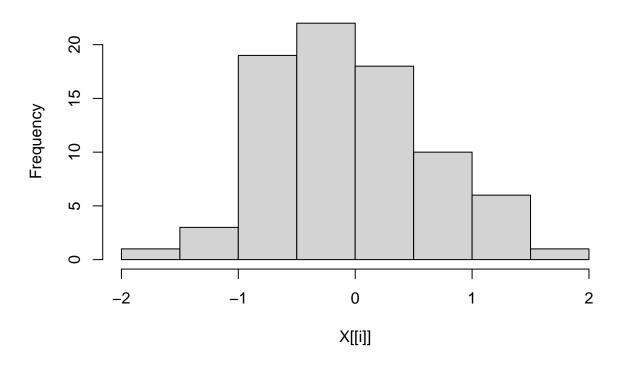


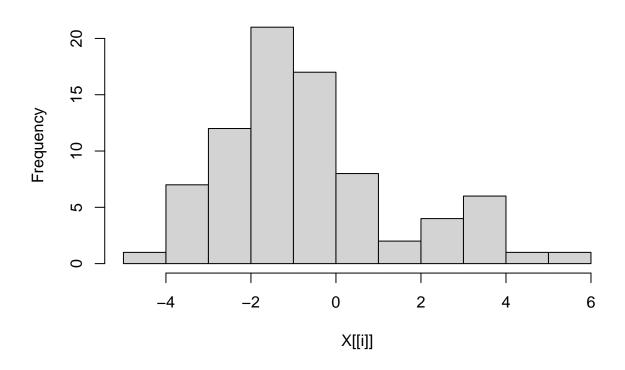


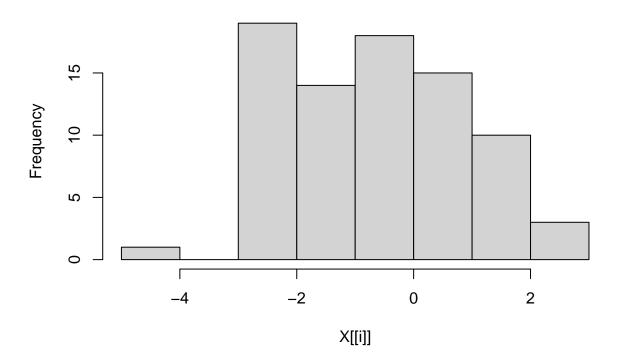


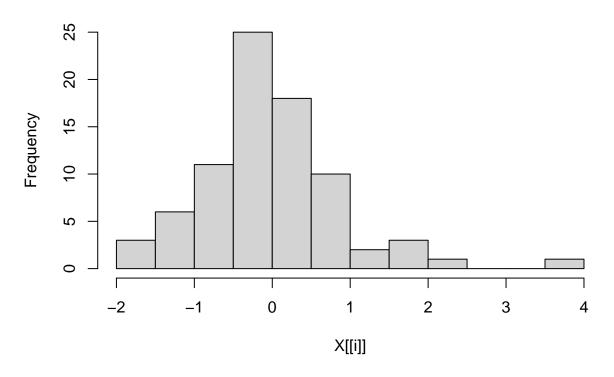


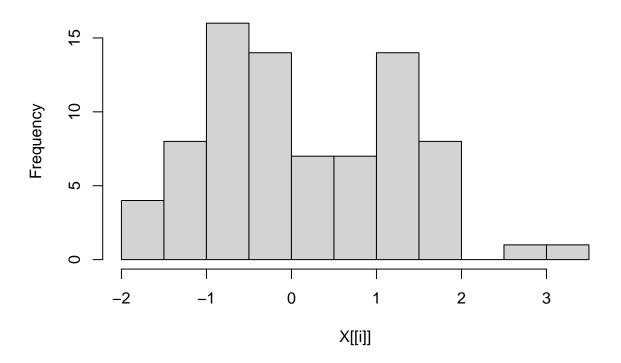


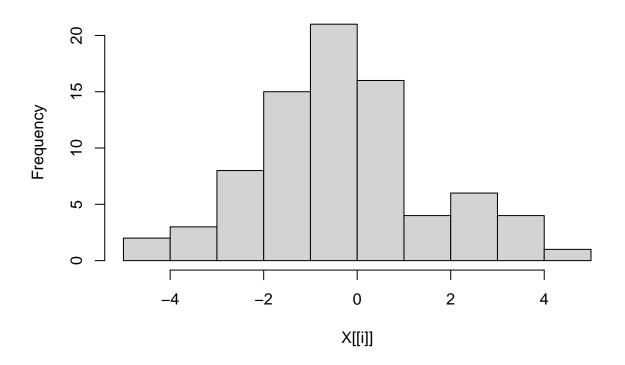


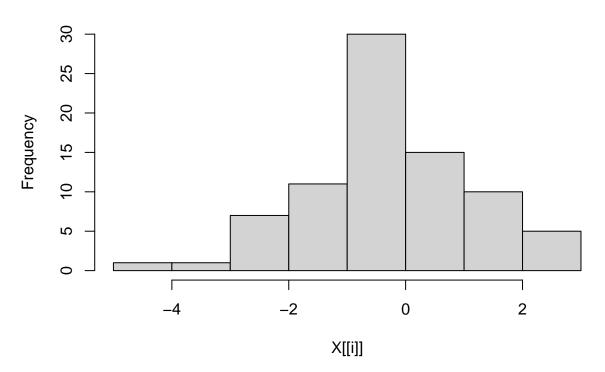


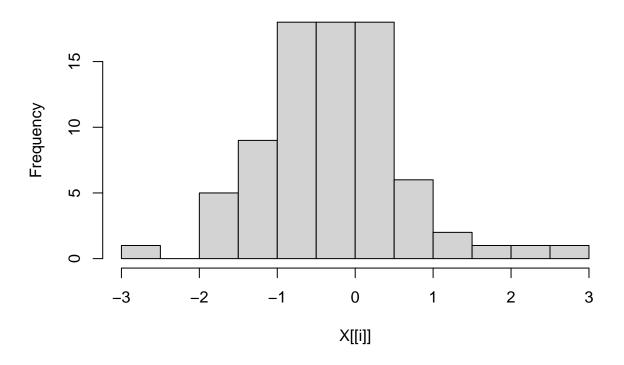


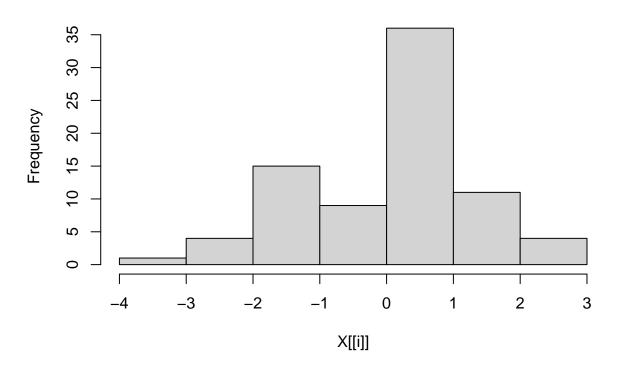


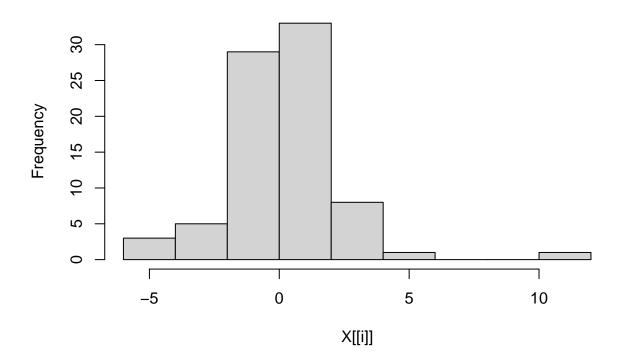


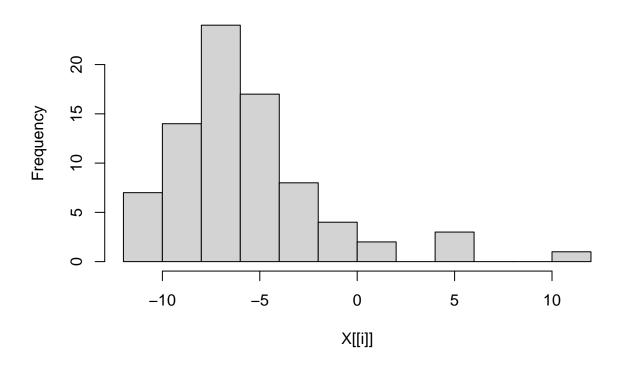


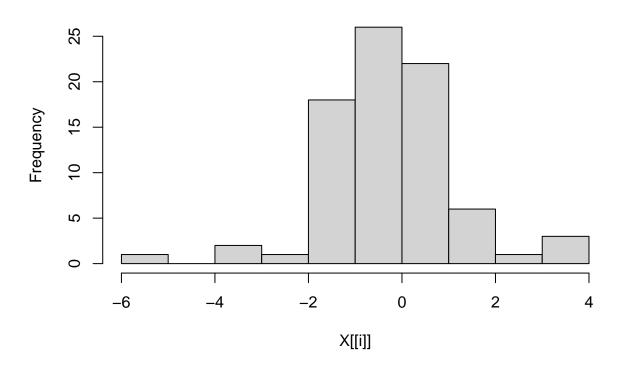


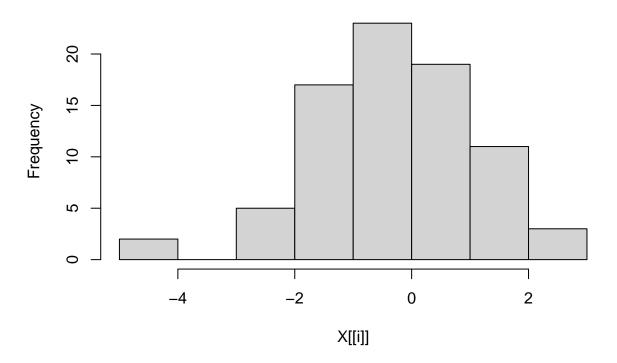


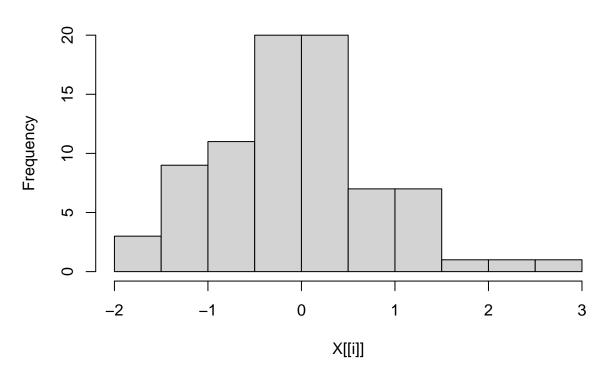


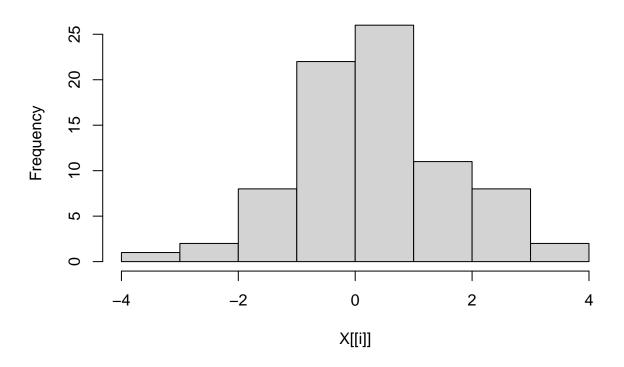


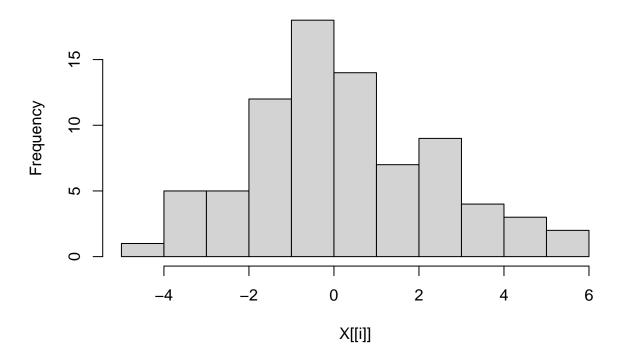












```
## $NP_038479
## $breaks
## [1] -4 -3 -2 -1 0 1 2 3 4
## $counts
## [1] 2 6 22 22 16 7 4 1
##
## $density
## [1] 0.0250 0.0750 0.2750 0.2750 0.2000 0.0875 0.0500 0.0125
## $mids
## [1] -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_001258898
## $breaks
## [1] -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0
##
```

```
## $counts
## [1] 10 18 11 18 15 5 3
## $density
## [1] 0.250 0.450 0.275 0.450 0.375 0.125 0.075
## [1] -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_542785
## $breaks
## [1] -10 -8 -6 -4 -2 0
                               2 4
                                      6
## $counts
## [1] 2 1 9 10 17 24 12 5
##
## $density
## [1] 0.01250 0.00625 0.05625 0.06250 0.10625 0.15000 0.07500 0.03125
## $mids
## [1] -9 -7 -5 -3 -1 1 3 5
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP 002005
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3 4
##
## $counts
## [1] 2 3 13 16 21 11 9 3 2
##
## $density
## [1] 0.0250 0.0375 0.1625 0.2000 0.2625 0.1375 0.1125 0.0375 0.0250
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5
##
## $xname
```

```
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
## $XP_003846537
## $breaks
## [1] -2.5 -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5
## $counts
## [1] 1 0 5 13 28 14 9 3 4 3
##
## $density
## [1] 0.025 0.000 0.125 0.325 0.700 0.350 0.225 0.075 0.100 0.075
##
## $mids
## [1] -2.25 -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_660208
## $breaks
## [1] -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0
## $counts
## [1] 6 14 20 19 14 5 2
##
## $density
## [1] 0.150 0.350 0.500 0.475 0.350 0.125 0.050
##
## $mids
## [1] -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
## $NP_004439
## $breaks
```

```
## [1] -6 -4 -2 0 2 4 6 8 10
##
## $counts
## [1] 17 32 15 5 5 4 0 2
## $density
## [1] 0.10625 0.20000 0.09375 0.03125 0.03125 0.02500 0.00000 0.01250
## $mids
## [1] -5 -3 -1 1 3 5 7 9
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_004388
## $breaks
## [1] -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5
## $counts
## [1] 6 13 26 16 11 5 2 1
## $density
## [1] 0.150 0.325 0.650 0.400 0.275 0.125 0.050 0.025
##
## $mids
## [1] -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_001229372
## $breaks
## [1] -8 -6 -4 -2 0 2 4 6 8 10
##
## $counts
## [1] 3 13 31 17 8 4 1 1 2
##
## $density
## [1] 0.01875 0.08125 0.19375 0.10625 0.05000 0.02500 0.00625 0.00625 0.01250
## $mids
## [1] -7 -5 -3 -1 1 3 5 7 9
```

```
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_005130
## $breaks
## [1] -10 -8 -6 -4 -2 0 2 4 6 8
##
## $counts
## [1] 1 2 15 18 23 9 7 4 1
##
## $density
## [1] 0.00625 0.01250 0.09375 0.11250 0.14375 0.05625 0.04375 0.02500 0.00625
## $mids
## [1] -9 -7 -5 -3 -1 1 3 5 7
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_004243
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3 4 5 6
## $counts
## [1] 1 10 17 15 16 5 6 4 2 3 1
##
## $density
## [1] 0.0125 0.1250 0.2125 0.1875 0.2000 0.0625 0.0750 0.0500 0.0250 0.0375
## [11] 0.0125
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5 4.5 5.5
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
```

```
##
## $NP_003144
## $breaks
  [1] -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5 3.0
## $counts
## [1] 1 4 12 14 19 18 8 3 0 1
##
## $density
## [1] 0.025 0.100 0.300 0.350 0.475 0.450 0.200 0.075 0.000 0.025
## $mids
## [1] -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25 2.75
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_001171551
## $breaks
  [1] -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5 3.0
##
## $counts
  [1] 1 5 10 14 21 17 7 4 0 1
##
##
## $density
  [1] 0.025 0.125 0.250 0.350 0.525 0.425 0.175 0.100 0.000 0.025
##
## $mids
   [1] -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25 2.75
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_072096
## $breaks
## [1] -3.0 -2.5 -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5
##
## $counts
## [1] 1 0 1 3 18 20 19 15 3
## $density
## [1] 0.025 0.000 0.025 0.075 0.450 0.500 0.475 0.375 0.075
```

```
##
## $mids
## [1] -2.75 -2.25 -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_005484
## $breaks
## [1] -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0
##
## $counts
## [1] 1 3 19 22 18 10 6 1
## $density
## [1] 0.025 0.075 0.475 0.550 0.450 0.250 0.150 0.025
##
## $mids
## [1] -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_061170
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3 4 5 6
##
## $counts
## [1] 1 7 12 21 17 8 2 4 6 1 1
##
## $density
## [1] 0.0125 0.0875 0.1500 0.2625 0.2125 0.1000 0.0250 0.0500 0.0750 0.0125
## [11] 0.0125
##
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5 4.5 5.5
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
```

```
##
## attr(,"class")
## [1] "histogram"
##
## $NP_653304
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3
##
## $counts
## [1] 1 0 19 14 18 15 10 3
## $density
## [1] 0.0125 0.0000 0.2375 0.1750 0.2250 0.1875 0.1250 0.0375
##
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_653081
## $breaks
## [1] -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5 3.0 3.5 4.0
##
## $counts
## [1] 3 6 11 25 18 10 2 3 1 0 0 1
##
## $density
## [1] 0.075 0.150 0.275 0.625 0.450 0.250 0.050 0.075 0.025 0.000 0.000 0.025
## $mids
## [1] -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25 2.75 3.25 3.75
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_057164
## $breaks
## [1] -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5 3.0 3.5
##
## $counts
## [1] 4 8 16 14 7 7 14 8 0 1 1
```

```
##
## $density
## [1] 0.100 0.200 0.400 0.350 0.175 0.175 0.350 0.200 0.000 0.025 0.025
##
## [1] -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25 2.75 3.25
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_004453
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3 4 5
##
## $counts
## [1] 2 3 8 15 21 16 4 6 4 1
##
## $density
## [1] 0.0250 0.0375 0.1000 0.1875 0.2625 0.2000 0.0500 0.0750 0.0500 0.0125
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5 4.5
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_003212
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3
##
## $counts
## [1] 1 1 7 11 30 15 10 5
## $density
## [1] 0.0125 0.0125 0.0875 0.1375 0.3750 0.1875 0.1250 0.0625
##
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5
##
## $xname
## [1] "X[[i]]"
##
```

```
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_060866
## $breaks
## [1] -3.0 -2.5 -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5 3.0
##
## $counts
## [1] 1 0 5 9 18 18 18 6 2 1 1 1
## $density
## [1] 0.025 0.000 0.125 0.225 0.450 0.450 0.450 0.150 0.050 0.025 0.025 0.025
##
## $mids
## [1] -2.75 -2.25 -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25 2.75
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_001160163
## $breaks
## [1] -4 -3 -2 -1 0 1 2 3
##
## $counts
## [1] 1 4 15 9 36 11 4
## $density
## [1] 0.0125 0.0500 0.1875 0.1125 0.4500 0.1375 0.0500
##
## $mids
## [1] -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
## $NP_003875
## $breaks
## [1] -6 -4 -2 0 2 4 6 8 10 12
##
```

```
## $counts
## [1] 3 5 29 33 8 1 0 0 1
## $density
## [1] 0.01875 0.03125 0.18125 0.20625 0.05000 0.00625 0.00000 0.00000 0.00625
## $mids
## [1] -5 -3 -1 1 3 5 7 9 11
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_002677
## $breaks
## [1] -12 -10 -8 -6 -4 -2 0 2 4 6 8 10 12
## $counts
## [1] 7 14 24 17 8 4 2 0 3 0 0 1
##
## $density
## [1] 0.04375 0.08750 0.15000 0.10625 0.05000 0.02500 0.01250 0.00000 0.01875
## [10] 0.00000 0.00000 0.00625
##
## $mids
## [1] -11 -9 -7 -5 -3 -1 1 3 5 7 9 11
##
## $xname
## [1] "X[[i]]"
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
## $NP_001135891
## $breaks
## [1] -6 -5 -4 -3 -2 -1 0 1 2 3 4
## $counts
## [1] 1 0 2 1 18 26 22 6 1 3
##
## $density
## [1] 0.0125 0.0000 0.0250 0.0125 0.2250 0.3250 0.2750 0.0750 0.0125 0.0375
##
## $mids
## [1] -5.5 -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5
##
```

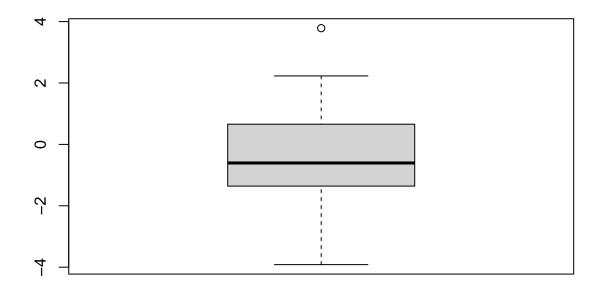
```
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_004522
## $breaks
## [1] -5 -4 -3 -2 -1 0 1 2 3
## $counts
## [1] 2 0 5 17 23 19 11 3
##
## $density
## [1] 0.0250 0.0000 0.0625 0.2125 0.2875 0.2375 0.1375 0.0375
## $mids
## [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_003767
## $breaks
## [1] -2.0 -1.5 -1.0 -0.5 0.0 0.5 1.0 1.5 2.0 2.5 3.0
## $counts
## [1] 3 9 11 20 20 7 7 1 1 1
##
## $density
## [1] 0.075 0.225 0.275 0.500 0.500 0.175 0.175 0.025 0.025 0.025
## $mids
## [1] -1.75 -1.25 -0.75 -0.25 0.25 0.75 1.25 1.75 2.25 2.75
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
##
## $NP_848597
```

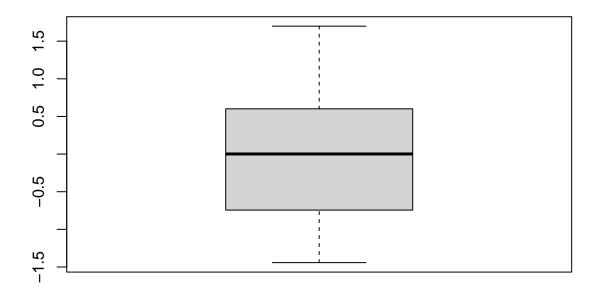
```
## $breaks
## [1] -4 -3 -2 -1 0 1 2 3 4
##
## $counts
## [1] 1 2 8 22 26 11 8 2
##
## $density
## [1] 0.0125 0.0250 0.1000 0.2750 0.3250 0.1375 0.1000 0.0250
##
## $mids
## [1] -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
##
## attr(,"class")
## [1] "histogram"
##
## $NP_005970
## $breaks
   [1] -5 -4 -3 -2 -1 0 1 2 3 4 5 6
##
##
## $counts
##
   [1] 1 5 5 12 18 14 7 9 4 3 2
##
## $density
   [1] 0.0125 0.0625 0.0625 0.1500 0.2250 0.1750 0.0875 0.1125 0.0500 0.0375
## [11] 0.0250
##
## $mids
   [1] -4.5 -3.5 -2.5 -1.5 -0.5 0.5 1.5 2.5 3.5 4.5 5.5
##
##
## $xname
## [1] "X[[i]]"
##
## $equidist
## [1] TRUE
## attr(,"class")
## [1] "histogram"
```

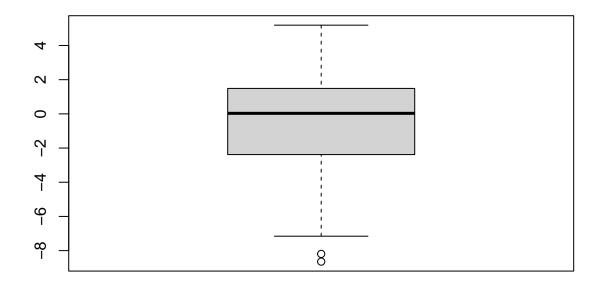
Detection of outliers and data imputation

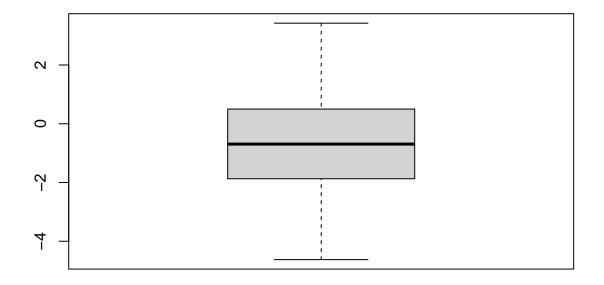
- On observing the box plot of 30 column, I got to know that it has a few outliers
- I have also created the function which can identify the presence of outliers in each variable
- I removed these outliers and imputed them with median value

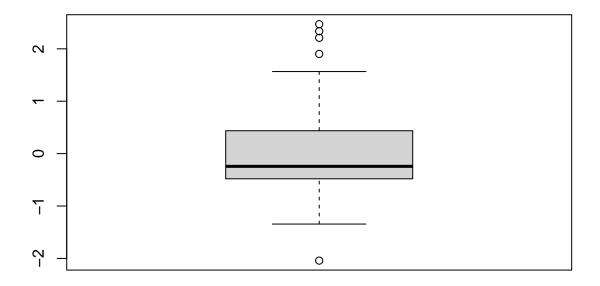
```
lapply(final_data[, 2:ncol(final_data)], boxplot)
```

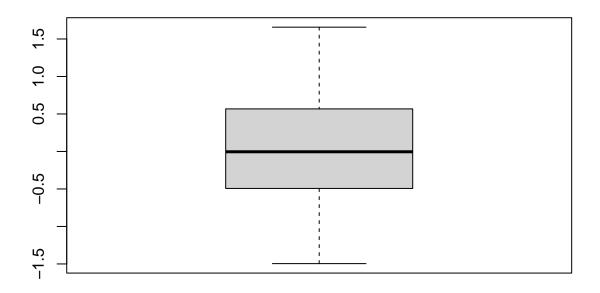


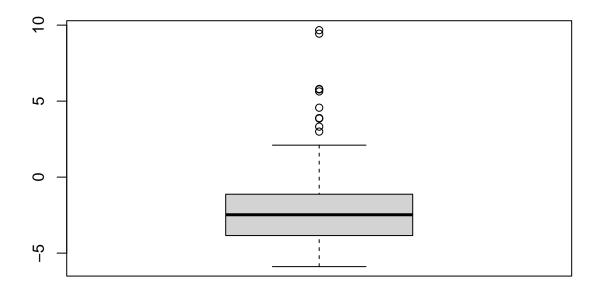


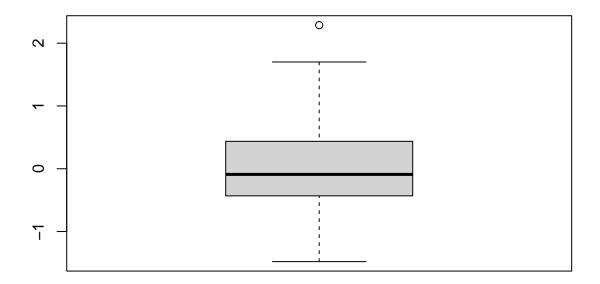


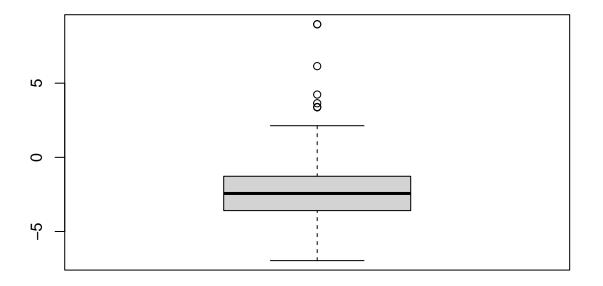


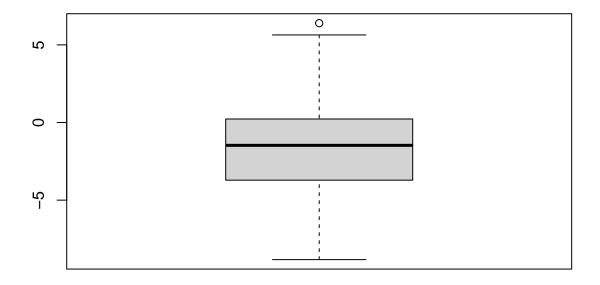


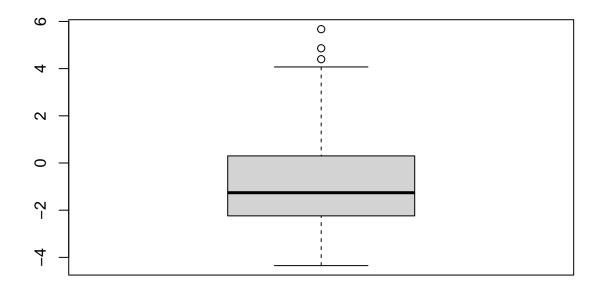


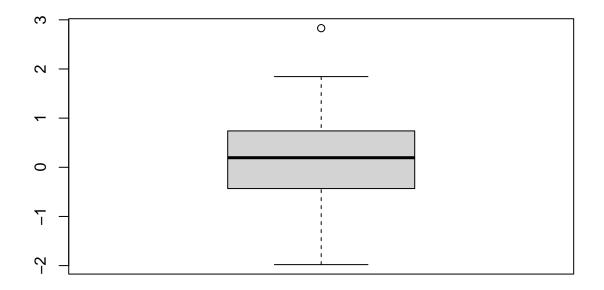


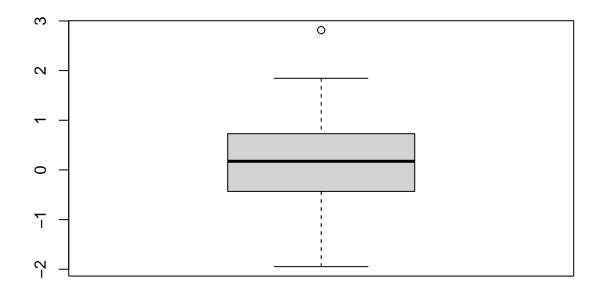


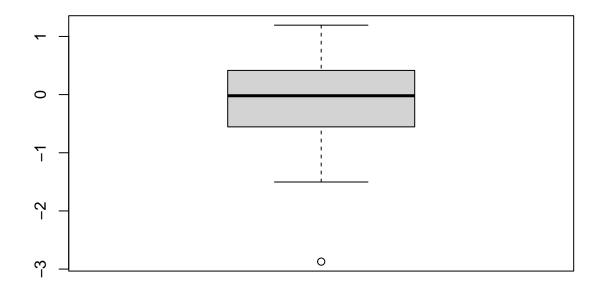


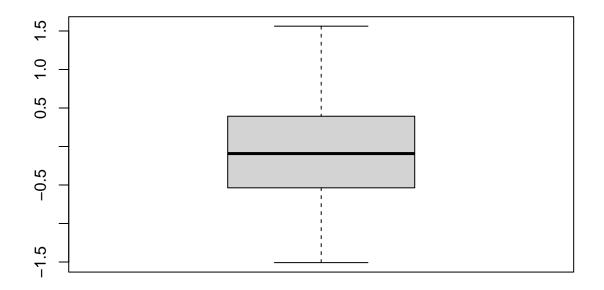


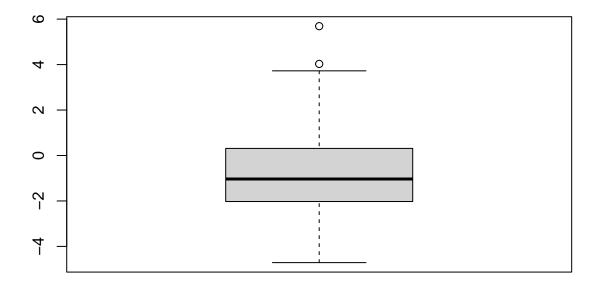


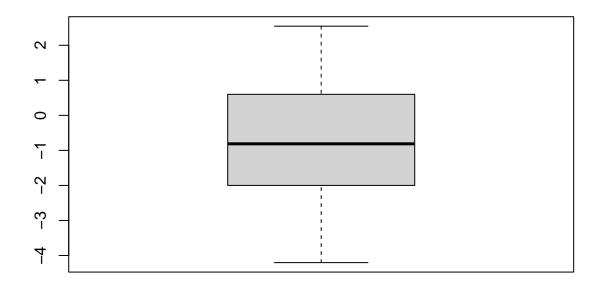


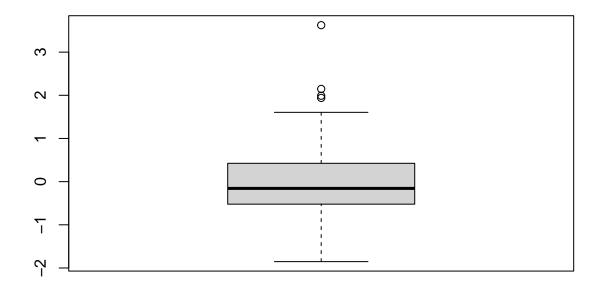


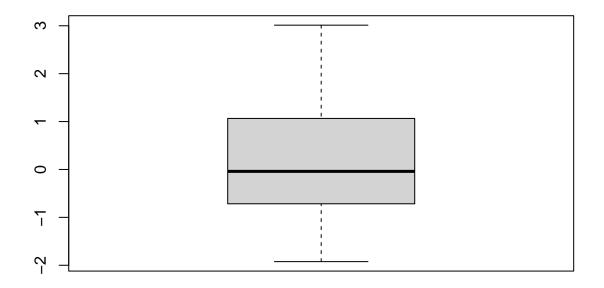


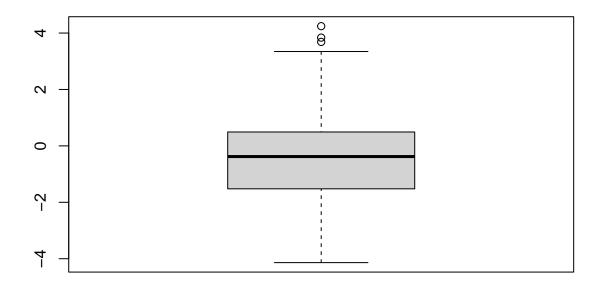


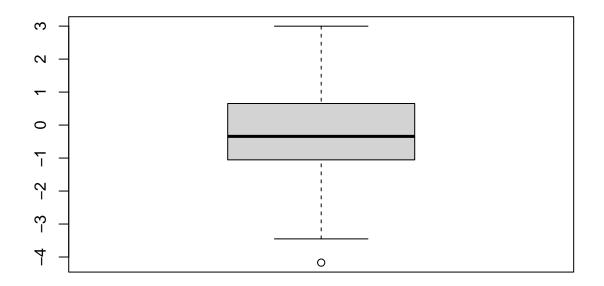


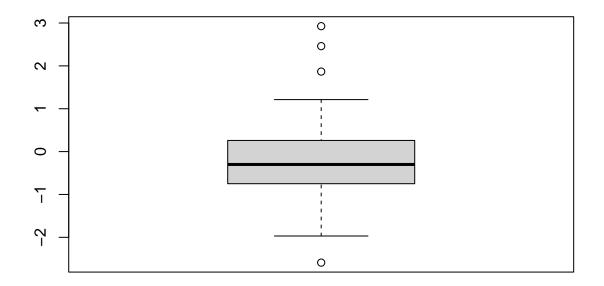


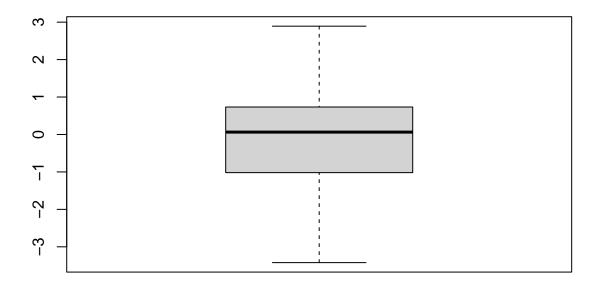


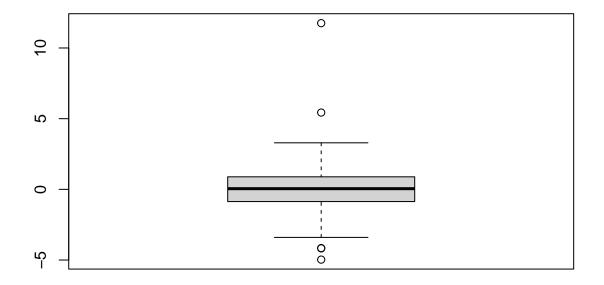


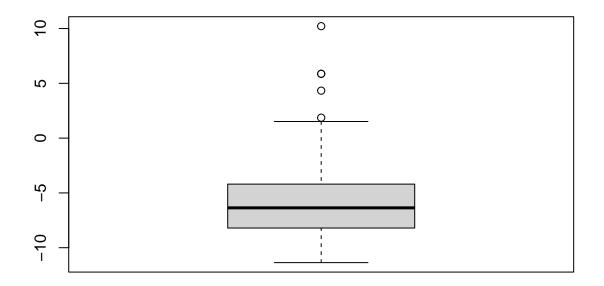


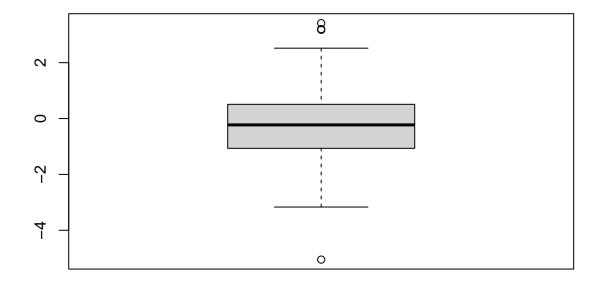


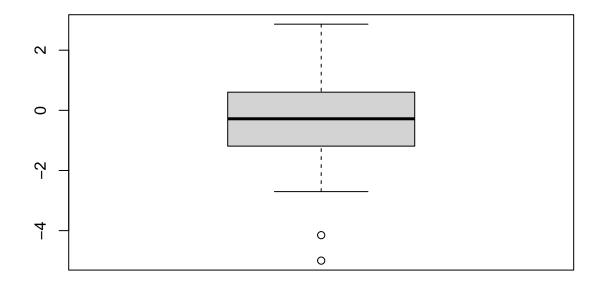


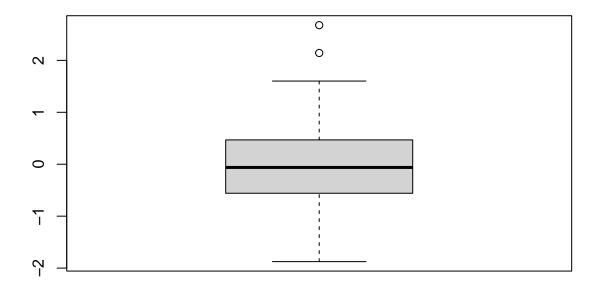


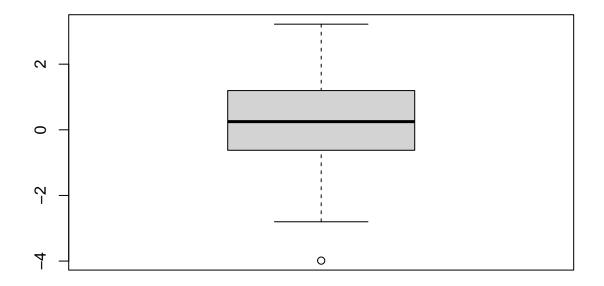


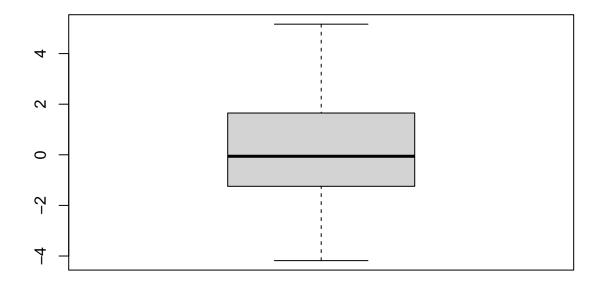












```
## $NP_038479
## $NP_038479$stats
##
              [,1]
## [1,] -3.9159043
## [2,] -1.3588005
## [3,] -0.6057669
## [4,] 0.6575331
## [5,] 2.2302073
##
## $NP_038479$n
## [1] 80
##
## $NP_038479$conf
##
              [,1]
## [1,] -0.9619510
## [2,] -0.2495829
##
## $NP_038479$out
## [1] 3.785229
##
## $NP_038479$group
## [1] 1
##
## $NP_038479$names
## [1] "1"
##
```

```
##
## $NP_001258898
## $NP_001258898$stats
               [,1]
## [1,] -1.44190438
## [2,] -0.74414488
## [3,] 0.00129172
## [4,] 0.60125977
## [5,] 1.69892745
## $NP_001258898$n
## [1] 80
## $NP_001258898$conf
              [,1]
## [1,] -0.2363732
## [2,] 0.2389566
## $NP_001258898$out
## numeric(0)
##
## $NP_001258898$group
## numeric(0)
## $NP_001258898$names
## [1] "1"
##
##
## $NP_542785
## $NP_542785$stats
##
## [1,] -7.15625291
## [2,] -2.38215725
## [3,] 0.03094425
## [4,] 1.48903278
## [5,] 5.18818807
## $NP_542785$n
## [1] 80
##
## $NP_542785$conf
              [,1]
## [1,] -0.6528990
## [2,] 0.7147875
## $NP_542785$out
## [1] -8.643742 -8.192980
## $NP_542785$group
## [1] 1 1
##
## $NP 542785$names
## [1] "1"
##
```

```
##
## $NP_002005
## $NP_002005$stats
              [,1]
## [1,] -4.6253957
## [2,] -1.8720616
## [3,] -0.6925908
## [4,] 0.5003105
## [5,] 3.4240101
## $NP_002005$n
## [1] 80
## $NP_002005$conf
              [,1]
## [1,] -1.1116688
## [2,] -0.2735127
## $NP_002005$out
## numeric(0)
##
## $NP_002005$group
## numeric(0)
## $NP_002005$names
## [1] "1"
##
## $XP_003846537
## $XP_003846537$stats
              [,1]
## [1,] -1.3445653
## [2,] -0.4809037
## [3,] -0.2429975
## [4,] 0.4367871
## [5,] 1.5662875
## $XP_003846537$n
## [1] 80
##
## $XP_003846537$conf
               [,1]
## [1,] -0.40510701
## [2,] -0.08088803
## $XP_003846537$out
## [1] 2.336572 1.902276 2.470569 2.210577 -2.042273
## $XP_003846537$group
## [1] 1 1 1 1 1
##
## $XP_003846537$names
## [1] "1"
##
```

```
##
## $NP_660208
## $NP_660208$stats
                [,1]
## [1,] -1.495746821
## [2,] -0.493137518
## [3,] -0.004159959
## [4,] 0.567662746
## [5,] 1.657515110
## $NP_660208$n
## [1] 80
## $NP_660208$conf
              [,1]
## [1,] -0.1915497
## [2,] 0.1832297
## $NP_660208$out
## numeric(0)
##
## $NP_660208$group
## numeric(0)
## $NP_660208$names
## [1] "1"
##
## $NP_004439
## $NP_004439$stats
##
## [1,] -5.887760
## [2,] -3.844681
## [3,] -2.475620
## [4,] -1.124023
## [5,] 2.103030
## $NP_004439$n
## [1] 80
##
## $NP_004439$conf
##
             [,1]
## [1,] -2.956223
## [2,] -1.995018
## $NP_004439$out
## [1] 9.668177 9.438237 2.995443 3.853287 5.635589 3.886441 3.322970 5.767942
## [9] 5.800973 4.567836
## $NP_004439$group
## [1] 1 1 1 1 1 1 1 1 1 1
## $NP_004439$names
## [1] "1"
```

```
##
##
## $NP_004388
## $NP_004388$stats
              [,1]
## [1,] -1.4793025
## [2,] -0.4315964
## [3,] -0.0896896
## [4,] 0.4370189
## [5,] 1.7005950
## $NP_004388$n
## [1] 80
##
## $NP_004388$conf
##
               [,1]
## [1,] -0.24312994
## [2,] 0.06375073
## $NP_004388$out
## [1] 2.286965
## $NP_004388$group
## [1] 1
##
## $NP_004388$names
## [1] "1"
##
##
## $NP_001229372
## $NP_001229372$stats
##
             [,1]
## [1,] -6.963267
## [2,] -3.590168
## [3,] -2.435468
## [4,] -1.275306
## [5,] 2.133122
##
## $NP_001229372$n
## [1] 80
## $NP_001229372$conf
             [,1]
## [1,] -2.844387
## [2,] -2.026549
##
## $NP_001229372$out
## [1] 8.978604 8.965470 6.148162 3.646817 3.369828 3.398403 4.232799
## $NP_001229372$group
## [1] 1 1 1 1 1 1 1
## $NP_001229372$names
## [1] "1"
```

```
##
##
## $NP_005130
## $NP_005130$stats
              [,1]
## [1,] -8.8307556
## [2,] -3.7147704
## [3,] -1.4701347
## [4,] 0.2255714
## [5,] 5.6405463
## $NP_005130$n
## [1] 80
##
## $NP_005130$conf
##
              [,1]
## [1,] -2.1661936
## [2,] -0.7740758
## $NP_005130$out
## [1] 6.397391
## $NP_005130$group
## [1] 1
##
## $NP_005130$names
## [1] "1"
##
##
## $NP_004243
## $NP_004243$stats
##
              [,1]
## [1,] -4.3493525
## [2,] -2.2404093
## [3,] -1.2588643
## [4,] 0.3011927
## [5,] 4.0708722
##
## $NP_004243$n
## [1] 80
## $NP_004243$conf
              [,1]
## [1,] -1.7078367
## [2,] -0.8098919
##
## $NP_004243$out
## [1] 4.399083 4.861847 5.671446
## $NP_004243$group
## [1] 1 1 1
## $NP_004243$names
## [1] "1"
```

```
##
##
## $NP_003144
## $NP_003144$stats
              [,1]
## [1,] -1.9798402
## [2,] -0.4325426
## [3,] 0.1947012
## [4,] 0.7404052
## [5,] 1.8449029
## $NP_003144$n
## [1] 80
##
## $NP_003144$conf
##
               [,1]
## [1,] -0.01249928
## [2,] 0.40190170
## $NP_003144$out
## [1] 2.830061
## $NP_003144$group
## [1] 1
##
## $NP_003144$names
## [1] "1"
##
##
## $NP_001171551
## $NP_001171551$stats
##
              [,1]
## [1,] -1.9470415
## [2,] -0.4322866
## [3,] 0.1742170
## [4,] 0.7294963
## [5,] 1.8449029
##
## $NP_001171551$n
## [1] 80
## $NP_001171551$conf
               [,1]
## [1,] -0.03101126
## [2,] 0.37944520
##
## $NP_001171551$out
## [1] 2.813958
## $NP_001171551$group
## [1] 1
## $NP_001171551$names
## [1] "1"
```

```
##
##
## $NP_072096
## $NP_072096$stats
               [,1]
## [1,] -1.50276929
## [2,] -0.55484927
## [3,] -0.01969573
## [4,] 0.41592377
## [5,] 1.19362341
## $NP_072096$n
## [1] 80
##
## $NP_072096$conf
##
              [,1]
## [1,] -0.1911822
## [2,] 0.1517907
## $NP_072096$out
## [1] -2.871865
## $NP_072096$group
## [1] 1
##
## $NP_072096$names
## [1] "1"
##
##
## $NP_005484
## $NP_005484$stats
##
               [,1]
## [1,] -1.50816552
## [2,] -0.53675374
## [3,] -0.09190035
## [4,] 0.39283208
## [5,] 1.56245999
##
## $NP_005484$n
## [1] 80
## $NP_005484$conf
               [,1]
## [1,] -0.25611110
## [2,] 0.07231039
##
## $NP_005484$out
## numeric(0)
## $NP_005484$group
## numeric(0)
## $NP_005484$names
## [1] "1"
```

```
##
##
## $NP_061170
## $NP_061170$stats
             [,1]
## [1,] -4.712975
## [2,] -2.025168
## [3,] -1.033795
## [4,] 0.311528
## [5,] 3.724644
## $NP_061170$n
## [1] 80
##
## $NP_061170$conf
##
              [,1]
## [1,] -1.4465707
## [2,] -0.6210191
## $NP_061170$out
## [1] 4.029373 5.689747
## $NP_061170$group
## [1] 1 1
##
## $NP_061170$names
## [1] "1"
##
##
## $NP_653304
## $NP_653304$stats
##
              [,1]
## [1,] -4.2030116
## [2,] -1.9997588
## [3,] -0.8129121
## [4,] 0.6002230
## [5,] 2.5447948
##
## $NP_653304$n
## [1] 80
## $NP_653304$conf
              [,1]
## [1,] -1.2721972
## [2,] -0.3536269
##
## $NP_653304$out
## numeric(0)
## $NP_653304$group
## numeric(0)
## $NP_653304$names
## [1] "1"
```

```
##
##
## $NP_653081
## $NP_653081$stats
              [,1]
## [1,] -1.8526870
## [2,] -0.5220216
## [3,] -0.1564469
## [4,] 0.4245559
## [5,] 1.6049629
## $NP_653081$n
## [1] 80
##
## $NP_653081$conf
##
               [,1]
## [1,] -0.32365917
## [2,] 0.01076545
## $NP_653081$out
## [1] 2.147486 1.989637 3.624280 1.938567
## $NP_653081$group
## [1] 1 1 1 1
##
## $NP_653081$names
## [1] "1"
##
##
## $NP_057164
## $NP_057164$stats
##
               [,1]
## [1,] -1.92397415
## [2,] -0.71653848
## [3,] -0.03968425
## [4,] 1.06566335
## [5,] 3.01239495
##
## $NP_057164$n
## [1] 80
## $NP_057164$conf
              [,1]
## [1,] -0.3545091
## [2,] 0.2751406
##
## $NP_057164$out
## numeric(0)
## $NP_057164$group
## numeric(0)
## $NP_057164$names
## [1] "1"
```

```
##
##
## $NP_004453
## $NP_004453$stats
              [,1]
## [1,] -4.1386672
## [2,] -1.5225765
## [3,] -0.3784592
## [4,] 0.4924532
## [5,] 3.3454346
## $NP_004453$n
## [1] 80
##
## $NP_004453$conf
##
               [,1]
## [1,] -0.73441287
## [2,] -0.02250543
## $NP_004453$out
## [1] 3.686225 3.837949 4.243675
## $NP_004453$group
## [1] 1 1 1
##
## $NP_004453$names
## [1] "1"
##
##
## $NP_003212
## $NP_003212$stats
##
              [,1]
## [1,] -3.4500588
## [2,] -1.0539582
## [3,] -0.3422749
## [4,] 0.6543596
## [5,] 2.9986201
##
## $NP_003212$n
## [1] 80
## $NP_003212$conf
               [,1]
## [1,] -0.64404820
## [2,] -0.04050166
##
## $NP_003212$out
## [1] -4.1699
## $NP_003212$group
## [1] 1
##
## $NP_003212$names
## [1] "1"
```

```
##
##
## $NP_060866
## $NP_060866$stats
              [,1]
## [1,] -1.9687931
## [2,] -0.7490088
## [3,] -0.3001057
## [4,] 0.2604499
## [5,] 1.2134275
## $NP_060866$n
## [1] 80
##
## $NP_060866$conf
##
              [,1]
## [1,] -0.4784260
## [2,] -0.1217855
## $NP_060866$out
## [1] -2.589974 2.460358 1.867033 2.926207
## $NP_060866$group
## [1] 1 1 1 1
##
## $NP_060866$names
## [1] "1"
##
##
## $NP_001160163
## $NP_001160163$stats
##
               [,1]
## [1,] -3.42210803
## [2,] -1.01917050
## [3,] 0.06316141
## [4,] 0.73348041
## [5,] 2.89217103
##
## $NP_001160163$n
## [1] 80
## $NP_001160163$conf
              [,1]
## [1,] -0.2464433
## [2,] 0.3727661
##
## $NP_001160163$out
## numeric(0)
## $NP_001160163$group
## numeric(0)
## $NP_001160163$names
## [1] "1"
```

```
##
##
## $NP_003875
## $NP_003875$stats
               [,1]
## [1,] -3.39803927
## [2,] -0.86429123
## [3,] 0.04810062
## [4,] 0.88773048
## [5,] 3.29279523
## $NP_003875$n
## [1] 80
##
## $NP_003875$conf
##
              [,1]
## [1,] -0.2613929
## [2,] 0.3575941
## $NP_003875$out
## [1] -4.970156 5.432549 11.755122 -4.153530 -4.176455
## $NP_003875$group
## [1] 1 1 1 1 1
##
## $NP_003875$names
## [1] "1"
##
##
## $NP_002677
## $NP_002677$stats
##
              [,1]
## [1,] -11.361115
## [2,] -8.205643
## [3,] -6.362650
## [4,] -4.197724
## [5,]
        1.516256
##
## $NP_002677$n
## [1] 80
## $NP_002677$conf
            [,1]
## [1,] -7.070647
## [2,] -5.654654
##
## $NP_002677$out
## [1] 4.331248 1.864520 5.868250 10.215922 5.871759
## $NP_002677$group
## [1] 1 1 1 1 1
## $NP_002677$names
## [1] "1"
```

```
##
##
## $NP_001135891
## $NP_001135891$stats
              [,1]
## [1,] -3.1680635
## [2,] -1.0661463
## [3,] -0.2271340
## [4,] 0.5092675
## [5,] 2.5193274
## $NP_001135891$n
## [1] 80
## $NP_001135891$conf
##
               [,1]
## [1,] -0.50542981
## [2,] 0.05116191
## $NP_001135891$out
## [1] -5.050298 3.201698 3.175491 3.414158
## $NP_001135891$group
## [1] 1 1 1 1
##
## $NP_001135891$names
## [1] "1"
##
##
## $NP_004522
## $NP_004522$stats
##
              [,1]
## [1,] -2.7021650
## [2,] -1.1902918
## [3,] -0.2807051
## [4,] 0.6040038
## [5,] 2.8653784
##
## $NP_004522$n
## [1] 80
## $NP_004522$conf
               [,1]
## [1,] -0.59766633
## [2,] 0.03625606
##
## $NP_004522$out
## [1] -4.150392 -4.996855
## $NP_004522$group
## [1] 1 1
## $NP_004522$names
## [1] "1"
```

```
##
##
## $NP_003767
## $NP_003767$stats
               [,1]
## [1,] -1.87517546
## [2,] -0.55882109
## [3,] -0.06146679
## [4,] 0.46892980
## [5,] 1.60169464
## $NP_003767$n
## [1] 80
##
## $NP_003767$conf
##
              [,1]
## [1,] -0.2430183
## [2,] 0.1200848
## $NP_003767$out
## [1] 2.142263 2.677884
## $NP_003767$group
## [1] 1 1
##
## $NP_003767$names
## [1] "1"
##
##
## $NP_848597
## $NP_848597$stats
##
              [,1]
## [1,] -2.8023343
## [2,] -0.6238947
## [3,] 0.2488007
## [4,] 1.1953659
## [5,] 3.2187065
##
## $NP_848597$n
## [1] 80
## $NP_848597$conf
               [,1]
## [1,] -0.07257056
## [2,] 0.57017190
##
## $NP_848597$out
## [1] -3.986077
## $NP_848597$group
## [1] 1
##
## $NP_848597$names
## [1] "1"
```

```
##
##
## $NP 005970
## $NP_005970$stats
                [,1]
## [1,] -4.18213251
## [2,] -1.24546089
## [3,] -0.05750858
## [4,] 1.65100594
## [5,] 5.16274745
## $NP_005970$n
## [1] 80
##
## $NP_005970$conf
##
               [,1]
## [1,] -0.5691676
## [2,] 0.4541505
## $NP_005970$out
## numeric(0)
## $NP_005970$group
## numeric(0)
##
## $NP_005970$names
## [1] "1"
# function for detection of outliers in each column
outliers <- function(x)</pre>
{
  for(i in 1:ncol(x))
    sd_i \leftarrow sd(x[,i])
    mean_i <- mean(x[,i])</pre>
    out = x[x[,i] > 3*sd_i+mean_i | x[,i] < mean_i-3*sd_i, ]
    if(nrow(out) > 0)
    {
      print(colnames(x)[i])
      paste("The outliers are -", out)
    }else
      print(paste("No outliers for",colnames(x)[i]))
    }
  }
# Detecting outliers in the dataset
outliers(final_data[,c(2:ncol(final_data))])
## [1] "NP_038479"
## [1] "No outliers for NP_001258898"
```

```
## [1] "No outliers for NP_542785"
## [1] "No outliers for NP_002005"
## [1] "No outliers for XP_003846537"
## [1] "No outliers for NP_660208"
## [1] "NP_004439"
## [1] "No outliers for NP 004388"
## [1] "NP_001229372"
## [1] "No outliers for NP_005130"
## [1] "No outliers for NP_004243"
## [1] "NP_003144"
## [1] "NP_001171551"
## [1] "NP_072096"
## [1] "No outliers for NP_005484"
## [1] "No outliers for NP_061170"
## [1] "No outliers for NP_653304"
## [1] "NP_653081"
## [1] "No outliers for NP_057164"
## [1] "No outliers for NP 004453"
## [1] "No outliers for NP_003212"
## [1] "NP 060866"
## [1] "No outliers for NP_001160163"
## [1] "NP 003875"
## [1] "NP_002677"
## [1] "NP 001135891"
## [1] "NP 004522"
## [1] "NP 003767"
## [1] "NP_848597"
## [1] "No outliers for NP_005970"
# replacing outliers with median imputation
outlier <- function(x) {</pre>
x[x < quantile(x,0.25) - 1.5 * IQR(x) | x > quantile(x,0.75) + 1.5 * IQR(x)] <- median(x)
х
}
data_out <- as.data.frame(lapply(final_data[,c(2:ncol(final_data))], outlier))</pre>
data_norm <- data_out</pre>
data_norm$PAM50.mRNA <- final_data$PAM50.mRNA</pre>
```

Feature Engineering -

I have not derived new features. - converted the PAM50.mRNA as a factor type as this is a response variable. for this I have used factor()

```
# changing the names
data_norm$PAM50.mRNA[which(data_norm$PAM50.mRNA == "Basal-like")] = "Basal.like"
data_norm$PAM50.mRNA[which(data_norm$PAM50.mRNA == "HER2-enriched")] = "HER2.enriched"
data_norm$PAM50.mRNA[which(data_norm$PAM50.mRNA == "Luminal A")] = "Luminal.A"
data_norm$PAM50.mRNA[which(data_norm$PAM50.mRNA == "Luminal B")] = "Luminal.B"

# converting PAM50.mRNA to a factor type
data_norm$PAM50.mRNA <- factor(data_norm$PAM50.mRNA)</pre>
```

Correlation/Collinearity analysis

- Numerical data is required for calculating correlation, so I have used only numerical variables to interpret the correlation
- Correlation plot is shown for whole data
- I have shown the plot of correlation between numeric features
- I also tried pairs.panels function for correlation but since there are more than 15 features. Plots are not clearly visible pairs.panels(data_n)
- Thefore, build a plot with top ten features
- I cannot apply chi-square test as the datapoints should be non-zero and non-negative

```
#Creating a correlation plot of whole dataset
cormat <- round(cor(data_norm[,1:30]),2)
cormat</pre>
```

##		_	_	_	_	XP_003846537	_
	NP_038479	1.00	-0.41	0.20	-0.14	0.33	-0.05
##	NP_001258898	-0.41	1.00	-0.04	0.33	-0.30	-0.13
##	NP_542785	0.20	-0.04	1.00	-0.34	0.07	-0.35
	NP_002005	-0.14	0.33	-0.34	1.00	0.09	0.18
	XP_003846537	0.33	-0.30	0.07	0.09	1.00	0.06
	NP_660208	-0.05	-0.13	-0.35	0.18	0.06	1.00
	NP_004439	0.13	0.05	0.07	0.06	-0.05	-0.03
	NP_004388	-0.45	0.56	0.07	0.03	-0.22	-0.10
##	NP_001229372	0.09	0.06	0.13	0.00	0.00	-0.07
##	NP_005130	0.05	-0.04	0.42	-0.49	-0.09	-0.18
##	NP_004243	0.01	0.12	-0.39	0.64	0.17	0.24
##	NP_003144	0.36	-0.51	0.16	-0.24	0.19	0.13
##	NP_001171551	0.35	-0.53	0.16	-0.25	0.19	0.14
##	NP_072096	0.12	-0.16	-0.29	0.07	-0.04	0.52
##	NP_005484	-0.36	0.34	-0.15	0.02	-0.18	-0.08
##	NP_061170	-0.01	-0.04	0.06	0.01	0.17	-0.08
##	NP_653304	-0.25	0.53	0.09	0.17	-0.05	-0.21
##	NP_653081	0.05	0.18	0.15	0.09	0.02	-0.28
##	NP_057164	0.11	-0.51	-0.01	0.06	0.26	0.35
##	NP_004453	-0.13	0.32	0.07	0.07	-0.19	-0.25
##	NP_003212	-0.22	0.23	0.23	-0.08	-0.05	-0.38
##	NP_060866	-0.02	0.18	-0.32	0.32	-0.01	0.28
##	NP_001160163	-0.40	0.46	-0.03	-0.11	-0.37	-0.05
##	NP_003875	0.34	-0.30	-0.18	-0.01	0.15	0.09
##	NP_002677	-0.04	0.09	-0.03	0.02	-0.14	0.13
##	NP_001135891	-0.03	0.18	-0.17	0.34	0.04	-0.04
##	NP_004522	0.00	-0.08	-0.33	0.27	0.08	0.26
##	NP_003767	-0.19	0.06	-0.30	0.36	-0.11	0.25
##	NP_848597	0.38	-0.36	-0.03	0.18	0.26	0.23
##	NP_005970	0.27	-0.41	0.10	-0.15	0.21	0.28
##		NP_004439	NP_004388 NP	_001229372	NP_005130	NP_004243 NP_	_003144
##	NP_038479	0.13	-0.45	0.09	0.05	0.01	0.36
##	NP_001258898	0.05	0.56	0.06	-0.04	0.12	-0.51
##	NP_542785	0.07	0.07	0.13	0.42	-0.39	0.16
##	NP_002005	0.06	0.03	0.00	-0.49	0.64	-0.24
##	XP_003846537	-0.05	-0.22	0.00	-0.09	0.17	0.19
##	NP_660208	-0.03	-0.10	-0.07	-0.18	0.24	0.13

##	NP_004439	1.00	0.06	0.78	0.01	0.05	-0.04
##	NP_004388	0.06	1.00	0.19	0.16	-0.09	-0.48
##	NP_001229372	0.78	0.19	1.00	0.09	-0.04	-0.04
##	NP_005130	0.01	0.16	0.09	1.00	-0.49	0.05
##	NP_004243	0.05	-0.09	-0.04	-0.49	1.00	-0.04
##	NP_003144	-0.04	-0.48	-0.04	0.05	-0.04	1.00
##	NP_001171551	-0.05	-0.48	-0.05	0.05	-0.04	1.00
##	NP_072096	-0.01	-0.17	-0.10	-0.10	0.15	0.17
##	NP_005484	0.13	0.30	0.17	0.02	0.02	-0.37
##	NP_061170	0.06	-0.02	0.13	-0.17	0.13	-0.10
##	NP_653304	0.15	0.67	0.22	0.02	0.10	-0.58
##	NP_653081	0.05	0.25	0.16	0.22	0.14	-0.23
##	NP_057164	-0.13	-0.36	-0.17	-0.01	0.15	0.49
##	NP_004453	0.17	0.22	0.29	-0.05	0.06	-0.28
##	NP_003212	0.10	0.28	0.13	0.20	-0.15	-0.17
##	NP_060866	0.03	0.12	0.07	-0.28	0.37	-0.17
##	NP_001160163	0.01	0.38	0.07	0.11	-0.20	-0.33
##	NP_003875	-0.13	-0.21	-0.13	-0.13	0.11	0.23
##	NP_002677	0.38	0.23	0.41	0.12	0.03	-0.02
##	NP_001135891	-0.13	0.00	-0.14	-0.16	0.27	-0.16
##	NP_004522	0.01	-0.16	-0.12	-0.21	0.38	0.14
##	NP_003767	0.23	0.07	0.15	-0.27	0.41	-0.24
	NP_848597	0.10	-0.26	0.13	-0.05	0.23	0.20
##	NP_005970	-0.16	-0.20	-0.10	0.14	0.00	0.47
##	_	NP_001171551		NP_005484	NP_061170	NP_653304	NP_653081
##	NP 038479	0.35	0.12	-0.36	-0.01	-0.25	0.05
##	NP_001258898	-0.53	-0.16	0.34	-0.04	0.53	0.18
	NP_542785	0.16	-0.29	-0.15	0.06	0.09	0.15
	NP_002005	-0.25	0.07	0.02	0.01	0.17	0.09
	XP_003846537	0.19	-0.04	-0.18	0.17	-0.05	0.02
	NP_660208	0.14	0.52	-0.08	-0.08	-0.21	-0.28
	NP_004439	-0.05	-0.01	0.13	0.06	0.15	0.05
	NP_004388	-0.48	-0.17	0.30	-0.02	0.67	0.25
	NP_001229372	-0.05	-0.10	0.17	0.13	0.22	0.16
	NP_005130	0.05	-0.10	0.02	-0.17	0.02	0.22
##	NP_004243	-0.04	0.15	0.02	0.13	0.10	0.14
##	NP_003144	1.00	0.17	-0.37	-0.10	-0.58	-0.23
##	NP_001171551	1.00	0.18	-0.37	-0.08	-0.59	-0.23
	NP_072096	0.18	1.00	-0.14	-0.18	-0.29	-0.19
	NP_005484	-0.37	-0.14	1.00	0.18	0.26	0.04
	NP 061170	-0.08	-0.18	0.18	1.00	0.13	0.18
##	NP_653304	-0.59	-0.29	0.26	0.13	1.00	0.40
##	NP_653081	-0.23	-0.19	0.04	0.18	0.40	1.00
	NP_057164	0.50	0.31	-0.33	-0.13	-0.50	-0.19
	NP_004453	-0.29	-0.25	0.36	0.25	0.24	0.22
	NP_003212	-0.18	-0.32	0.24	-0.01	0.23	0.16
	NP 060866	-0.18	0.11	0.14	0.14	0.16	0.04
##	NP_001160163	-0.32	0.03	0.33	-0.02	0.30	0.14
	NP_003875	0.22	0.06	-0.30	-0.15	-0.18	-0.08
	NP_002677	0.00	0.10	0.23	0.04	0.06	0.18
	NP_001135891	-0.16	-0.03	-0.06	0.01	0.13	0.14
	NP_004522	0.15	0.13	0.18	0.14	-0.23	-0.08
	NP_003767	-0.21	0.07	-0.08	0.14	0.18	0.09
			0.01	0.00		0.10	0.03
##	NP_848597	0.22	0.27	-0.28	-0.15	-0.23	0.09

##	NP_005970	0	. 48	0	. 28	-0.	22	-0.	16	-0.43	-0.11
##	NF_003970									0.43	
	NP_038479	0.11	*** -	-0.13	*** -	-0.22	*** -	-0.02	*** -	-0.40	0.34
	NP_001258898	-0.51		0.32		0.23		0.18		0.46	-0.30
	NP_542785	-0.01		0.07		0.23		-0.32		-0.03	-0.18
	NP_002005	0.06		0.07		-0.08		0.32		-0.11	-0.01
	XP_003846537	0.26		-0.19		-0.05		-0.01		-0.37	0.15
	NP_660208	0.35		-0.25		-0.38		0.28		-0.05	0.09
	NP_004439	-0.13		0.17		0.10		0.03		0.01	-0.13
##	NP_004388	-0.36		0.22		0.28		0.12		0.38	-0.21
##	NP_001229372	-0.17		0.29		0.13		0.07		0.07	-0.13
##	NP_005130	-0.01		-0.05		0.20		-0.28		0.11	-0.13
##	NP_004243	0.15		0.06		-0.15		0.37		-0.20	0.11
##	NP_003144	0.49		-0.28		-0.17		-0.17		-0.33	0.23
##	NP_001171551	0.50		-0.29		-0.18		-0.18		-0.32	0.22
##	NP_072096	0.31		-0.25		-0.32		0.11		0.03	0.06
##	NP_005484	-0.33		0.36		0.24		0.14		0.33	-0.30
##	NP_061170	-0.13		0.25		-0.01		0.14		-0.02	-0.15
##	NP_653304	-0.50		0.24		0.23		0.16		0.30	-0.18
	NP_653081	-0.19		0.22		0.16		0.04		0.14	-0.08
##	NP_057164	1.00		-0.47		-0.24		-0.16		-0.37	0.14
##	NP_004453	-0.47		1.00		-0.01		0.20		0.26	-0.25
##	NP_003212	-0.24		-0.01		1.00		-0.14		-0.08	-0.16
	NP_060866	-0.16		0.20		-0.14		1.00		-0.01	0.14
##	NP_001160163	-0.37		0.26		-0.08		-0.01		1.00	-0.38
##	NP_003875	0.14		-0.25		-0.16		0.14		-0.38	1.00
	NP_002677	-0.06		0.20		0.07		0.06		0.20	-0.12
	NP_001135891	-0.09		0.03		-0.09		0.22		0.19	-0.04
	NP_004522	0.22		0.02		-0.19		0.20		-0.07	-0.01
	NP_003767	0.04		0.08		-0.17		0.23		-0.05	0.01
	NP_848597	0.36		-0.29		-0.17		0.06		-0.35	0.20
	NP_005970	0.56		-0.45		-0.21		-0.15		-0.20	0.25
##	ND 000470	_	NP_	_		_		_		NP_848597	_
	NP_038479	-0.04			.03		00	-0.		0.38	0.27
	NP_001258898	0.09			. 18	-0.			06	-0.36	-0.41
	NP_542785	-0.03			. 17		33	-0.		-0.03	0.10
	NP_002005	0.02			.34		27		36	0.18	-0.15
	XP_003846537 NP_660208	-0.14 0.13			.04		08	-0.	25	0.26 0.23	0.21 0.28
	NP_000208 NP_004439	0.13			. 13		01		23	0.23	-0.16
	NP_004439	0.33			.00	-0.			07	-0.26	-0.20
	NP_001229372	0.23			. 14	-0.			15	0.13	-0.10
	NP_005130	0.12			. 16	-0.		-0.		-0.05	0.14
	NP_004243	0.12			. 27		38		41	0.03	0.00
	NP_003144	-0.02			. 16		14	-0.		0.20	0.47
	NP_001171551	0.02			. 16		15	-0.		0.22	0.48
	NP_072096	0.10			.03		13		07	0.27	0.28
	NP 005484	0.10			.06		18	-0.		-0.28	-0.22
	NP_061170	0.04			.01		14		14	-0.15	-0.16
	NP_653304	0.04			. 13	-0.			18	-0.23	-0.43
	NP_653081	0.18			. 14	-0.			09	0.10	-0.11
	NP_057164	-0.06			.09		22		04	0.10	0.56
	NP_004453	0.20			.03		02		08	-0.29	-0.45
	NP_003212	0.07			.09	-0.		-0.		-0.17	-0.21
		0.01			- •	٠.		٠.		· · - ·	· ·

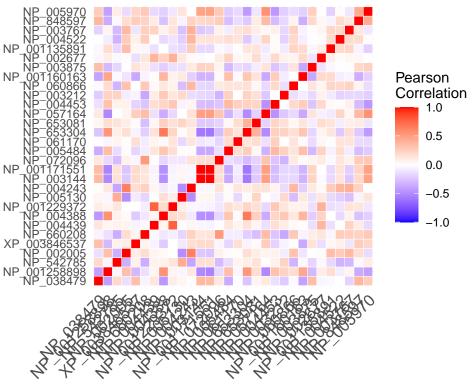
```
0.06
                                   0.22
                                              0.20
                                                        0.23
## NP_060866
                                                                   0.06
                                                                            -0.15
## NP_001160163
                      0.20
                                   0.19
                                             -0.07
                                                       -0.05
                                                                  -0.35
                                                                            -0.20
## NP 003875
                                             -0.01
                                                        0.01
                                                                   0.20
                     -0.12
                                  -0.04
                                                                             0.25
## NP_002677
                                              0.02
                                                        0.05
                                                                            -0.01
                      1.00
                                   0.09
                                                                   0.04
## NP_001135891
                      0.09
                                   1.00
                                              0.19
                                                        0.21
                                                                  -0.05
                                                                            -0.20
## NP 004522
                      0.02
                                   0.19
                                              1.00
                                                        0.19
                                                                   0.19
                                                                             0.02
## NP 003767
                      0.05
                                   0.21
                                              0.19
                                                        1.00
                                                                   0.12
                                                                            -0.10
## NP_848597
                                  -0.05
                                                        0.12
                                                                             0.43
                      0.04
                                              0.19
                                                                   1.00
## NP_005970
                     -0.01
                                  -0.20
                                              0.02
                                                       -0.10
                                                                   0.43
                                                                             1.00
```

We can say that the proteins are not highly correlated

```
melted_cormat <- reshape2::melt(cormat)</pre>
```

Visualizing the correlation plot

Correlation plot



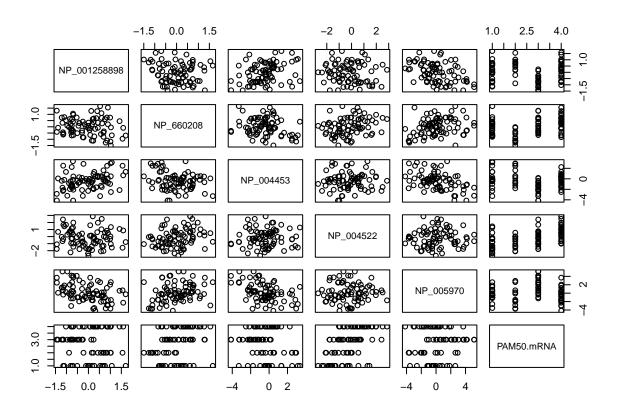
Top 5 variables and response variable

```
top_variables <- c("PAM50.mRNA", "NP_660208", "NP_005970", "NP_004453", "NP_004522", "NP_001258898")

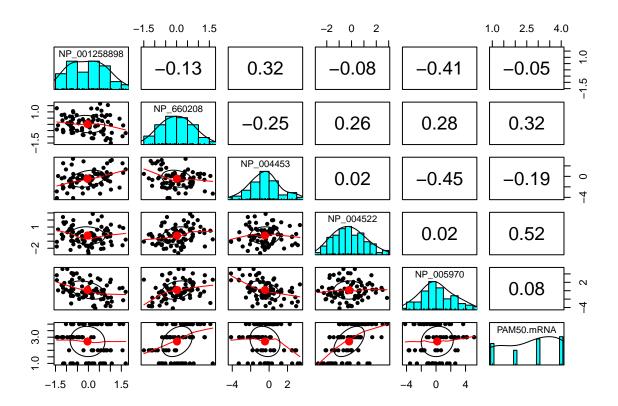
#"NP_001160163", "NP_003767", "NP_542785", "NP_060866", "NP_004388"

data_viz= data_norm[,(names(data_norm) %in% top_variables)]
```

pairs(data_viz)

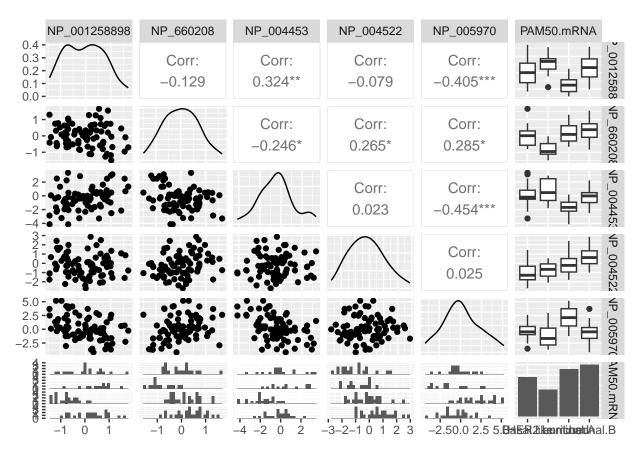


pairs.panels(data_viz)



ggpairs(data_viz)

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



The variable names are displayed on the outer edges of the matrix. The boxes along the diagonals display the density plot for each variable. The boxes in the lower left corner display the scatterplot between each variable. The boxes in the upper right corner display the Pearson correlation coefficient between each variable. The variables are not correlated but according to the distribution they have normal distribution

As this is a multi-class classification, I am going to use SVM, Neural Networks, Decison tree and Naive Baye's. 1. SVM: I chose this algorithm as it does complex data transformations depending on the selected kernel function and based on that transformations, it tries to maximize the separation boundaries between the data points depending on the labels or classes defined. SVM works well for binary classification. But, For multi-class classification, the same principle for binary classification is utilized after breaking down the multiclassification problem into multiple binary classification problems. The main reason behind choosing this algorithm is it works well with expression data. As I am dealing with the protein expression data, this algorithm will work well and compatiable with the features in the data-set. The kernel function is used where the complexity of the problem is high and data is linearly separable. where it adds multiple polynomial features at a very high degree this way it prevents the computational complexity or burden that comes along adding multiple features to data. The dependencies of the variable are always taken into consideration

- 2. Neural Networks: I chose this algorithm as this works well with complex datset. The output layer contains one neuron per class rather than just one neuron. If the dataset contains four classes, then the output layer has four neurons. Therefore, it works well with multi class classification problems.
- 3. Random Forests: I chose this algorithm as this a classifier is a systematic approach for multiclass classification. It poses a set of questions to the dataset (related to its attributes/features). The benefit of this method is that it can handle missing values and maintains accuracy for missing data. It won't over fit the model. t handles dataset with higher dimensionality

4. Naive bayes: I chose naive bayes algorithm as this is highly compatiable for protein expression datset and the calculations of the probabilities for each class are simplified to make their calculations tractable.

As this is a classification problem, Classification Metrics are :Confusion Matrix, Precision, Recall, F1-score and AUC I chose these metrics as it is a well-balanced dataset and these metrics works well.

Creation of training & validation subsets

- Data splitting is done in 70:20 ratio
- Partition is created using createDataPartition function

Construction of at least three related models

- I built 4 models which are as follows:
 - Support Vector Machine (svm())
 - Neural Network (neuralnet())
 - Naive Bayes (naive_bayes())
 - Random Forest (randomForest())

Evaluation of fit of models with holdout method

• For model evaluation, I have calculated accuracy of each model using the confusionMatrix function. I have also compared the sensitivity, specificity, precision, recall and AUC of classification models

For a balanced dataset, we use a confusion matrix and the derived performance metrics; accuracy, precision, recall, F1-score and AUC

```
#creating test/train split index
set.seed(1000)
samp <- createDataPartition(data_norm$PAM50.mRNA, p = 0.7, list = FALSE)
train <- data_norm[samp, ]
test<- data_norm[-samp, ]</pre>
```

```
#exploring training dataset
head(train)
```

```
##
     NP_038479 NP_001258898
                              NP_542785 NP_002005 XP_003846537
                                                                  NP_660208
## 1 -0.3708959 -0.01456473
                             0.05890562 -2.1305106
                                                     -0.4957955 -0.09170859
## 2 -1.6044748 -0.51053492 3.17273051 -0.9688522
                                                     -0.6711132 -0.51388030
## 3 -3.0516640
                 0.53387720
                             3.71857004 -0.3435704
                                                     -0.3498830 1.65751511
                             2.96770134 -3.2821848
## 4 -1.0663490
                 1.69892745
                                                     -0.4373841 -0.70126018
## 5 -0.3009396 -1.09705962 0.03501174 -1.2946781
                                                     -0.3376402 0.78031558
## 6 0.6787788 -0.70384668 -1.91679672 -2.7445377
                                                     -0.4103540 0.41280109
    NP_004439 NP_004388 NP_001229372 NP_005130 NP_004243 NP_003144 NP_001171551
                                                                        0.1948258
## 1 -4.419112 0.1691111
                            -3.004808 2.2373013 -3.717470 0.1948258
## 2 -5.187379 0.4863889
                            -3.273820 -2.2066435 -2.905828 0.9748146
                                                                        0.9547423
                            -2.969601 -0.4698218 -3.076914 0.3539689
## 3 -3.099008 -0.7349499
                                                                        0.3508126
                            -2.859260 3.8894601 -3.180972 0.3397851
## 4 -3.130366 0.7843981
                                                                        0.3470145
## 5 -2.062567 -0.1823685
                            -1.774609 -0.2388310 -1.785901 1.5453813
                                                                        1.5990205
## 6 -3.322351 -0.2659006
                            -2.127745 -0.6167161 -1.937433 1.4835906
                                                                        1.5432063
```

```
NP 072096
                 NP 005484
                             NP 061170 NP 653304 NP 653081 NP 057164
## 1 -0.5215101 -0.96600568 -1.41417478 0.5548304 -0.2753844 -0.6243686
## 2 -0.5506795 -0.29643048 -2.19660738 -0.8751815 0.2890113 0.6034772
## 3 0.8747561 0.29399945 -3.01378855 -1.5334903 -1.4072388 1.8279546
## 4 -1.2145533 -0.11205750 -1.51096203 0.1301301 0.3723177 -0.1771228
## 5 0.2693307 0.62504394 0.48671103 -1.4414803 0.6448058 1.4155177
     0.3623570 0.07115731 -0.06412445 -2.9967579 -0.3690816 0.6375064
     NP 004453
                  NP 003212
                               NP_060866 NP_001160163
                                                       NP 003875 NP 002677
## 1 -2.3105130 0.007476375 -1.399480707 1.340963113 0.04810062 -7.684869
## 2 -0.3499566 -0.376719645 -1.457277986 0.740637896 1.71079864 -8.877335
## 3 -3.2915417 -0.476134402 0.335031167
                                         2.493930969 -3.12110229 -7.479934
     1.3374534 0.159048035 -0.451843095
                                         1.055503657 -0.82416136 -7.963274
## 5 -0.2021304 -1.819778494 -0.600190397
                                         1.886978849 0.83677799 -4.388818
    3.3454346 -2.334106568 0.002369978 0.002369978 -0.66257428 -3.342988
    NP_001135891 NP_004522 NP_003767 NP_848597 NP_005970 PAM50.mRNA
## 1
       0.9038146 -2.3839833 -0.3892635 0.3819222
                                                  0.8193237 Basal.like
## 2
      -1.2933543 -0.7514024 -1.0558322 -0.7346755 -0.9655069 Basal.like
      -0.4603530 -0.7412624 -1.0095468 -0.1952249 2.6107136 Basal.like
## 4
      -1.7784528 -0.7771697 -0.7446371 0.2277281 -0.2638766 Basal.like
## 5
       0.6306902 1.4493952 -0.5521973 1.9688493
                                                 1.8389858 Basal.like
## 6
      -1.0913487 1.2864003 0.2889839 -0.5502216 1.5019339 Basal.like
```

dim(train)

[1] 59 31

head(test)

```
NP_038479 NP_001258898 NP_542785 NP_002005 XP_003846537
                                                               NP_660208
                   0.6862870 1.2085340 -1.994803
     -1.6621615
                                                  -0.4081033 -0.80727303
## 12 -1.0424112
                  -0.1083678 -1.0174124 -1.919639
                                                  -0.5742532 0.01208057
                  -0.7426185 1.6022647 -2.311497
                                                  -0.8640801 -0.25677221
## 14 -1.3904136
## 17 0.9718405
                  0.1667926 1.4009461 -3.102765
                                                   -0.6610396 0.19337440
                   0.7547515 1.3304690 -4.288958
## 18 -1.9684279
                                                   -0.2695313 0.60993907
## 23 -2.0569715
                  0.2201496 -0.3960125 -2.099070
                                                  -0.8055116 -1.11933337
                 NP 004388 NP 001229372 NP 005130 NP 004243 NP 003144
     NP 004439
##
     -5.421010 -0.05217698
                             -4.053853 -1.0833654 -3.112478 -0.9070655
## 12 -1.396939 -0.08968960
                             -1.969637 3.2369163 -0.967415 -0.4265335
## 14 -1.532119 0.38764892
                           -0.766236 -0.3748599 -4.349352 0.9274781
## 17 -3.383772 -0.27750263
                           -2.685051 4.9629030 -3.679969 -0.6914187
## 18 -4.140614 1.20684870
                             -5.323837 0.7264954 -3.681453 -0.9794653
                             -2.902759 3.5956468 -2.114378 0.5645881
## 23 -2.795601 -0.11280757
     NP_001171551 NP_072096
                              NP_005484 NP_061170 NP_653304
                                                               NP_{653081}
##
## 7
       -0.9070655 -0.4779580
                            0.48670214 -0.8804541 0.1241230
                                                              0.35031915
## 12
       2.3551430 0.05071496
        0.9072345 0.3910229 -0.72574885 -1.5624841 -2.3182450 -0.26014614
## 14
## 17
       -0.6800265 -0.3306662 \ 0.36805459 -3.4103538 -0.6154708 -0.23193388
## 18
       -0.9335491 1.0302482 0.65232319 0.5710870 1.2598288 -0.58741219
## 23
        0.5684152 -0.8131658 -0.05540115 -4.7129749 -0.7902033 -0.29650809
                  NP 004453 NP 003212 NP 060866 NP 001160163
##
      NP 057164
                                                               NP 003875
     -0.7074806 -0.19188638 1.3183057 0.2571795
                                                  2.08670742 -0.07878830
## 7
## 12 0.1188933 -0.02428122 0.8756729 -0.9151449 -1.36284939
                                                             2.66194558
## 14  0.4146404 -1.74805043 -0.3141291 -0.7729839  0.72504218  0.04810062
```

```
## 17 -1.1850801 -0.11421462 -0.1825677 -0.6686344
                                                      0.87690557 -1.37115250
## 18 -1.8271476 -0.49911195 -0.4143437 -1.1030856
                                                      2.35475182 -0.94061315
## 23 -0.4074938 -1.80055616
                             2.9986201 -0.5414421
                                                      0.06316141 -0.41897511
##
      NP_002677 NP_001135891
                              NP_004522
                                                       NP_848597
                                                                  NP_005970
                                           NP_003767
## 7
      -7.080890
                   0.4667437 -1.2896031 -1.359457807 -1.7752596 -0.4746316
## 12 -3.814997
                  -0.2879042 -1.6423806 1.077935362 -1.1696774 -1.1787679
## 14 -5.732665
                  -1.7446765 -0.2837637 0.006394535
                                                       0.6676853
## 17 -9.444415
                  -0.8850859 -1.9103826 -1.731905061 -0.1977573 -0.9724260
## 18 -9.745913
                  -1.5481188 0.5569589 -1.254962003 -0.5838802 -1.0818935
                   0.4957004 -0.2807051 -0.503171187 -0.6639091 -3.7141033
## 23 -3.737066
##
         PAM50.mRNA
         Basal.like
## 7
## 12
         Basal.like
## 14
         Basal.like
## 17
         Basal.like
## 18
         Basal.like
## 23 HER2.enriched
dim(test)
```

[1] 21 31

Model building

As this is a multi-class classification, I am implement 4 models - SVM, Neural Networks, Naive Baye's and Random forests

SVM linear classifier

Support Vector Machines are generalized extension of a maximal margin classifier, SVM are intended for the binary classification setting when there are two classes. However, this designed intention does not disqualify from using the SVM method with cases of more than two classes. SVM determines the best line separator by identifying closest points in Convex hull, a hyperplane bisects the closest point to the convex hull. The support vector classifies a test observation depending on which side of a plane it lies; this is based on boundaries-support vectors. SVM method allows some observations to be on the incorrect side of the margin and in some cases the incorrect side of the hyperplane in the interest of performing better in classifying the remaining observations further away from the hydroplane. This is known as a soft margin classifier; training observations can violate this area. Advantages of using a SVM model are; can be adapted to work well with nonlinear boundaries, uses kernels, less overfitting of data, performs well with clear margin of separation among data. The kernel function is used where the complexity of the problem is high and data is linearly separable. where it adds multiple polynomial features at a very high degree this way it prevents the computational complexity or burden that comes along adding multiple features to data. The dependencies of the variable are always taken into consideration

```
svm_model1 <- svm(PAM50.mRNA~., data= train, type="C-classification", kernel = 'linear')
svm_model1

##
## Call:
## svm(formula = PAM50.mRNA ~ ., data = train, type = "C-classification",</pre>
```

```
##
       kernel = "linear")
##
##
## Parameters:
##
      SVM-Type: C-classification
  SVM-Kernel: linear
##
##
         cost: 1
##
## Number of Support Vectors: 41
svm_pred <- predict(svm_model1, newdata = test)</pre>
confusionMatrix(svm_pred, factor(data_norm$PAM50.mRNA[-samp]), mode = "everything")
## Confusion Matrix and Statistics
##
##
                  Reference
## Prediction
                  Basal.like HER2.enriched Luminal.A Luminal.B
                                          0
                                                    0
##
    Basal.like
                           5
    HER2.enriched
                            0
                                          2
                                                     0
                                                               0
##
    Luminal.A
                            0
                                          0
                                                     6
                                                               1
##
    Luminal.B
                                                     0
##
## Overall Statistics
##
                  Accuracy: 0.8095
##
##
                    95% CI: (0.5809, 0.9455)
##
       No Information Rate: 0.3333
##
       P-Value [Acc > NIR] : 1.026e-05
##
##
                     Kappa: 0.7399
##
  Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
                        Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                   1.0000
                                                       0.66667
                                                                          1.0000
## Specificity
                                   0.8750
                                                        1.00000
                                                                          0.9333
## Pos Pred Value
                                   0.7143
                                                        1.00000
                                                                          0.8571
## Neg Pred Value
                                   1.0000
                                                        0.94737
                                                                          1.0000
## Precision
                                   0.7143
                                                        1.00000
                                                                          0.8571
## Recall
                                   1.0000
                                                        0.66667
                                                                          1.0000
## F1
                                   0.8333
                                                        0.80000
                                                                          0.9231
## Prevalence
                                   0.2381
                                                        0.14286
                                                                          0.2857
## Detection Rate
                                   0.2381
                                                        0.09524
                                                                          0.2857
## Detection Prevalence
                                   0.3333
                                                        0.09524
                                                                          0.3333
## Balanced Accuracy
                                   0.9375
                                                        0.83333
                                                                          0.9667
##
                        Class: Luminal.B
## Sensitivity
                                  0.5714
## Specificity
                                  0.9286
## Pos Pred Value
                                  0.8000
## Neg Pred Value
                                  0.8125
## Precision
                                  0.8000
## Recall
                                  0.5714
```

```
## F1 0.6667
## Prevalence 0.3333
## Detection Rate 0.1905
## Detection Prevalence 0.2381
## Balanced Accuracy 0.7500
```

```
accuracy_svm <-confusionMatrix(test$PAM50.mRNA, svm_pred)$overall["Accuracy"]
```

Model evaluation for SVM:

- 1. Accuracy: The overall model accuracy of SVM model is 81%
- 2. Precision,Recall,F1: The Precision,Recall,F1 for Basal.like class is 0.7143, 1.0000, 0.8333 The Precision,Recall,F1 for HER2.enriched class is 1.00000, 0.33333, 0.80000

 The Precision,Recall,F1 for Luminal.A class is 0.8571, 1.0000, 0.9231 The Precision,Recall,F1 for Luminal.B class is 0.8000, 0.5714, 0.6667
- 3. Sensitivity and Specificity: The Sensitivity and Specificity for Basal.like class is 1.0000 and 0.8750 The Sensitivity and Specificity for HER2.enriched class is 0.66667 and 1.00000 The Sensitivity and Specificity for Luminal.A is 1.0000 and 0.9333 The Sensitivity and Specificity for Luminal.B class is 0.5714 and 0.9286
- 4. The Kappa statistic: The kappa value for this model is 0.7399 which states that it is a good agreement.

Macro-averaged Metrics: The per-class metrics can be averaged over all the classes resulting in macro-averaged precision, recall and F-1.

```
# macro-averaged precision
precision_svm <- c(0.7143, 1.00000,0.8571, 0.8000)
macro_precision_svm <- mean(precision_svm)
# macro-averaged recall
recall_svm <- c(1.0000, 0.66667, 1.0000, 0.5714)
macro_recall_svm <- mean(recall_svm)
# macro-averaged F-1
F1_svm<- c(0.8333,0.80000, 0.9231, 0.6667)
macroF1_svm <- mean(F1_svm)
macro_avg_svm <- data.frame( macro_precision_svm, macro_recall_svm, macroF1_svm)
macro_avg_svm</pre>
```

AUC for SVM

```
svm_auc <- multiclass.roc(test$PAM50.mRNA, as.ordered(svm_pred))
auc(svm_auc)</pre>
```

```
## Multi-class area under the curve: 0.7897
```

The AUC of SVM model is : 0.7897

```
Name_metrics <- c("Accuracy", "Precision", "Recall", "F-1", "AUC", "Kappa")
values_svm <- c(0.8095 , 0.84285,  0.8095175, 0.805775,0.7897, 0.7399 )
metrics_svm <- data.frame(Name_metrics, values_svm)
print (metrics_svm)</pre>
```

```
## Name_metrics values_svm
## 1 Accuracy 0.8095000
## 2 Precision 0.8428500
## 3 Recall 0.8095175
## 4 F-1 0.8057750
## 5 AUC 0.7897000
## 6 Kappa 0.7399000
```

Neural Networks

Inspired by biological neural networks, the Neural Network method is a supervised machine learning algorithm which consists of units arranged in layers which coverts an input vector (independent variable) into a prediction/classification. "The algorithm learns a function by training on a dataset without prior knowledge about the dataset."

```
# I have tried out many hidden layers but got optimal results with hidden layer = 5
neuralnet_model2 <- train(PAM50.mRNA~., data= train, hidden = 5, method = "nnet")</pre>
```

```
## # weights: 39
## initial value 90.443203
## iter 10 value 33.522949
## iter 20 value 29.081551
## iter 30 value 25.631000
## iter
       40 value 25.517964
       50 value 25.395198
## iter
## iter 60 value 25.275651
## iter 70 value 25.125594
## iter 80 value 25.083164
## iter 90 value 25.080908
## iter 100 value 25.080875
## final value 25.080875
## stopped after 100 iterations
## # weights: 109
## initial value 95.605172
## iter 10 value 31.720806
## iter 20 value 21.847426
## iter 30 value 18.902135
## iter
       40 value 16.921231
## iter 50 value 16.448390
## iter
        60 value 14.611413
## iter 70 value 14.256455
## iter 80 value 13.923560
## iter 90 value 4.786360
## iter 100 value 4.500853
## final value 4.500853
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 115.225552
## iter 10 value 16.288091
## iter 20 value 4.194251
## iter 30 value 0.250901
## iter 40 value 0.023452
## iter 50 value 0.010547
## iter 60 value 0.002400
## iter 70 value 0.000788
## iter 80 value 0.000229
## iter 90 value 0.000140
## final value 0.000085
## converged
## # weights:
              39
## initial value 99.122639
## iter 10 value 55.167500
## iter 20 value 44.898518
## iter 30 value 41.116677
## iter 40 value 40.029601
## iter 50 value 38.251906
## iter 60 value 36.067644
## iter 70 value 34.828721
## iter 80 value 34.678452
## final value 34.678386
## converged
## # weights: 109
## initial value 125.755935
## iter 10 value 49.563245
## iter 20 value 30.684360
## iter 30 value 17.693543
## iter 40 value 13.908672
## iter 50 value 13.614344
## iter 60 value 13.433069
## iter 70 value 13.086003
## iter 80 value 11.668672
## iter 90 value 11.218409
## iter 100 value 11.214485
## final value 11.214485
## stopped after 100 iterations
## # weights: 179
## initial value 91.656334
## iter 10 value 33.201743
## iter 20 value 12.681427
## iter 30 value 9.869531
## iter 40 value 9.214958
## iter 50 value 9.131412
## iter 60 value 9.066018
## iter 70 value 8.996023
## iter 80 value 8.895696
## iter 90 value 8.776035
## iter 100 value 8.769974
## final value 8.769974
## stopped after 100 iterations
## # weights: 39
## initial value 87.201502
```

```
## iter 10 value 41.199495
## iter 20 value 38.238237
## iter 30 value 37.143784
## iter 40 value 35.524119
## iter 50 value 35.018233
## iter 60 value 34.042314
## iter 70 value 28.906279
## iter 80 value 26.316558
## iter 90 value 25.287137
## iter 100 value 23.321447
## final value 23.321447
## stopped after 100 iterations
## # weights: 109
## initial value 83.188527
## iter 10 value 28.268240
## iter 20 value 21.936391
## iter 30 value 20.065399
## iter
       40 value 18.630961
## iter 50 value 17.981643
## iter 60 value 7.789546
## iter 70 value 3.318506
## iter 80 value 1.922637
## iter 90 value 0.182460
## iter 100 value 0.162306
## final value 0.162306
## stopped after 100 iterations
## # weights: 179
## initial value 88.663771
## iter 10 value 11.817962
## iter 20 value 4.425132
## iter 30 value 4.356586
## iter 40 value 4.330111
## iter
       50 value 3.455265
## iter 60 value 1.619989
## iter 70 value 1.546958
## iter 80 value 1.538854
## iter 90 value 1.520210
## iter 100 value 0.157354
## final value 0.157354
## stopped after 100 iterations
## # weights: 39
## initial value 94.783065
## iter 10 value 53.592345
## iter 20 value 47.413612
## iter 30 value 47.405922
## final value 47.405898
## converged
## # weights: 109
## initial value 87.084715
## iter 10 value 26.023255
## iter 20 value 10.114774
## iter 30 value 6.342580
## iter 40 value 2.451402
## iter 50 value 0.385441
```

```
## iter 60 value 0.004775
## final value 0.000074
## converged
## # weights: 179
## initial value 97.163515
## iter 10 value 2.318107
## iter 20 value 0.034487
## iter 30 value 0.003940
## iter 40 value 0.000306
## final value 0.000056
## converged
## # weights: 39
## initial value 86.806419
## iter 10 value 75.907903
## iter 20 value 54.688468
## iter 30 value 50.443480
## iter 40 value 46.229664
## iter 50 value 45.961644
## final value 45.958668
## converged
## # weights: 109
## initial value 85.421652
## iter 10 value 36.382018
## iter 20 value 18.181113
## iter 30 value 13.355138
## iter 40 value 12.954718
## iter 50 value 12.933512
## iter 60 value 12.932679
## iter 70 value 12.932642
## final value 12.932639
## converged
## # weights: 179
## initial value 83.841104
## iter 10 value 18.837481
## iter 20 value 11.161662
## iter 30 value 9.930130
## iter 40 value 9.566609
## iter 50 value 9.197280
## iter 60 value 9.032886
## iter 70 value 9.028200
## final value 9.028183
## converged
## # weights: 39
## initial value 84.493812
## iter 10 value 48.961837
## iter 20 value 47.120962
## iter 30 value 47.082685
## iter 40 value 47.041195
## iter 50 value 44.496628
## iter 60 value 41.900770
## iter 70 value 41.830721
## iter 80 value 41.820574
## iter 90 value 41.815581
```

iter 100 value 39.414005

```
## final value 39.414005
## stopped after 100 iterations
## # weights: 109
## initial value 88.138094
## iter 10 value 21.747293
## iter 20 value 15.048828
## iter 30 value 4.739651
## iter 40 value 3.025010
## iter 50 value 0.208822
## iter 60 value 0.137060
## iter 70 value 0.124229
## iter 80 value 0.115481
## iter 90 value 0.102526
## iter 100 value 0.098002
## final value 0.098002
## stopped after 100 iterations
## # weights: 179
## initial value 91.373849
## iter 10 value 20.036539
## iter 20 value 16.034971
## iter 30 value 13.476093
## iter 40 value 8.681448
## iter 50 value 8.207185
## iter 60 value 8.184184
## iter 70 value 5.238894
## iter 80 value 0.161448
## iter 90 value 0.133480
## iter 100 value 0.121600
## final value 0.121600
## stopped after 100 iterations
## # weights: 39
## initial value 86.274487
## iter 10 value 54.969200
## iter 20 value 46.295371
## iter 30 value 43.451786
## iter 40 value 35.825560
## iter 50 value 30.188028
## iter 60 value 29.345814
## iter 70 value 29.332487
## iter 80 value 29.311485
## iter 90 value 28.577676
## iter 100 value 27.005402
## final value 27.005402
## stopped after 100 iterations
## # weights: 109
## initial value 80.318196
## iter 10 value 33.230696
## iter 20 value 29.954567
## iter 30 value 21.628038
## iter 40 value 19.174917
## iter 50 value 19.114320
## iter 60 value 19.108960
## iter 70 value 17.916892
## iter 80 value 17.492409
```

```
## iter 90 value 17.491062
## iter 100 value 16.469333
## final value 16.469333
## stopped after 100 iterations
## # weights: 179
## initial value 116.205737
## iter 10 value 25.710360
## iter 20 value 0.588619
## iter 30 value 0.001871
## iter 40 value 0.000492
## final value 0.000041
## converged
## # weights: 39
## initial value 85.707050
## iter 10 value 59.869708
## iter 20 value 50.880603
## iter 30 value 47.953242
## iter 40 value 47.821941
## iter 50 value 47.708402
## iter 60 value 45.499533
## iter 70 value 44.250236
## iter 80 value 44.003961
## final value 44.003956
## converged
## # weights: 109
## initial value 82.029705
## iter 10 value 43.266080
## iter 20 value 18.106270
## iter 30 value 14.673940
## iter 40 value 14.289562
## iter 50 value 14.087079
## iter 60 value 14.022807
## iter 70 value 14.002406
## iter 80 value 13.996983
## iter 90 value 13.817015
## iter 100 value 13.636175
## final value 13.636175
## stopped after 100 iterations
## # weights: 179
## initial value 98.834823
## iter 10 value 36.380523
## iter 20 value 16.515008
## iter 30 value 12.663731
## iter 40 value 10.549273
## iter 50 value 10.060107
## iter 60 value 9.870984
## iter 70 value 9.487019
## iter 80 value 9.401886
## iter 90 value 9.366593
## iter 100 value 9.363588
## final value 9.363588
## stopped after 100 iterations
## # weights: 39
## initial value 92.269604
```

```
## iter 10 value 45.213232
## iter 20 value 33.655342
## iter 30 value 25.408527
## iter 40 value 21.498229
## iter 50 value 19.126444
## iter 60 value 18.590005
## iter 70 value 17.211918
## iter 80 value 15.679628
## iter 90 value 15.526127
## iter 100 value 15.486981
## final value 15.486981
## stopped after 100 iterations
## # weights: 109
## initial value 86.556071
## iter 10 value 21.126144
## iter 20 value 12.730770
## iter 30 value 8.742250
## iter
       40 value 5.295685
## iter 50 value 5.189397
## iter 60 value 2.870770
## iter 70 value 2.670501
## iter 80 value 2.568537
## iter 90 value 2.551319
## iter 100 value 2.543506
## final value 2.543506
## stopped after 100 iterations
## # weights: 179
## initial value 85.899796
## iter 10 value 19.404934
## iter 20 value 1.100866
## iter 30 value 0.267923
## iter 40 value 0.246658
## iter 50 value 0.208982
## iter 60 value 0.191439
## iter 70 value 0.169347
## iter 80 value 0.153511
## iter 90 value 0.131053
## iter 100 value 0.116548
## final value 0.116548
## stopped after 100 iterations
## # weights: 39
## initial value 85.430115
## iter 10 value 57.225354
## iter 20 value 53.384902
## iter 30 value 51.057077
## iter 40 value 50.925104
## iter 50 value 50.922126
## iter 60 value 50.921837
## iter 70 value 50.920673
## final value 50.920351
## converged
## # weights: 109
## initial value 85.167665
## iter 10 value 43.130540
```

```
## iter 20 value 23.554205
## iter 30 value 21.552252
## iter 40 value 20.071984
## iter 50 value 19.914447
## iter 60 value 19.901990
## iter 70 value 19.892129
## iter 80 value 18.423996
## iter 90 value 10.923661
## iter 100 value 10.823521
## final value 10.823521
## stopped after 100 iterations
## # weights: 179
## initial value 88.286280
## iter 10 value 4.681288
## iter 20 value 0.043541
## iter 30 value 0.018550
## iter 40 value 0.003462
## iter 50 value 0.001255
## final value 0.000069
## converged
## # weights: 39
## initial value 94.891027
## iter 10 value 55.411517
## iter 20 value 48.375786
## iter 30 value 46.518448
## iter 40 value 46.394271
## iter 50 value 46.219659
## iter 60 value 45.602768
## iter 70 value 45.547223
## final value 45.547177
## converged
## # weights: 109
## initial value 87.111745
## iter 10 value 30.400489
## iter 20 value 19.456495
## iter 30 value 14.712802
## iter 40 value 13.830381
## iter 50 value 13.701561
## iter 60 value 13.696985
## iter 70 value 13.696311
## iter 80 value 13.696173
## final value 13.696161
## converged
## # weights: 179
## initial value 104.196537
## iter 10 value 39.403897
## iter 20 value 17.959820
## iter 30 value 12.788810
## iter 40 value 10.185183
## iter 50 value 9.919076
## iter 60 value 9.778906
## iter 70 value 9.770507
## iter 80 value 9.770228
## iter 90 value 9.770166
```

```
## final value 9.770163
## converged
## # weights: 39
## initial value 88.889867
## iter 10 value 52.154530
## iter 20 value 47.815507
## iter 30 value 46.865824
## iter 40 value 46.568644
## iter 50 value 46.541225
## iter 60 value 45.921930
## iter 70 value 45.329138
## iter 80 value 45.327353
## iter 90 value 45.317866
## iter 100 value 45.312219
## final value 45.312219
## stopped after 100 iterations
## # weights: 109
## initial value 95.002202
## iter 10 value 28.105194
## iter 20 value 0.797318
## iter 30 value 0.283174
## iter 40 value 0.243165
## iter 50 value 0.213412
## iter 60 value 0.179481
## iter 70 value 0.158125
## iter 80 value 0.147890
## iter 90 value 0.138695
## iter 100 value 0.133685
## final value 0.133685
## stopped after 100 iterations
## # weights: 179
## initial value 94.277604
## iter 10 value 13.652133
## iter 20 value 4.123864
## iter 30 value 3.446456
## iter 40 value 3.020463
## iter 50 value 2.823368
## iter 60 value 2.772622
## iter 70 value 2.289223
## iter 80 value 1.651327
## iter 90 value 0.414348
## iter 100 value 0.227118
## final value 0.227118
## stopped after 100 iterations
## # weights: 39
## initial value 94.460149
## iter 10 value 51.389012
## iter 20 value 43.140962
## iter 30 value 33.502693
## iter 40 value 27.848324
## iter 50 value 26.824194
## iter 60 value 26.641870
## iter 70 value 26.620122
## iter 80 value 26.598139
```

```
## iter 90 value 26.588809
## iter 100 value 26.566875
## final value 26.566875
## stopped after 100 iterations
## # weights: 109
## initial value 90.391599
## iter 10 value 7.930567
## iter 20 value 3.633502
## iter 30 value 3.526730
## iter 40 value 3.443888
## iter 50 value 2.982633
## iter 60 value 0.006821
## iter 70 value 0.002643
## iter 80 value 0.000379
## iter 90 value 0.000223
## iter 100 value 0.000128
## final value 0.000128
## stopped after 100 iterations
## # weights: 179
## initial value 87.133673
## iter 10 value 12.481079
## iter 20 value 3.318230
## iter 30 value 1.831807
## iter 40 value 0.056883
## iter 50 value 0.013627
## iter 60 value 0.006736
## iter 70 value 0.003355
## iter 80 value 0.001945
## iter 90 value 0.001460
## iter 100 value 0.000847
## final value 0.000847
## stopped after 100 iterations
## # weights: 39
## initial value 88.563420
## iter 10 value 47.640507
## iter 20 value 44.315735
## iter 30 value 42.430785
## iter 40 value 41.526490
## iter 50 value 41.412117
## iter 60 value 40.455560
## iter 70 value 40.143907
## final value 40.139625
## converged
## # weights: 109
## initial value 85.979869
## iter 10 value 37.610561
## iter 20 value 18.706558
## iter 30 value 14.114685
## iter 40 value 13.069311
## iter 50 value 12.948879
## iter 60 value 12.924478
## iter 70 value 12.920273
## iter 80 value 12.843811
## iter 90 value 12.714612
```

```
## iter 100 value 12.711120
## final value 12.711120
## stopped after 100 iterations
## # weights: 179
## initial value 87.609021
## iter 10 value 29.697635
## iter 20 value 11.896680
## iter 30 value 9.670379
## iter 40 value 9.141076
## iter 50 value 9.085234
## iter 60 value 9.072865
## iter 70 value 9.072378
## iter 80 value 9.072302
## final value 9.072299
## converged
## # weights: 39
## initial value 84.685863
## iter 10 value 61.851769
## iter 20 value 42.560284
## iter 30 value 41.078396
## iter 40 value 40.657067
## iter 50 value 40.600519
## iter 60 value 40.594214
## iter 70 value 40.592003
## iter 80 value 40.580621
## iter 90 value 37.091250
## iter 100 value 34.069179
## final value 34.069179
## stopped after 100 iterations
## # weights: 109
## initial value 86.006150
## iter 10 value 12.014673
## iter 20 value 8.523674
## iter 30 value 4.504624
## iter 40 value 3.986184
## iter 50 value 3.940254
## iter 60 value 3.464024
## iter 70 value 3.362368
## iter 80 value 3.345236
## iter 90 value 3.301024
## iter 100 value 3.275223
## final value 3.275223
## stopped after 100 iterations
## # weights: 179
## initial value 91.636387
## iter 10 value 19.423609
## iter 20 value 2.315661
## iter 30 value 0.213472
## iter 40 value 0.181507
## iter 50 value 0.164156
## iter 60 value 0.143868
## iter 70 value 0.129505
## iter 80 value 0.119260
## iter 90 value 0.108010
```

```
## iter 100 value 0.101534
## final value 0.101534
## stopped after 100 iterations
## # weights: 39
## initial value 96.098385
## iter 10 value 39.375213
## iter 20 value 37.798208
## iter 30 value 37.538787
## iter 40 value 35.200980
## iter 50 value 33.220584
## iter 60 value 33.181591
## iter 70 value 33.164383
## iter 80 value 33.125503
## iter 90 value 33.024983
## iter 100 value 32.948634
## final value 32.948634
## stopped after 100 iterations
## # weights: 109
## initial value 100.613408
## iter 10 value 34.385032
## iter 20 value 22.629568
## iter 30 value 22.206673
## iter 40 value 22.099804
## iter 50 value 18.525970
## iter 60 value 18.522409
## final value 18.522345
## converged
## # weights: 179
## initial value 109.623683
## iter 10 value 18.936297
## iter 20 value 3.276659
## iter 30 value 2.780370
## iter 40 value 2.772626
## iter 50 value 2.772606
## final value 2.772589
## converged
## # weights: 39
## initial value 100.819689
## iter 10 value 66.110514
## iter 20 value 51.872886
## iter 30 value 47.839847
## iter 40 value 46.910582
## iter 50 value 44.477582
## iter 60 value 42.358120
## iter 70 value 42.182193
## iter 80 value 42.179317
## iter 90 value 42.113525
## iter 100 value 41.294815
## final value 41.294815
## stopped after 100 iterations
## # weights: 109
## initial value 83.972304
## iter 10 value 31.973653
## iter 20 value 18.513537
```

```
## iter 30 value 14.075397
## iter 40 value 13.374546
## iter 50 value 12.785221
## iter 60 value 11.878226
## iter 70 value 11.744054
## final value 11.743838
## converged
## # weights: 179
## initial value 75.849269
## iter 10 value 32.628312
## iter 20 value 14.725494
## iter 30 value 10.022148
## iter 40 value 9.189537
## iter 50 value 9.130171
## iter 60 value 9.080929
## iter 70 value 8.908773
## iter 80 value 8.841742
## final value 8.841481
## converged
## # weights:
## initial value 85.799499
## iter 10 value 43.617362
## iter 20 value 34.249951
## iter 30 value 31.305164
## iter 40 value 30.361464
## iter 50 value 30.290953
## iter 60 value 29.599607
## iter 70 value 28.454951
## iter 80 value 26.964471
## iter 90 value 26.503637
## iter 100 value 26.233916
## final value 26.233916
## stopped after 100 iterations
## # weights: 109
## initial value 84.346112
## iter 10 value 31.238751
## iter 20 value 24.851702
## iter 30 value 17.040986
## iter 40 value 16.468365
## iter 50 value 14.804808
## iter 60 value 14.706717
## iter 70 value 14.691450
## iter 80 value 9.232789
## iter 90 value 7.927592
## iter 100 value 5.251767
## final value 5.251767
## stopped after 100 iterations
## # weights: 179
## initial value 135.816014
## iter 10 value 28.633434
## iter 20 value 3.210789
## iter 30 value 0.332123
## iter 40 value 0.287206
## iter 50 value 0.267787
```

```
## iter 60 value 0.241176
## iter 70 value 0.204326
## iter 80 value 0.161111
## iter 90 value 0.131655
## iter 100 value 0.122657
## final value 0.122657
## stopped after 100 iterations
## # weights: 39
## initial value 84.555277
## iter 10 value 52.401706
## iter 20 value 46.716274
## iter 30 value 44.489603
## iter 40 value 41.728293
## iter 50 value 41.441091
## iter 60 value 40.317956
## iter 70 value 40.163857
## iter 80 value 40.090201
## iter 90 value 40.084586
## iter 100 value 40.084034
## final value 40.084034
## stopped after 100 iterations
## # weights: 109
## initial value 84.461243
## iter 10 value 25.676559
## iter 20 value 10.681158
## iter 30 value 10.241056
## iter 40 value 9.941434
## iter 50 value 9.937546
## iter 60 value 9.937409
## final value 9.937406
## converged
## # weights: 179
## initial value 85.647588
## iter 10 value 1.113502
## iter 20 value 0.015207
## iter 30 value 0.001123
## iter 40 value 0.000121
## final value 0.000095
## converged
## # weights: 39
## initial value 84.493059
## iter 10 value 55.337174
## iter 20 value 49.473363
## iter 30 value 45.855050
## iter 40 value 45.236948
## iter 50 value 45.231476
## final value 45.231474
## converged
## # weights: 109
## initial value 88.576805
## iter 10 value 47.901369
## iter 20 value 19.973965
## iter 30 value 16.340185
## iter 40 value 14.075666
```

```
## iter 50 value 13.241925
## iter 60 value 13.230843
## final value 13.230825
## converged
## # weights: 179
## initial value 96.244694
## iter 10 value 36.815255
## iter 20 value 16.444696
## iter 30 value 13.402294
## iter 40 value 10.660775
## iter 50 value 9.741046
## iter 60 value 9.718537
## iter 70 value 9.717106
## iter 80 value 9.717017
## iter 90 value 9.717015
## final value 9.717014
## converged
## # weights: 39
## initial value 87.527507
## iter 10 value 59.470647
## iter 20 value 54.261550
## iter 30 value 50.852477
## iter 40 value 43.821572
## iter 50 value 37.665862
## iter 60 value 34.906300
## iter 70 value 32.763432
## iter 80 value 30.381801
## iter 90 value 29.891834
## iter 100 value 29.415800
## final value 29.415800
## stopped after 100 iterations
## # weights: 109
## initial value 95.897478
## iter 10 value 19.551051
## iter 20 value 6.736674
## iter 30 value 0.306956
## iter 40 value 0.151007
## iter 50 value 0.130962
## iter 60 value 0.125228
## iter 70 value 0.119796
## iter 80 value 0.114569
## iter 90 value 0.104433
## iter 100 value 0.097994
## final value 0.097994
## stopped after 100 iterations
## # weights: 179
## initial value 92.894485
## iter 10 value 7.813962
## iter 20 value 2.245462
## iter 30 value 2.056088
## iter 40 value 1.739704
## iter 50 value 1.564316
## iter 60 value 1.539415
## iter 70 value 0.203462
```

```
## iter 80 value 0.147995
## iter 90 value 0.138873
## iter 100 value 0.119363
## final value 0.119363
## stopped after 100 iterations
## # weights: 39
## initial value 82.045220
## iter 10 value 50.460103
## iter 20 value 40.359635
## iter 30 value 39.868212
## iter 40 value 39.801420
## iter 50 value 39.796008
## iter 60 value 39.777249
## iter 70 value 39.744564
## iter 80 value 39.715944
## final value 39.700617
## converged
## # weights: 109
## initial value 93.497383
## iter 10 value 23.968601
## iter 20 value 11.199831
## iter 30 value 9.462070
## iter 40 value 9.142694
## iter 50 value 9.125637
## iter 60 value 9.124723
## iter 70 value 9.124622
## iter 80 value 9.124597
## final value 9.124593
## converged
## # weights: 179
## initial value 89.102480
## iter 10 value 2.568511
## iter 20 value 0.004981
## iter 30 value 0.000125
## final value 0.000087
## converged
## # weights: 39
## initial value 90.422485
## iter 10 value 55.439180
## iter 20 value 51.210726
## iter 30 value 49.466750
## iter 40 value 47.377306
## iter 50 value 46.510887
## iter 60 value 46.510403
## iter 70 value 46.505202
## iter 80 value 45.837614
## iter 90 value 44.594650
## final value 44.551711
## converged
## # weights: 109
## initial value 84.935148
## iter 10 value 40.757184
## iter 20 value 19.629173
## iter 30 value 15.348745
```

```
## iter 40 value 12.681702
## iter 50 value 12.199866
## iter 60 value 12.185067
## iter 70 value 12.184624
## iter 80 value 12.184611
## iter 80 value 12.184611
## iter 80 value 12.184611
## final value 12.184611
## converged
## # weights: 179
## initial value 83.910426
## iter 10 value 34.998274
## iter 20 value 15.773805
## iter 30 value 12.113621
## iter 40 value 10.561321
## iter 50 value 10.240869
## iter 60 value 10.221188
## iter 70 value 10.218426
## iter 80 value 10.217704
## iter 90 value 10.217680
## final value 10.217679
## converged
## # weights: 39
## initial value 87.161684
## iter 10 value 48.919987
## iter 20 value 37.120230
## iter 30 value 33.619473
## iter 40 value 31.964313
## iter 50 value 31.846112
## iter 60 value 31.833067
## iter 70 value 31.788296
## iter 80 value 31.785319
## iter 90 value 31.771949
## iter 100 value 31.770205
## final value 31.770205
## stopped after 100 iterations
## # weights: 109
## initial value 84.472374
## iter 10 value 24.637282
## iter 20 value 6.256709
## iter 30 value 3.710129
## iter 40 value 2.301791
## iter 50 value 0.196661
## iter 60 value 0.154112
## iter 70 value 0.142138
## iter 80 value 0.124443
## iter 90 value 0.110161
## iter 100 value 0.100825
## final value 0.100825
## stopped after 100 iterations
## # weights: 179
## initial value 86.311468
## iter 10 value 19.606938
## iter 20 value 0.175800
```

```
## iter 30 value 0.140464
## iter 40 value 0.127027
## iter 50 value 0.105444
## iter 60 value 0.088970
## iter 70 value 0.081223
## iter 80 value 0.078218
## iter 90 value 0.073560
## iter 100 value 0.068932
## final value 0.068932
## stopped after 100 iterations
## # weights: 39
## initial value 95.240893
## iter 10 value 54.335887
## iter 20 value 44.230801
## iter 30 value 40.626475
## iter 40 value 40.607851
## final value 40.607827
## converged
## # weights: 109
## initial value 91.822970
## iter 10 value 33.620848
## iter 20 value 21.077220
## iter 30 value 17.886135
## iter 40 value 10.304131
## iter 50 value 5.168990
## iter 60 value 3.379538
## iter 70 value 3.028947
## iter 80 value 3.017272
## iter 90 value 3.016402
## iter 100 value 3.014590
## final value 3.014590
## stopped after 100 iterations
## # weights: 179
## initial value 78.654200
## iter 10 value 21.477523
## iter 20 value 5.268091
## iter 30 value 2.269101
## iter 40 value 0.116834
## iter 50 value 0.041768
## iter 60 value 0.002297
## iter 70 value 0.000696
## iter 80 value 0.000405
## final value 0.000078
## converged
## # weights: 39
## initial value 88.612030
## iter 10 value 49.622789
## iter 20 value 46.781365
## iter 30 value 45.469918
## iter 40 value 44.221383
## iter 50 value 43.291555
## iter 60 value 42.952887
## iter 70 value 42.777442
## iter 80 value 41.746352
```

```
## iter 90 value 41.701759
## final value 41.701757
## converged
## # weights: 109
## initial value 112.009100
## iter 10 value 39.822394
## iter 20 value 18.341515
## iter 30 value 13.739736
## iter 40 value 12.873716
## iter 50 value 12.801245
## iter 60 value 12.730344
## iter 70 value 12.373070
## iter 80 value 12.300504
## final value 12.300475
## converged
## # weights: 179
## initial value 90.899929
## iter 10 value 25.380800
## iter 20 value 11.075211
## iter 30 value 9.434821
## iter 40 value 9.309890
## iter 50 value 9.293179
## iter 60 value 9.292766
## final value 9.292761
## converged
## # weights: 39
## initial value 84.660999
## iter 10 value 49.398835
## iter 20 value 38.087474
## iter 30 value 32.891198
## iter 40 value 32.856073
## iter 50 value 32.853953
## iter 60 value 32.842141
## iter 70 value 32.841166
## iter 80 value 32.535814
## iter 90 value 26.878177
## iter 100 value 22.543408
## final value 22.543408
## stopped after 100 iterations
## # weights: 109
## initial value 81.280071
## iter 10 value 20.730092
## iter 20 value 12.601547
## iter 30 value 10.933330
## iter 40 value 10.309970
## iter 50 value 10.261066
## iter 60 value 10.249574
## iter 70 value 10.232252
## iter 80 value 10.220140
## iter 90 value 9.715515
## iter 100 value 5.224129
## final value 5.224129
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 92.642853
## iter 10 value 18.447145
## iter 20 value 1.291731
## iter 30 value 0.247518
## iter 40 value 0.176579
## iter 50 value 0.147135
## iter 60 value 0.125422
## iter 70 value 0.111623
## iter 80 value 0.098209
## iter 90 value 0.090113
## iter 100 value 0.085958
## final value 0.085958
## stopped after 100 iterations
## # weights: 39
## initial value 84.634485
## iter 10 value 60.329517
## iter 20 value 48.805937
## iter 30 value 47.747798
## iter 40 value 47.676858
## iter 50 value 47.667324
## iter 60 value 47.666017
## iter 70 value 47.665841
## final value 47.665822
## converged
## # weights: 109
## initial value 76.844182
## iter 10 value 12.473045
## iter 20 value 6.860033
## iter 30 value 6.482738
## iter 40 value 6.228858
## iter 50 value 5.926711
## iter 60 value 5.830242
## iter 70 value 5.713317
## iter 80 value 5.487873
## iter 90 value 5.035578
## iter 100 value 5.027107
## final value 5.027107
## stopped after 100 iterations
## # weights: 179
## initial value 97.227967
## iter 10 value 28.335348
## iter 20 value 13.570519
## iter 30 value 0.172348
## iter 40 value 0.008744
## iter 50 value 0.001066
## iter 60 value 0.000187
## final value 0.000086
## converged
## # weights: 39
## initial value 99.102172
## iter 10 value 62.321742
## iter 20 value 43.079288
## iter 30 value 41.755140
## iter 40 value 41.708582
```

```
## iter 50 value 41.345610
## iter 60 value 39.218102
## iter 70 value 38.196157
## iter 80 value 37.835824
## iter 90 value 37.425590
## iter 100 value 37.207597
## final value 37.207597
## stopped after 100 iterations
## # weights: 109
## initial value 79.457797
## iter 10 value 42.195002
## iter 20 value 23.916126
## iter 30 value 14.692237
## iter 40 value 13.636215
## iter 50 value 13.571933
## iter 60 value 13.570770
## final value 13.570758
## converged
## # weights: 179
## initial value 80.324008
## iter 10 value 26.593789
## iter 20 value 13.256101
## iter 30 value 10.056126
## iter 40 value 9.635744
## iter 50 value 9.599798
## iter 60 value 9.598230
## iter 70 value 9.598190
## final value 9.598190
## converged
## # weights: 39
## initial value 106.372760
## iter 10 value 50.764720
## iter 20 value 45.703467
## iter 30 value 45.118499
## iter 40 value 44.996638
## iter 50 value 44.977079
## iter 60 value 44.163350
## iter 70 value 44.138067
## iter 80 value 43.776471
## iter 90 value 42.809833
## iter 100 value 42.640595
## final value 42.640595
## stopped after 100 iterations
## # weights: 109
## initial value 94.368265
## iter 10 value 37.537841
## iter 20 value 9.237348
## iter 30 value 5.607793
## iter 40 value 3.957762
## iter 50 value 3.582007
## iter 60 value 3.457202
## iter 70 value 3.422662
## iter 80 value 3.420760
## iter 90 value 3.118261
```

```
## iter 100 value 2.963361
## final value 2.963361
## stopped after 100 iterations
## # weights: 179
## initial value 77.328495
## iter 10 value 6.997263
## iter 20 value 6.087295
## iter 30 value 2.731081
## iter 40 value 2.697898
## iter 50 value 2.625033
## iter 60 value 0.303728
## iter 70 value 0.196102
## iter 80 value 0.179382
## iter 90 value 0.170921
## iter 100 value 0.140077
## final value 0.140077
## stopped after 100 iterations
## # weights: 39
## initial value 87.726434
## iter 10 value 56.789121
## iter 20 value 51.328945
## iter 30 value 50.727551
## iter 40 value 50.565547
## iter 50 value 50.544471
## iter 60 value 50.541772
## iter 70 value 50.515839
## iter 80 value 50.493003
## final value 50.492708
## converged
## # weights: 109
## initial value 85.146082
## iter 10 value 17.247189
## iter 20 value 4.934125
## iter 30 value 1.789466
## iter 40 value 0.057120
## iter 50 value 0.001257
## iter 60 value 0.000343
## iter 70 value 0.000103
## iter 70 value 0.000091
## iter 70 value 0.000091
## final value 0.000091
## converged
## # weights: 179
## initial value 91.033007
## iter 10 value 31.708816
## iter 20 value 18.209857
## iter 30 value 13.518443
## iter
       40 value 11.694392
## iter 50 value 6.317443
## iter 60 value 6.163341
## iter 70 value 5.380551
## iter 80 value 5.030062
## iter 90 value 4.782621
## iter 100 value 4.598614
```

```
## final value 4.598614
## stopped after 100 iterations
## # weights: 39
## initial value 90.979391
## iter 10 value 63.840346
## iter 20 value 52.615424
## iter 30 value 45.795996
## iter 40 value 43.481910
## iter 50 value 43.400406
## final value 43.400337
## converged
## # weights: 109
## initial value 94.210037
## iter 10 value 34.748607
## iter 20 value 17.477748
## iter 30 value 13.690121
## iter 40 value 13.313755
## iter 50 value 13.103820
## iter 60 value 12.300829
## iter 70 value 12.274187
## final value 12.273916
## converged
## # weights: 179
## initial value 107.781479
## iter 10 value 46.113231
## iter 20 value 19.705543
## iter 30 value 13.620713
## iter 40 value 11.207289
## iter 50 value 10.508540
## iter 60 value 10.257546
## iter 70 value 10.057342
## iter 80 value 9.993388
## iter 90 value 9.980311
## iter 100 value 9.979237
## final value 9.979237
## stopped after 100 iterations
## # weights: 39
## initial value 85.637996
## iter 10 value 51.653682
## iter 20 value 47.024852
## iter 30 value 46.141526
## iter 40 value 44.813804
## iter 50 value 43.907115
## iter 60 value 43.822581
## iter 70 value 41.708839
## iter 80 value 37.321334
## iter 90 value 37.051513
## iter 100 value 36.959961
## final value 36.959961
## stopped after 100 iterations
## # weights: 109
## initial value 101.922916
## iter 10 value 40.093598
## iter 20 value 26.172665
```

```
## iter 30 value 23.959575
## iter 40 value 22.213973
## iter 50 value 21.606453
## iter 60 value 20.313021
## iter 70 value 19.728464
## iter 80 value 19.198195
## iter 90 value 19.036258
## iter 100 value 18.552648
## final value 18.552648
## stopped after 100 iterations
## # weights: 179
## initial value 88.500525
## iter 10 value 16.236081
## iter 20 value 9.359549
## iter 30 value 9.170230
## iter 40 value 9.113127
## iter 50 value 9.100655
## iter 60 value 8.683629
## iter 70 value 7.743205
## iter 80 value 5.314661
## iter 90 value 3.824127
## iter 100 value 3.724713
## final value 3.724713
## stopped after 100 iterations
## # weights: 39
## initial value 85.504158
## iter 10 value 51.035108
## iter 20 value 39.313031
## iter 30 value 37.099374
## iter 40 value 32.067651
## iter 50 value 24.718211
## iter 60 value 23.044729
## iter 70 value 22.401147
## iter 80 value 21.508237
## iter 90 value 19.292912
## iter 100 value 17.854128
## final value 17.854128
## stopped after 100 iterations
## # weights: 109
## initial value 86.368387
## iter 10 value 19.955508
## iter 20 value 3.918924
## iter 30 value 3.866134
## iter 40 value 3.862093
## final value 3.862085
## converged
## # weights: 179
## initial value 93.404603
## iter 10 value 6.995081
## iter 20 value 0.032685
## iter 30 value 0.000865
## final value 0.000054
## converged
## # weights: 39
```

```
## initial value 88.314914
## iter 10 value 55.186139
## iter 20 value 48.766242
## iter 30 value 46.880364
## iter 40 value 44.863689
## iter 50 value 44.716557
## final value 44.715521
## converged
## # weights: 109
## initial value 84.554258
## iter 10 value 42.769099
## iter 20 value 23.746488
## iter 30 value 15.237052
## iter 40 value 13.910516
## iter 50 value 13.855195
## iter 60 value 13.498744
## iter 70 value 13.340045
## final value 13.339054
## converged
## # weights: 179
## initial value 89.700336
## iter 10 value 19.369542
## iter 20 value 14.319760
## iter 30 value 11.368265
## iter 40 value 10.238950
## iter 50 value 10.019604
## iter 60 value 9.545527
## iter 70 value 9.456907
## iter 80 value 9.454461
## final value 9.454459
## converged
## # weights: 39
## initial value 83.217860
## iter 10 value 72.301523
## iter 20 value 54.547624
## iter 30 value 52.650064
## iter 40 value 48.000310
## iter 50 value 43.628003
## iter 60 value 41.195167
## iter 70 value 40.590890
## iter 80 value 40.282709
## iter 90 value 40.247224
## iter 100 value 30.270018
## final value 30.270018
## stopped after 100 iterations
## # weights: 109
## initial value 88.484848
## iter 10 value 27.234702
## iter 20 value 21.714300
## iter 30 value 21.677894
## iter 40 value 21.668467
## iter 50 value 21.650414
## iter 60 value 17.602823
## iter 70 value 15.479977
```

```
## iter 80 value 12.035298
## iter 90 value 0.558783
## iter 100 value 0.324180
## final value 0.324180
## stopped after 100 iterations
## # weights: 179
## initial value 85.158137
## iter 10 value 12.325689
## iter 20 value 5.530979
## iter 30 value 5.478463
## iter 40 value 2.125036
## iter 50 value 0.324598
## iter 60 value 0.196629
## iter 70 value 0.176477
## iter 80 value 0.156427
## iter 90 value 0.135103
## iter 100 value 0.117536
## final value 0.117536
## stopped after 100 iterations
## # weights: 39
## initial value 86.135040
## iter 10 value 58.849612
## iter 20 value 58.394005
## iter 30 value 57.215602
## iter 40 value 57.133695
## iter 50 value 56.957008
## iter 60 value 56.945747
## iter 70 value 56.203676
## iter 80 value 56.199832
## final value 56.199496
## converged
## # weights: 109
## initial value 85.263801
## iter 10 value 27.988399
## iter 20 value 10.459206
## iter 30 value 7.577708
## iter 40 value 7.554635
## iter 50 value 7.553103
## iter 60 value 7.553016
## iter 70 value 7.552982
## iter 80 value 7.552955
## final value 7.552946
## converged
## # weights: 179
## initial value 83.454380
## iter 10 value 5.220893
## iter 20 value 0.029194
## iter 30 value 0.000437
## iter 40 value 0.000113
## final value 0.000098
## converged
## # weights: 39
## initial value 85.070266
## iter 10 value 54.291042
```

```
## iter 20 value 50.846707
## iter 30 value 48.493029
## iter 40 value 43.253338
## iter 50 value 40.909423
## iter 60 value 40.789941
## final value 40.789597
## converged
## # weights: 109
## initial value 94.846875
## iter 10 value 45.152128
## iter 20 value 20.695816
## iter 30 value 13.259740
## iter 40 value 12.642599
## iter 50 value 12.606915
## iter 60 value 12.605400
## final value 12.605382
## converged
## # weights: 179
## initial value 99.763452
## iter 10 value 25.733380
## iter 20 value 11.276527
## iter 30 value 9.864735
## iter 40 value 8.510892
## iter 50 value 8.339314
## iter 60 value 8.330291
## iter 70 value 8.330149
## final value 8.330147
## converged
## # weights: 39
## initial value 80.735696
## iter 10 value 38.502479
## iter 20 value 38.032847
## iter 30 value 36.681442
## iter 40 value 33.928201
## iter 50 value 31.669386
## iter 60 value 31.330738
## iter 70 value 31.319905
## iter 80 value 31.309715
## iter 90 value 31.302847
## iter 100 value 31.300733
## final value 31.300733
## stopped after 100 iterations
## # weights: 109
## initial value 88.127854
## iter 10 value 48.603377
## iter 20 value 24.800636
## iter 30 value 21.634161
## iter
       40 value 15.351928
## iter 50 value 14.447558
## iter 60 value 12.132197
## iter 70 value 10.720194
## iter 80 value 10.443053
## iter 90 value 10.410678
## iter 100 value 10.360546
```

```
## final value 10.360546
## stopped after 100 iterations
## # weights: 179
## initial value 82.382368
## iter 10 value 3.335850
## iter 20 value 0.139071
## iter 30 value 0.126641
## iter 40 value 0.119501
## iter 50 value 0.110842
## iter 60 value 0.100903
## iter 70 value 0.092511
## iter 80 value 0.084827
## iter 90 value 0.079321
## iter 100 value 0.072681
## final value 0.072681
## stopped after 100 iterations
## # weights: 39
## initial value 88.569429
## iter 10 value 52.057311
## iter 20 value 49.489339
## iter 30 value 49.478717
## final value 49.478695
## converged
## # weights: 109
## initial value 85.658770
## iter 10 value 12.616459
## iter 20 value 9.979147
## iter 30 value 6.967466
## iter 40 value 5.472989
## iter 50 value 5.439773
## iter 60 value 5.436268
## iter 70 value 5.435758
## iter 80 value 5.435393
## iter 90 value 5.435109
## iter 100 value 5.435085
## final value 5.435085
## stopped after 100 iterations
## # weights: 179
## initial value 94.014756
## iter 10 value 9.614060
## iter 20 value 0.234099
## iter 30 value 0.022176
## iter 40 value 0.001376
## final value 0.000092
## converged
## # weights: 39
## initial value 87.228509
## iter 10 value 57.631346
## iter 20 value 51.427902
## iter 30 value 43.972267
## iter 40 value 40.559650
## iter 50 value 39.313782
## iter 60 value 39.002215
## iter 70 value 38.973746
```

```
## final value 38.973734
## converged
## # weights: 109
## initial value 84.306901
## iter 10 value 34.525259
## iter 20 value 20.977475
## iter 30 value 14.037339
## iter 40 value 13.771425
## iter 50 value 13.716870
## iter 60 value 13.671838
## iter 70 value 13.584142
## iter 80 value 13.559821
## iter 90 value 13.559421
## final value 13.559420
## converged
## # weights: 179
## initial value 92.205157
## iter 10 value 25.996815
## iter 20 value 10.571112
## iter 30 value 8.990252
## iter 40 value 8.840946
## iter 50 value 8.836100
## final value 8.836079
## converged
## # weights: 39
## initial value 96.536657
## iter 10 value 47.667920
## iter 20 value 40.664939
## iter 30 value 35.682169
## iter 40 value 32.022732
## iter 50 value 30.912800
## iter 60 value 28.485915
## iter 70 value 27.267841
## iter 80 value 24.395301
## iter 90 value 24.301131
## iter 100 value 23.949851
## final value 23.949851
## stopped after 100 iterations
## # weights: 109
## initial value 97.964712
## iter 10 value 16.029815
## iter 20 value 13.199702
## iter 30 value 8.474738
## iter 40 value 7.036124
## iter 50 value 6.135850
## iter 60 value 5.607218
## iter 70 value 5.568675
## iter 80 value 5.554491
## iter 90 value 4.605932
## iter 100 value 4.137162
## final value 4.137162
## stopped after 100 iterations
## # weights: 179
## initial value 110.375823
```

```
## iter 10 value 8.369583
## iter 20 value 5.567377
## iter 30 value 3.875420
## iter 40 value 2.057444
## iter 50 value 2.037662
## iter 60 value 2.021935
## iter 70 value 0.178508
## iter 80 value 0.114990
## iter 90 value 0.110359
## iter 100 value 0.105149
## final value 0.105149
## stopped after 100 iterations
## # weights: 39
## initial value 80.957537
## iter 10 value 50.055387
## iter 20 value 41.115415
## iter 30 value 40.857893
## final value 40.857541
## converged
## # weights: 109
## initial value 82.797846
## iter 10 value 14.267774
## iter 20 value 8.640742
## iter 30 value 8.491461
## iter 40 value 7.934376
## iter 50 value 6.920458
## iter 60 value 6.091865
## iter 70 value 5.895446
## iter 80 value 5.643293
## iter 90 value 5.091204
## iter 100 value 4.228674
## final value 4.228674
## stopped after 100 iterations
## # weights: 179
## initial value 101.835834
## iter 10 value 26.695753
## iter 20 value 0.251809
## iter 30 value 0.006019
## iter 40 value 0.001160
## iter 50 value 0.000163
## iter 50 value 0.000087
## iter 50 value 0.000087
## final value 0.000087
## converged
## # weights: 39
## initial value 89.748768
## iter 10 value 64.560955
## iter 20 value 56.987643
## iter 30 value 52.296687
## iter 40 value 48.628113
## iter 50 value 47.283428
## iter 60 value 46.072337
## iter 70 value 45.582333
## final value 45.578561
```

```
## converged
## # weights: 109
## initial value 101.878225
## iter 10 value 32.576616
## iter 20 value 17.956448
## iter 30 value 14.510480
## iter 40 value 13.695566
## iter 50 value 13.353246
## iter 60 value 13.340087
## iter 70 value 13.339678
## final value 13.339662
## converged
## # weights: 179
## initial value 104.304063
## iter 10 value 22.654255
## iter 20 value 11.808058
## iter 30 value 10.135618
## iter 40 value 9.584948
## iter 50 value 9.553848
## final value 9.553761
## converged
## # weights: 39
## initial value 89.619804
## iter 10 value 54.870380
## iter 20 value 50.080364
## iter 30 value 49.220888
## iter 40 value 46.067817
## iter 50 value 44.704431
## iter 60 value 44.500894
## iter 70 value 44.248418
## iter 80 value 43.983202
## iter 90 value 43.912536
## iter 100 value 43.897615
## final value 43.897615
## stopped after 100 iterations
## # weights: 109
## initial value 82.148497
## iter 10 value 18.504338
## iter 20 value 16.534819
## iter 30 value 12.146233
## iter 40 value 10.851083
## iter 50 value 3.146123
## iter 60 value 0.709989
## iter 70 value 0.331581
## iter 80 value 0.290400
## iter 90 value 0.274366
## iter 100 value 0.260629
## final value 0.260629
## stopped after 100 iterations
## # weights: 179
## initial value 90.879574
## iter 10 value 12.276186
## iter 20 value 0.175685
## iter 30 value 0.135455
```

```
## iter 40 value 0.118041
## iter 50 value 0.110250
## iter 60 value 0.099248
## iter 70 value 0.090586
## iter 80 value 0.086845
## iter 90 value 0.084314
## iter 100 value 0.079697
## final value 0.079697
## stopped after 100 iterations
## # weights: 39
## initial value 90.885403
## iter 10 value 50.235584
## iter 20 value 41.293811
## iter 30 value 32.253233
## iter 40 value 28.547413
## iter 50 value 28.050529
## iter 60 value 28.023540
## iter 70 value 27.246526
## iter 80 value 25.150358
## iter 90 value 24.836179
## iter 100 value 24.783391
## final value 24.783391
## stopped after 100 iterations
## # weights: 109
## initial value 90.689643
## iter 10 value 44.000867
## iter 20 value 21.982741
## iter 30 value 19.441264
## iter 40 value 18.373514
## iter 50 value 16.516993
## iter 60 value 16.281038
## iter 70 value 16.033907
## iter 80 value 15.667252
## iter 90 value 12.893773
## iter 100 value 6.852092
## final value 6.852092
## stopped after 100 iterations
## # weights: 179
## initial value 99.004364
## iter 10 value 13.336342
## iter 20 value 0.175991
## iter 30 value 0.001295
## final value 0.000090
## converged
## # weights: 39
## initial value 87.859719
## iter 10 value 55.557296
## iter 20 value 50.743608
## iter 30 value 46.335764
## iter 40 value 45.690365
## iter 50 value 45.687632
## iter 50 value 45.687632
## iter 50 value 45.687632
## final value 45.687632
```

```
## converged
## # weights: 109
## initial value 95.752011
## iter 10 value 33.296873
## iter 20 value 15.201010
## iter 30 value 13.355825
## iter 40 value 12.000015
## iter 50 value 11.969858
## final value 11.969847
## converged
## # weights: 179
## initial value 88.645177
## iter 10 value 35.025365
## iter 20 value 12.234525
## iter 30 value 10.146700
## iter 40 value 9.608180
## iter 50 value 9.576469
## iter 60 value 9.527474
## iter 70 value 9.500696
## final value 9.500285
## converged
## # weights: 39
## initial value 84.162095
## iter 10 value 45.208478
## iter 20 value 36.056182
## iter 30 value 35.668986
## iter 40 value 35.645635
## iter 50 value 35.635089
## iter 60 value 33.746947
## iter 70 value 33.132181
## iter 80 value 33.070633
## iter 90 value 30.685026
## iter 100 value 30.559358
## final value 30.559358
## stopped after 100 iterations
## # weights: 109
## initial value 98.288410
## iter 10 value 64.716515
## iter 20 value 27.328697
## iter 30 value 23.962148
## iter 40 value 16.745102
## iter 50 value 13.808241
## iter 60 value 12.895011
## iter 70 value 12.861148
## iter 80 value 12.841490
## iter 90 value 12.800089
## iter 100 value 10.496632
## final value 10.496632
## stopped after 100 iterations
## # weights: 179
## initial value 92.586932
## iter 10 value 5.577873
## iter 20 value 2.646779
## iter 30 value 1.578574
```

```
## iter 40 value 0.448454
## iter 50 value 0.287208
## iter 60 value 0.260006
## iter 70 value 0.207982
## iter 80 value 0.165087
## iter 90 value 0.153485
## iter 100 value 0.123694
## final value 0.123694
## stopped after 100 iterations
## # weights: 39
## initial value 82.285651
## iter 10 value 49.576909
## iter 20 value 43.496566
## iter 30 value 43.367276
## iter 40 value 41.968308
## iter 50 value 40.155264
## iter 60 value 40.150604
## final value 40.150354
## converged
## # weights: 109
## initial value 81.113319
## iter 10 value 15.710306
## iter 20 value 8.650494
## iter 30 value 6.481966
## iter 40 value 6.407040
## iter 50 value 6.401020
## iter 60 value 6.397249
## iter 70 value 6.394094
## iter 80 value 6.393849
## iter 90 value 6.392785
## iter 100 value 6.196127
## final value 6.196127
## stopped after 100 iterations
## # weights: 179
## initial value 74.165043
## iter 10 value 10.374266
## iter 20 value 2.205221
## iter 30 value 1.433166
## iter 40 value 1.388080
## iter 50 value 1.386627
## iter 60 value 1.386295
## final value 1.386295
## converged
## # weights: 39
## initial value 78.806613
## iter 10 value 50.157954
## iter 20 value 43.376447
## iter 30 value 42.305114
## iter 40 value 40.775328
## iter 50 value 39.388854
## iter 60 value 38.689967
## iter 70 value 38.656501
## final value 38.656137
## converged
```

```
## # weights: 109
## initial value 89.419148
## iter 10 value 41.428152
## iter 20 value 25.601646
## iter 30 value 17.834672
## iter 40 value 12.642926
## iter 50 value 11.595612
## iter 60 value 10.358117
## iter 70 value 10.321988
## iter 80 value 10.321725
## final value 10.321724
## converged
## # weights: 179
## initial value 83.130812
## iter 10 value 24.345634
## iter 20 value 10.714409
## iter 30 value 8.870532
## iter 40 value 8.533697
## iter 50 value 8.509163
## iter 60 value 8.497566
## iter 70 value 8.497112
## final value 8.497104
## converged
## # weights: 39
## initial value 82.915363
## iter 10 value 44.195114
## iter 20 value 39.187604
## iter 30 value 38.445664
## iter 40 value 38.413880
## iter 50 value 38.408266
## iter 60 value 38.391194
## iter 70 value 33.906355
## iter 80 value 33.198792
## iter 90 value 33.095255
## iter 100 value 33.006481
## final value 33.006481
## stopped after 100 iterations
## # weights: 109
## initial value 75.536522
## iter 10 value 24.378592
## iter 20 value 13.316891
## iter 30 value 11.949264
## iter 40 value 11.488736
## iter 50 value 11.146315
## iter 60 value 10.409760
## iter 70 value 10.014863
## iter 80 value 9.579057
## iter 90 value 5.923311
## iter 100 value 4.793947
## final value 4.793947
## stopped after 100 iterations
## # weights: 179
## initial value 91.459843
## iter 10 value 17.012083
```

```
## iter 20 value 0.383727
## iter 30 value 0.184435
## iter 40 value 0.145075
## iter 50 value 0.118865
## iter 60 value 0.095490
## iter 70 value 0.087759
## iter 80 value 0.083136
## iter 90 value 0.078991
## iter 100 value 0.075963
## final value 0.075963
## stopped after 100 iterations
## # weights: 39
## initial value 86.925422
## iter 10 value 75.089706
## iter 20 value 51.597660
## iter 30 value 50.322401
## iter 40 value 50.135663
## iter 50 value 49.890374
## iter 60 value 49.468568
## iter 70 value 48.244121
## iter 80 value 47.471842
## iter 90 value 45.438795
## iter 100 value 45.367848
## final value 45.367848
## stopped after 100 iterations
## # weights: 109
## initial value 80.216185
## iter 10 value 10.797088
## iter 20 value 0.492589
## iter 30 value 0.023043
## iter 40 value 0.003348
## iter 50 value 0.000462
## iter 60 value 0.000189
## final value 0.000085
## converged
## # weights: 179
## initial value 92.942133
## iter 10 value 18.558935
## iter 20 value 9.135916
## iter 30 value 7.594404
## iter 40 value 5.998985
## iter 50 value 4.516330
## iter 60 value 3.997780
## iter 70 value 3.901493
## iter 80 value 3.869751
## iter 90 value 3.838543
## iter 100 value 3.769553
## final value 3.769553
## stopped after 100 iterations
## # weights: 39
## initial value 95.099086
## iter 10 value 56.038815
## iter 20 value 46.996571
## iter 30 value 46.541976
```

```
## iter 40 value 45.899882
## iter 50 value 44.852815
## iter 60 value 44.247409
## iter 70 value 44.080407
## final value 44.079338
## converged
## # weights: 109
## initial value 114.807598
## iter 10 value 47.933184
## iter 20 value 30.095918
## iter 30 value 19.618417
## iter 40 value 14.888073
## iter 50 value 13.973251
## iter 60 value 13.450400
## iter 70 value 11.919411
## iter 80 value 11.834875
## iter 90 value 11.832884
## final value 11.832580
## converged
## # weights: 179
## initial value 103.989063
## iter 10 value 48.679498
## iter 20 value 16.529536
## iter 30 value 12.114281
## iter 40 value 11.247450
## iter 50 value 11.104977
## iter 60 value 11.098787
## iter 70 value 11.098566
## final value 11.098552
## converged
## # weights: 39
## initial value 84.167025
## iter 10 value 55.517782
## iter 20 value 52.204104
## iter 30 value 51.519064
## iter 40 value 50.648501
## iter 50 value 48.581665
## iter 60 value 48.538721
## iter 70 value 48.029915
## iter 80 value 47.688230
## iter 90 value 47.665156
## iter 100 value 47.593373
## final value 47.593373
## stopped after 100 iterations
## # weights: 109
## initial value 111.186645
## iter 10 value 23.941476
## iter 20 value 7.791402
## iter 30 value 5.995794
## iter 40 value 0.497591
## iter 50 value 0.340925
## iter 60 value 0.331554
## iter 70 value 0.320785
## iter 80 value 0.308540
```

```
## iter 90 value 0.301206
## iter 100 value 0.296787
## final value 0.296787
## stopped after 100 iterations
## # weights: 179
## initial value 79.230253
## iter 10 value 16.340692
## iter 20 value 1.058126
## iter 30 value 0.170543
## iter 40 value 0.151643
## iter 50 value 0.127515
## iter 60 value 0.112913
## iter 70 value 0.104021
## iter 80 value 0.099967
## iter 90 value 0.096261
## iter 100 value 0.091898
## final value 0.091898
## stopped after 100 iterations
## # weights: 39
## initial value 84.115611
## iter 10 value 51.888072
## iter 20 value 42.919513
## iter 30 value 40.510190
## iter 40 value 40.462647
## iter 50 value 40.461453
## final value 40.461439
## converged
## # weights: 109
## initial value 88.593368
## iter 10 value 43.543549
## iter 20 value 6.637161
## iter 30 value 0.038278
## iter 40 value 0.008543
## iter 50 value 0.001907
## iter 60 value 0.000242
## final value 0.000093
## converged
## # weights: 179
## initial value 84.414587
## iter 10 value 18.519580
## iter 20 value 3.546465
## iter 30 value 0.151060
## iter 40 value 0.002930
## iter 50 value 0.000135
## iter 50 value 0.000082
## iter 50 value 0.000082
## final value 0.000082
## converged
## # weights: 39
## initial value 90.958227
## iter 10 value 56.066105
## iter 20 value 48.316351
## iter 30 value 48.031360
## iter 40 value 46.327371
```

```
## iter 50 value 44.724163
## iter 60 value 44.620837
## final value 44.620744
## converged
## # weights: 109
## initial value 86.736202
## iter 10 value 34.612246
## iter 20 value 16.580279
## iter 30 value 13.804108
## iter 40 value 13.607868
## iter 50 value 13.260563
## iter 60 value 13.154417
## iter 70 value 13.025304
## iter 80 value 12.966463
## final value 12.966462
## converged
## # weights: 179
## initial value 101.403698
## iter 10 value 35.958371
## iter 20 value 15.177125
## iter 30 value 12.051414
## iter 40 value 10.679620
## iter 50 value 10.343655
## iter 60 value 10.304440
## iter 70 value 10.299128
## iter 80 value 10.298832
## iter 90 value 10.298806
## iter 100 value 10.298803
## final value 10.298803
## stopped after 100 iterations
## # weights: 39
## initial value 85.541371
## iter 10 value 40.462490
## iter 20 value 39.144683
## iter 30 value 38.120839
## iter 40 value 36.371696
## iter 50 value 31.631251
## iter 60 value 31.271470
## iter 70 value 31.226951
## iter 80 value 31.172060
## iter 90 value 31.108114
## iter 100 value 29.891872
## final value 29.891872
## stopped after 100 iterations
## # weights: 109
## initial value 89.328189
## iter 10 value 16.856487
## iter 20 value 9.407713
## iter 30 value 7.715865
## iter 40 value 7.699714
## iter 50 value 7.070750
## iter 60 value 3.108212
## iter 70 value 0.298989
## iter 80 value 0.142514
```

```
## iter 90 value 0.133465
## iter 100 value 0.128921
## final value 0.128921
## stopped after 100 iterations
## # weights: 179
## initial value 94.432791
## iter 10 value 19.139524
## iter 20 value 2.816546
## iter 30 value 1.594628
## iter 40 value 0.224761
## iter 50 value 0.206274
## iter 60 value 0.185097
## iter 70 value 0.155340
## iter 80 value 0.126522
## iter 90 value 0.113178
## iter 100 value 0.100636
## final value 0.100636
## stopped after 100 iterations
## # weights: 39
## initial value 95.581378
## iter 10 value 59.749052
## iter 20 value 53.598882
## iter 30 value 52.162214
## iter 40 value 50.862310
## iter 50 value 46.575284
## iter 60 value 46.371992
## iter 70 value 46.184034
## iter 80 value 45.930161
## iter 90 value 45.876062
## iter 100 value 45.867834
## final value 45.867834
## stopped after 100 iterations
## # weights: 109
## initial value 89.923919
## iter 10 value 29.004092
## iter 20 value 21.413658
## iter 30 value 19.603756
## iter 40 value 18.642277
## iter 50 value 18.152229
## iter 60 value 18.092563
## iter 70 value 18.065668
## iter 80 value 12.138861
## iter 90 value 4.764871
## iter 100 value 0.202869
## final value 0.202869
## stopped after 100 iterations
## # weights: 179
## initial value 90.315614
## iter 10 value 40.845016
## iter 20 value 17.904206
## iter 30 value 15.699832
## iter 40 value 9.098825
## iter 50 value 2.036497
## iter 60 value 0.033425
```

```
## iter 70 value 0.001418
## iter 80 value 0.000492
## final value 0.000085
## converged
## # weights: 39
## initial value 85.169125
## iter 10 value 64.972884
## iter 20 value 52.467356
## iter 30 value 46.753811
## iter 40 value 45.058500
## iter 50 value 43.553622
## iter 60 value 41.378322
## iter 70 value 40.437582
## iter 80 value 40.419734
## final value 40.419730
## converged
## # weights: 109
## initial value 89.255698
## iter 10 value 30.551451
## iter 20 value 15.049853
## iter 30 value 13.823229
## iter 40 value 12.961858
## iter 50 value 11.828680
## iter 60 value 11.727350
## iter 70 value 11.724338
## final value 11.724324
## converged
## # weights: 179
## initial value 92.643328
## iter 10 value 32.356206
## iter 20 value 12.043873
## iter 30 value 9.610067
## iter 40 value 9.171488
## iter 50 value 9.127067
## iter 60 value 9.126000
## final value 9.125989
## converged
## # weights: 39
## initial value 86.436602
## iter 10 value 51.324144
## iter 20 value 41.189983
## iter 30 value 36.271056
## iter 40 value 23.473972
## iter 50 value 19.037549
## iter 60 value 14.682916
## iter 70 value 10.087899
## iter 80 value 7.390502
## iter 90 value 7.352893
## iter 100 value 7.283568
## final value 7.283568
## stopped after 100 iterations
## # weights: 109
## initial value 92.446436
## iter 10 value 7.748643
```

```
## iter 20 value 3.646915
## iter 30 value 3.250936
## iter 40 value 3.020992
## iter 50 value 1.715864
## iter 60 value 1.639587
## iter 70 value 1.633453
## iter 80 value 1.602874
## iter 90 value 1.596140
## iter 100 value 1.583929
## final value 1.583929
## stopped after 100 iterations
## # weights: 179
## initial value 84.105863
## iter 10 value 28.459374
## iter 20 value 3.942282
## iter 30 value 2.747349
## iter 40 value 2.726469
## iter 50 value 2.700101
## iter 60 value 2.680525
## iter 70 value 1.644744
## iter 80 value 1.597143
## iter 90 value 0.338664
## iter 100 value 0.209367
## final value 0.209367
## stopped after 100 iterations
## # weights: 39
## initial value 87.885213
## iter 10 value 44.213304
## iter 20 value 42.301055
## iter 30 value 41.199692
## iter 40 value 39.422634
## iter 50 value 39.165472
## iter 60 value 39.099620
## iter 70 value 39.098662
## iter 80 value 39.098118
## iter 90 value 39.092692
## iter 100 value 38.860467
## final value 38.860467
## stopped after 100 iterations
## # weights: 109
## initial value 80.616963
## iter 10 value 24.804407
## iter 20 value 19.850656
## iter 30 value 19.548855
## iter 40 value 19.477124
## iter 50 value 15.694441
## iter 60 value 7.555785
## iter 70 value 7.320940
## iter 80 value 7.276279
## iter 90 value 4.503162
## iter 100 value 3.376580
## final value 3.376580
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 87.602008
## iter 10 value 20.302238
## iter 20 value 7.532904
## iter 30 value 3.263114
## iter 40 value 2.996990
## iter 50 value 2.321430
## iter 60 value 1.393459
## iter 70 value 1.389881
## iter 80 value 1.388395
## iter 90 value 1.384602
## iter 100 value 1.155743
## final value 1.155743
## stopped after 100 iterations
## # weights: 39
## initial value 91.305407
## iter 10 value 58.618654
## iter 20 value 53.080022
## iter 30 value 52.363335
## iter 40 value 51.065755
## iter 50 value 50.313108
## iter 60 value 49.461887
## iter 70 value 47.284153
## iter 80 value 47.221814
## final value 47.221812
## converged
## # weights: 109
## initial value 92.760705
## iter 10 value 27.576014
## iter 20 value 17.265204
## iter 30 value 13.778948
## iter 40 value 13.131824
## iter 50 value 13.130236
## final value 13.130235
## converged
## # weights: 179
## initial value 96.355526
## iter 10 value 33.826390
## iter 20 value 16.460663
## iter 30 value 10.778998
## iter 40 value 10.302176
## iter 50 value 10.172901
## iter 60 value 9.645958
## iter 70 value 9.617659
## iter 80 value 9.616849
## final value 9.616825
## converged
## # weights: 39
## initial value 87.074427
## iter 10 value 71.008343
## iter 20 value 56.164297
## iter 30 value 49.275056
## iter 40 value 47.825219
## iter 50 value 47.780336
## iter 60 value 47.751163
```

```
## iter 70 value 47.695421
## iter 80 value 47.658673
## iter 90 value 47.651104
## iter 100 value 46.269704
## final value 46.269704
## stopped after 100 iterations
## # weights: 109
## initial value 77.269200
## iter 10 value 35.617660
## iter 20 value 10.169319
## iter 30 value 5.360920
## iter 40 value 3.211887
## iter 50 value 2.393310
## iter 60 value 0.700776
## iter 70 value 0.428502
## iter 80 value 0.359364
## iter 90 value 0.339497
## iter 100 value 0.323250
## final value 0.323250
## stopped after 100 iterations
## # weights: 179
## initial value 79.680889
## iter 10 value 3.807592
## iter 20 value 0.126699
## iter 30 value 0.090285
## iter 40 value 0.083949
## iter 50 value 0.077477
## iter 60 value 0.073347
## iter 70 value 0.064331
## iter 80 value 0.058672
## iter 90 value 0.056376
## iter 100 value 0.054383
## final value 0.054383
## stopped after 100 iterations
## # weights: 39
## initial value 88.012074
## iter 10 value 71.405366
## iter 20 value 56.100726
## iter 30 value 56.076457
## iter 40 value 55.954540
## iter 50 value 55.646566
## iter 60 value 55.507321
## iter 70 value 55.458438
## iter 80 value 48.353150
## iter 90 value 46.934789
## iter 100 value 43.826322
## final value 43.826322
## stopped after 100 iterations
## # weights: 109
## initial value 81.471815
## iter 10 value 36.050582
## iter 20 value 15.036087
## iter 30 value 10.437203
## iter 40 value 0.553493
```

```
## iter 50 value 0.013375
## iter 60 value 0.003862
## iter 70 value 0.000989
## iter 80 value 0.000699
## iter 90 value 0.000588
## iter 100 value 0.000207
## final value 0.000207
## stopped after 100 iterations
## # weights: 179
## initial value 109.870317
## iter 10 value 7.851044
## iter 20 value 2.072646
## iter 30 value 1.914825
## iter 40 value 1.909850
## iter 50 value 1.909558
## final value 1.909545
## converged
## # weights: 39
## initial value 82.532556
## iter 10 value 50.471275
## iter 20 value 48.589357
## iter 30 value 46.680328
## iter 40 value 43.489034
## iter 50 value 43.372329
## final value 43.372243
## converged
## # weights: 109
## initial value 83.301827
## iter 10 value 50.239755
## iter 20 value 40.670091
## iter 30 value 31.821810
## iter 40 value 24.198455
## iter 50 value 15.692491
## iter 60 value 14.007768
## iter 70 value 13.350579
## iter 80 value 13.095092
## iter 90 value 13.067776
## final value 13.067539
## converged
## # weights: 179
## initial value 122.642259
## iter 10 value 34.125404
## iter 20 value 15.284064
## iter 30 value 11.129062
## iter 40 value 10.331692
## iter 50 value 9.949609
## iter 60 value 9.896971
## iter 70 value 9.896053
## iter 80 value 9.895976
## final value 9.895975
## converged
## # weights: 39
## initial value 84.988710
## iter 10 value 53.414109
```

```
## iter 20 value 48.781218
## iter 30 value 46.450066
## iter 40 value 45.855967
## iter 50 value 45.504685
## iter 60 value 44.001229
## iter 70 value 43.582510
## iter 80 value 43.501098
## iter 90 value 43.483127
## iter 100 value 42.489067
## final value 42.489067
## stopped after 100 iterations
## # weights: 109
## initial value 78.544943
## iter 10 value 34.269147
## iter 20 value 9.851332
## iter 30 value 7.574756
## iter 40 value 6.818164
## iter 50 value 6.345905
## iter 60 value 6.322714
## iter 70 value 6.295090
## iter 80 value 6.154941
## iter 90 value 5.619595
## iter 100 value 5.451086
## final value 5.451086
## stopped after 100 iterations
## # weights: 179
## initial value 95.661557
## iter 10 value 39.354820
## iter 20 value 12.037084
## iter 30 value 7.005599
## iter 40 value 0.186445
## iter 50 value 0.146341
## iter 60 value 0.135677
## iter 70 value 0.130774
## iter 80 value 0.115389
## iter 90 value 0.110102
## iter 100 value 0.103114
## final value 0.103114
## stopped after 100 iterations
## # weights: 39
## initial value 84.890458
## iter 10 value 56.609915
## iter 20 value 44.507153
## iter 30 value 35.914186
## iter 40 value 30.905719
## iter 50 value 26.887704
## iter 60 value 26.560696
## iter 70 value 25.306715
## iter 80 value 24.792564
## iter 90 value 24.715344
## iter 100 value 24.710941
## final value 24.710941
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 99.098165
## iter 10 value 28.786201
## iter 20 value 7.929812
## iter 30 value 0.031978
## iter 40 value 0.006280
## iter 50 value 0.002113
## final value 0.000012
## converged
## # weights: 179
## initial value 115.439256
## iter 10 value 7.128630
## iter 20 value 2.070233
## iter 30 value 1.945668
## iter 40 value 1.921497
## iter 50 value 1.678276
## iter 60 value 0.391117
## iter 70 value 0.076208
## iter 80 value 0.029519
## iter 90 value 0.010626
## iter 100 value 0.003848
## final value 0.003848
## stopped after 100 iterations
## # weights: 39
## initial value 103.705680
## iter 10 value 53.211195
## iter 20 value 47.834961
## iter 30 value 44.319229
## iter 40 value 41.892499
## iter 50 value 38.594866
## iter 60 value 37.469823
## iter 70 value 37.362836
## iter 70 value 37.362835
## iter 70 value 37.362835
## final value 37.362835
## converged
## # weights: 109
## initial value 80.405311
## iter 10 value 31.917459
## iter 20 value 16.169677
## iter 30 value 11.327703
## iter 40 value 10.814074
## iter 50 value 10.778966
## iter 60 value 10.778060
## final value 10.778056
## converged
## # weights: 179
## initial value 84.189209
## iter 10 value 26.108640
## iter 20 value 12.013497
## iter 30 value 9.373290
## iter 40 value 8.958735
## iter 50 value 8.943334
## iter 60 value 8.943022
## final value 8.943015
```

```
## converged
## # weights: 39
## initial value 84.540267
## iter 10 value 44.420054
## iter 20 value 41.305691
## iter 30 value 31.367397
## iter 40 value 31.061725
## iter 50 value 31.059840
## iter 60 value 31.056806
## iter 70 value 31.055420
## iter 80 value 31.053729
## iter 90 value 31.049046
## iter 100 value 29.998840
## final value 29.998840
## stopped after 100 iterations
## # weights: 109
## initial value 81.228345
## iter 10 value 24.272873
## iter 20 value 7.497484
## iter 30 value 6.753635
## iter 40 value 6.483648
## iter 50 value 6.352880
## iter 60 value 6.264558
## iter 70 value 4.403690
## iter 80 value 3.359493
## iter 90 value 3.333130
## iter 100 value 3.185023
## final value 3.185023
## stopped after 100 iterations
## # weights: 179
## initial value 113.126314
## iter 10 value 19.505697
## iter 20 value 8.627369
## iter 30 value 8.443156
## iter 40 value 6.717805
## iter 50 value 4.099864
## iter 60 value 0.224275
## iter 70 value 0.166034
## iter 80 value 0.153501
## iter 90 value 0.145471
## iter 100 value 0.122332
## final value 0.122332
## stopped after 100 iterations
## # weights: 39
## initial value 86.097523
## iter 10 value 49.936030
## iter 20 value 43.890027
## iter
       30 value 42.729551
## iter 40 value 42.591851
## iter 50 value 42.572274
## iter 60 value 42.569087
## iter 70 value 42.568819
## final value 42.568313
## converged
```

```
## # weights: 109
## initial value 84.320196
## iter 10 value 23.041738
## iter 20 value 7.016379
## iter 30 value 4.522576
## iter 40 value 4.498505
## iter 50 value 4.257110
## iter 60 value 4.187923
## iter 70 value 4.187898
## final value 4.187898
## converged
## # weights: 179
## initial value 119.704908
## iter 10 value 8.974781
## iter 20 value 3.231054
## iter 30 value 2.202606
## iter 40 value 0.317631
## iter 50 value 0.119630
## iter 60 value 0.017991
## iter 70 value 0.001912
## iter 80 value 0.000749
## iter 90 value 0.000415
## iter 100 value 0.000260
## final value 0.000260
## stopped after 100 iterations
## # weights: 39
## initial value 96.359105
## iter 10 value 64.894576
## iter 20 value 51.891959
## iter 30 value 48.754777
## iter 40 value 47.959488
## final value 47.954624
## converged
## # weights: 109
## initial value 97.005897
## iter 10 value 33.293174
## iter 20 value 18.525335
## iter 30 value 14.925314
## iter 40 value 14.565280
## iter 50 value 14.506521
## iter 60 value 14.483830
## iter 70 value 14.464465
## iter 80 value 14.463810
## iter 90 value 14.463780
## iter 90 value 14.463780
## iter 90 value 14.463780
## final value 14.463780
## converged
## # weights: 179
## initial value 94.403785
## iter 10 value 32.147049
## iter 20 value 17.150787
## iter 30 value 11.956276
## iter 40 value 10.328922
```

```
## iter 50 value 10.071316
## iter 60 value 10.049796
## iter 70 value 10.049173
## iter 80 value 10.049164
## iter 80 value 10.049164
## iter 80 value 10.049164
## final value 10.049164
## converged
## # weights: 39
## initial value 90.472972
## iter 10 value 48.666843
## iter 20 value 43.194239
## iter 30 value 38.995691
## iter 40 value 35.596253
## iter 50 value 35.483457
## iter 60 value 35.475603
## iter 70 value 35.466387
## iter 80 value 35.464677
## iter 90 value 35.416961
## iter 100 value 35.104182
## final value 35.104182
## stopped after 100 iterations
## # weights: 109
## initial value 89.592870
## iter 10 value 31.275267
## iter 20 value 18.922981
## iter 30 value 17.684342
## iter 40 value 16.672120
## iter 50 value 16.448323
## iter 60 value 15.837622
## iter 70 value 15.764838
## iter 80 value 13.451522
## iter 90 value 12.335129
## iter 100 value 12.325324
## final value 12.325324
## stopped after 100 iterations
## # weights: 179
## initial value 84.851448
## iter 10 value 23.438295
## iter 20 value 5.866080
## iter 30 value 4.022276
## iter 40 value 0.939609
## iter 50 value 0.441182
## iter 60 value 0.306766
## iter 70 value 0.301015
## iter 80 value 0.278531
## iter 90 value 0.258664
## iter 100 value 0.233725
## final value 0.233725
## stopped after 100 iterations
## # weights: 39
## initial value 82.020546
## iter 10 value 64.805221
## iter 20 value 41.690696
```

```
## iter 30 value 40.074506
## iter 40 value 40.072438
## final value 40.072432
## converged
## # weights: 109
## initial value 79.821808
## iter 10 value 27.937265
## iter 20 value 1.938374
## iter 30 value 0.198284
## iter 40 value 0.013956
## iter 50 value 0.004000
## iter 60 value 0.002540
## iter 70 value 0.000539
## iter 80 value 0.000220
## final value 0.000096
## converged
## # weights: 179
## initial value 85.287590
## iter 10 value 2.195586
## iter 20 value 0.019398
## iter 30 value 0.001846
## final value 0.000095
## converged
## # weights: 39
## initial value 84.317257
## iter 10 value 57.682888
## iter 20 value 50.444700
## iter 30 value 45.846096
## iter 40 value 45.692229
## final value 45.691802
## converged
## # weights: 109
## initial value 83.435358
## iter 10 value 31.643228
## iter 20 value 16.735840
## iter 30 value 13.280159
## iter 40 value 11.678191
## iter 50 value 11.580044
## iter 60 value 11.576538
## iter 70 value 11.576507
## iter 70 value 11.576507
## iter 70 value 11.576507
## final value 11.576507
## converged
## # weights: 179
## initial value 84.465331
## iter 10 value 36.219337
## iter 20 value 19.878112
## iter 30 value 11.322392
## iter 40 value 9.252890
## iter 50 value 8.599707
## iter 60 value 8.567128
## iter 70 value 8.565120
## iter 80 value 8.564941
```

```
## final value 8.564932
## converged
## # weights: 39
## initial value 89.738634
## iter 10 value 58.493085
## iter 20 value 57.203495
## iter 30 value 56.970584
## iter 40 value 56.953117
## iter 50 value 51.550252
## iter 60 value 50.715648
## iter 70 value 50.710615
## iter 80 value 50.575978
## iter 90 value 47.144963
## iter 100 value 47.131241
## final value 47.131241
## stopped after 100 iterations
## # weights: 109
## initial value 96.567744
## iter 10 value 37.915738
## iter 20 value 0.404338
## iter 30 value 0.280086
## iter 40 value 0.202904
## iter 50 value 0.157915
## iter 60 value 0.135142
## iter 70 value 0.128417
## iter 80 value 0.121374
## iter 90 value 0.112813
## iter 100 value 0.105724
## final value 0.105724
## stopped after 100 iterations
## # weights: 179
## initial value 119.659448
## iter 10 value 37.063911
## iter 20 value 17.493135
## iter 30 value 7.844024
## iter 40 value 7.159135
## iter 50 value 7.141693
## iter 60 value 7.130486
## iter 70 value 5.518796
## iter 80 value 5.502201
## iter 90 value 5.117201
## iter 100 value 0.186663
## final value 0.186663
## stopped after 100 iterations
## # weights: 179
## initial value 91.203441
## iter 10 value 27.811912
## iter 20 value 16.551713
## iter 30 value 12.649276
## iter 40 value 11.288908
## iter 50 value 10.775554
## iter 60 value 10.626294
## iter 70 value 10.618773
## iter 80 value 10.509283
```

```
## iter 90 value 10.451275
## iter 100 value 10.448875
## final value 10.448875
## stopped after 100 iterations
nnpred_model2 <- predict(neuralnet_model2, newdata= data_norm[-samp,])</pre>
#viewing confusion matrix
confusionMatrix(nnpred_model2, factor(data_norm$PAM50.mRNA[-samp]), mode = "everything")
## Confusion Matrix and Statistics
##
##
                  Reference
                   Basal.like HER2.enriched Luminal.A Luminal.B
## Prediction
    Basal.like
                            5
                                          1
##
    HER2.enriched
                            0
                                           1
                                                     0
                                                               0
    Luminal.A
                                                     6
                                                               2
##
                            0
                                          0
##
    Luminal.B
                            0
                                                     0
                                                               5
                                          1
## Overall Statistics
##
##
                  Accuracy: 0.8095
##
                    95% CI: (0.5809, 0.9455)
       No Information Rate: 0.3333
##
##
       P-Value [Acc > NIR] : 1.026e-05
##
##
                     Kappa: 0.7358
##
  Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##
                        Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                   1.0000
                                                        0.33333
                                                                          1.0000
## Specificity
                                   0.9375
                                                        1.00000
                                                                          0.8667
## Pos Pred Value
                                   0.8333
                                                        1.00000
                                                                          0.7500
## Neg Pred Value
                                   1.0000
                                                        0.90000
                                                                          1.0000
## Precision
                                   0.8333
                                                        1.00000
                                                                          0.7500
## Recall
                                   1.0000
                                                        0.33333
                                                                           1.0000
## F1
                                   0.9091
                                                        0.50000
                                                                          0.8571
## Prevalence
                                   0.2381
                                                        0.14286
                                                                          0.2857
## Detection Rate
                                   0.2381
                                                        0.04762
                                                                          0.2857
## Detection Prevalence
                                   0.2857
                                                        0.04762
                                                                          0.3810
## Balanced Accuracy
                                                        0.66667
                                                                          0.9333
                                   0.9688
##
                        Class: Luminal.B
## Sensitivity
                                  0.7143
## Specificity
                                  0.9286
## Pos Pred Value
                                  0.8333
## Neg Pred Value
                                  0.8667
## Precision
                                  0.8333
## Recall
                                  0.7143
## F1
                                  0.7692
```

0.3333

Prevalence

Model evaluation for NN

- 1. Accuracy: The overall model accuracy of Neural network model is 81%
- 2. Precision, Recall, F-1: The Precision, Recall, F1 for Basal.like class is 0.8333, 1.0000, 0.9091 The Precision, Recall, F1 for HER2.enriched class is 1.00000, 0.33333, 0.50000 The Precision, Recall, F1 for Luminal.A class is 0.7500, 1.0000, 0.8571 The Precision, Recall, F1 for Luminal.B class is 0.8333, 0.7143, 0.7692
- 3. Sensitivity and Specificity The Sensitivity and Specificity for Basal.like class is 1.0000 and 0.9375. The Sensitivity and Specificity for HER2.enriched class is 0.33333 and 1.00000 The Sensitivity and Specificity for Luminal.A is 1.0000 and 0.8667 The Sensitivity and Specificity for Luminal.B class is 0.7143 and 0.9286.
- 4. The Kappa statistic: The kappa value for this model is 0.7358 which states that it is a good agreement.

Macro-averaged Metrics: The per-class metrics can be averaged over all the classes resulting in macro-averaged precision, recall and F-1.

```
# macro-averaged precision
precision_nn <- c(0.8333,1.00000, 0.7500, 0.8333)
macro_precision_nn <- mean(precision_nn)</pre>
# macro-averaged recall
recall_nn <- c(1.0000, 0.33333, 1.0000, 0.7143)
macro_recall_nn<- mean(recall_nn)</pre>
# macro-averaged F-1
F1 nn<- c(0.9091, 0.50000, 0.8571, 0.7692)
macroF1_nn <- mean(F1_nn)</pre>
macro_average_nn <-data.frame(macro_precision_nn, macro_recall_nn, macroF1_nn)</pre>
macro_average_nn
##
     macro_precision_nn macro_recall_nn macroF1_nn
## 1
                                             0.75885
                0.85415
                               0.7619075
AUC
nn_auc <- multiclass.roc(test$PAM50.mRNA, as.ordered(nnpred_model2))</pre>
auc(nn_auc)
## Multi-class area under the curve: 0.8571
Name_metrics <- c("Accuracy", "Precision", "Recall", "F-1", "AUC", "Kappa")
values_nn <- c(0.8095 , 0.85415, 0.7619075, 0.75885,0.8571,0.7358 )</pre>
metrics_nn <- data.frame(Name_metrics, values_nn)</pre>
print (metrics_nn)
```

Naive Baye's

The Naive Bayes Algorithm is a classifier based on applying Bayes theorem with independent assumptions between features. Meaning that all features in the data set are equally important and independent of one another. Bayesian probability Is rooted in the theory that the likelihood of an event should be based on the evidence across multiple trials. Naïve Bayes uses probabilities to classify groups based on prior probability. One advantage is that Naïve Bayes works with mixed data: nominal, continuous and ordinal variables. Naïve Bayes is fast and effective, handles missing and noisy data well, and requires few records for training and can also work well with large records. Disadvantages of Naïve Bayes is that it assumes that all the data predictors are independent when in data is far from this faulty assumption. Also, estimated probabilities are less reliable than predicted classes.

```
nb_model3 <- naive_bayes(PAM50.mRNA ~ ., data = train, usekernel = T)
nbpred_model3 <- predict(nb_model3, newdata= data_norm[-samp,])</pre>
#viewing confusion matrix
confusionMatrix(nbpred_model3 , factor(data_norm$PAM50.mRNA[-samp]), mode = "everything")
## Confusion Matrix and Statistics
##
##
                  Reference
## Prediction
                    Basal.like HER2.enriched Luminal.A Luminal.B
##
                                                      0
                                                                 0
     Basal.like
                             3
                                            1
     HER2.enriched
                             1
                                            2
                                                      0
                                                                 0
##
##
     Luminal.A
                             1
                                            0
                                                      5
                                                                 1
     Luminal.B
                                            0
                                                      1
                                                                 6
##
##
##
  Overall Statistics
##
##
                  Accuracy : 0.7619
##
                     95% CI: (0.5283, 0.9178)
##
       No Information Rate: 0.3333
##
       P-Value [Acc > NIR] : 7.251e-05
##
##
                      Kappa: 0.6729
##
##
    Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                    0.6000
                                                          0.66667
                                                                             0.8333
## Specificity
                                    0.9375
                                                          0.94444
                                                                             0.8667
## Pos Pred Value
                                    0.7500
                                                          0.66667
                                                                             0.7143
## Neg Pred Value
                                    0.8824
                                                          0.94444
                                                                             0.9286
## Precision
                                    0.7500
                                                          0.66667
                                                                             0.7143
```

```
## Recall
                                     0.6000
                                                          0.66667
                                                                             0.8333
## F1
                                     0.6667
                                                          0.66667
                                                                             0.7692
                                                          0.14286
## Prevalence
                                     0.2381
                                                                             0.2857
## Detection Rate
                                                                             0.2381
                                     0.1429
                                                          0.09524
## Detection Prevalence
                                     0.1905
                                                          0.14286
                                                                             0.3333
## Balanced Accuracy
                                                          0.80556
                                    0.7688
                                                                             0.8500
                         Class: Luminal.B
## Sensitivity
                                   0.8571
## Specificity
                                   0.9286
## Pos Pred Value
                                   0.8571
## Neg Pred Value
                                   0.9286
## Precision
                                   0.8571
## Recall
                                   0.8571
                                   0.8571
## F1
## Prevalence
                                   0.3333
## Detection Rate
                                   0.2857
## Detection Prevalence
                                   0.3333
## Balanced Accuracy
                                   0.8929
```

Model evaluation for NB:

- 1. Accuracy: The overall model accuracy of Naive Baye's model is 76. 2%
- 2. Precision,Recall,F1: The Precision,Recall,F1 for Basal.like class is 0.7500, 0.60000, 0.6667 The Precision,Recall,F1 for HER2.enriched class is 0.6667, 0.6667, 0.6667 The Precision,Recall,F1 for Luminal.A class is 0.7143, 0.8333, 0.71692 The Precision,Recall,F1 for Luminal.B class is 0.8571, 0.8571, 0.8571
- 3. Sensitivity and Specificity The Sensitivity and Specificity for Basal.like class is 0.60000 and 0.9375. The Sensitivity and Specificity for HER2.enriched class is 0.66667 and 0.94444. The Sensitivity and Specificity for Luminal.A is 0.8333 and 0.8667. The Sensitivity and Specificity for Luminal.B class is 0.8571 and 0.9286.
- 4. The Kappa statistic: The kappa value for this model is 0.6729 which states that it is a good agreement.

Macro-averaged Metrics : The per-class metrics can be averaged over all the classes resulting in macro-averaged precision, recall and F-1.

```
# macro-averaged precision
precision_nb <- c(0.7500,0.66667,0.7143,0.8571)
macro_precision_nb <- mean(precision_nb)
# macro-averaged recall
recall_nb <- c(0.6000,0.66667,0.8333,0.8571)
macro_recall_nb<- mean(recall_nb)
# macro-averaged F-1
F1_nb<- c(0.6667,0.66667,0.7692,0.8571)
macroF1_nb <- mean(F1_nb)
macro_average_nb <-data.frame(macro_precision_nb, macro_recall_nb, macroF1_nb)
macro_average_nb</pre>
```

```
## macro_precision_nb macro_recall_nb macroF1_nb
## 1 0.7470175 0.7392675 0.7399175
```

AUC

```
nb_auc <- multiclass.roc(test$PAM50.mRNA, as.ordered(nbpred_model3))</pre>
auc(nb_auc)
## Multi-class area under the curve: 0.8857
Name_metrics <- c("Accuracy", "Precision", "Recall", "F-1", "AUC", "Kappa")
values_nb <- c(0.7619, 0.7470175, 0.7392675,0.7399175, 0.8857, 0.6729 )</pre>
metrics_nb <- data.frame(Name_metrics, values_nb)</pre>
print (metrics_nb)
##
     Name_metrics values_nb
## 1
         Accuracy 0.7619000
## 2
        Precision 0.7470175
## 3
           Recall 0.7392675
## 4
              F-1 0.7399175
              AUC 0.8857000
## 5
## 6
            Kappa 0.6729000
```

Random Forest

Random Forest is an ensemble of decision trees. It builds and combines multiple decision trees to get more accurate predictions. It's a non-linear classification algorithm. Each decision tree model is used when employed on its own. It works well with the multiclass classification.

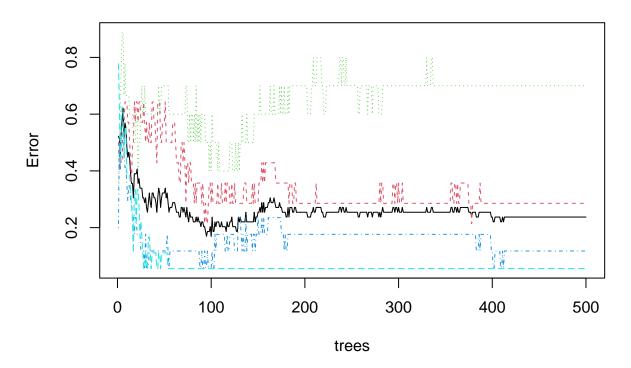
```
## Confusion Matrix and Statistics
##
##
                  Reference
## Prediction
                   Basal.like HER2.enriched Luminal.A Luminal.B
##
    Basal.like
                            4
                                          1
                                                    1
                                                              1
##
    HER2.enriched
                            0
                                          1
                                                    0
                                                               0
##
    Luminal.A
                            1
                                          0
                                                    4
                                                               2
    Luminal.B
                                          1
                                                    1
##
##
## Overall Statistics
##
##
                  Accuracy: 0.619
                    95% CI: (0.3844, 0.8189)
##
##
       No Information Rate: 0.3333
##
       P-Value [Acc > NIR] : 0.006807
##
##
                     Kappa : 0.4734
##
  Mcnemar's Test P-Value : NA
##
```

```
##
## Statistics by Class:
##
##
                        Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                    0.8000
                                                        0.33333
                                                                           0.6667
## Specificity
                                    0.8125
                                                        1.00000
                                                                           0.8000
## Pos Pred Value
                                    0.5714
                                                        1.00000
                                                                           0.5714
## Neg Pred Value
                                    0.9286
                                                        0.90000
                                                                           0.8571
## Precision
                                    0.5714
                                                        1.00000
                                                                           0.5714
## Recall
                                    0.8000
                                                        0.33333
                                                                           0.6667
## F1
                                    0.6667
                                                        0.50000
                                                                           0.6154
## Prevalence
                                    0.2381
                                                        0.14286
                                                                           0.2857
## Detection Rate
                                                        0.04762
                                    0.1905
                                                                           0.1905
## Detection Prevalence
                                    0.3333
                                                        0.04762
                                                                           0.3333
## Balanced Accuracy
                                    0.8063
                                                        0.66667
                                                                           0.7333
##
                        Class: Luminal.B
## Sensitivity
                                   0.5714
## Specificity
                                   0.8571
## Pos Pred Value
                                   0.6667
## Neg Pred Value
                                   0.8000
## Precision
                                   0.6667
## Recall
                                   0.5714
## F1
                                   0.6154
## Prevalence
                                   0.3333
## Detection Rate
                                   0.1905
## Detection Prevalence
                                   0.2857
## Balanced Accuracy
                                   0.7143
```

Plotting model

plot(classifier_RF)

classifier_RF



Importance plot

importance(classifier_RF)

```
MeanDecreaseGini
##
## NP_038479
                        0.8805231
## NP_001258898
                        1.0524888
## NP_542785
                        1.8547260
## NP_002005
                        1.8352126
## XP_003846537
                        1.6043369
  NP_660208
                        1.3260767
  NP_004439
                        0.5102471
                        1.0973934
## NP_004388
## NP_001229372
                        0.5588631
## NP_005130
                        1.2257370
## NP_004243
                        3.3422886
## NP_003144
                        2.6735076
## NP_001171551
                        3.1100743
## NP_072096
                        0.7202323
## NP_005484
                        1.6973190
## NP_061170
                        1.1520539
## NP_653304
                        1.9111497
## NP_653081
                        0.4687590
## NP_057164
                        0.9427079
## NP_004453
                        2.4038736
## NP_003212
                        1.6859924
## NP_060866
                        1.4245630
```

```
## NP_001160163
                        1.2619327
## NP_003875
                        1.4044818
## NP 002677
                        0.4121577
## NP_001135891
                        0.8641273
## NP_004522
                        1.6539665
## NP 003767
                        1.2076640
## NP 848597
                        1.1409587
## NP_005970
                        1.4146192
```

```
# Variable importance plot
varImpPlot(classifier_RF)
```

classifier_RF



Model evaluation for RF: 1. Accuracy : The overall model accuracy of model is 61.9% 2. Precision,Recall,F1: The Precision,Recall,F1 for Basal.like class is 0.5714 , 0.8000, 0.6667 The Precision,Recall,F1 for HER2.enriched class is 1.00000, 0.33333, 0.50000 The Precision,Recall,F1 for Luminal.A class is 0.5714, 0.6667, 0.6154 The Precision,Recall,F1 for Luminal.B class is 0.6667, 0.5714, 0.6154

3. Sensitivity and Specificity The Sensitivity and Specificity for Basal.like class is 0.80000 and 0.8125 The Sensitivity and Specificity for HER2.enriched class is 0.333333 and 1.00000 The Sensitivity and Specificity for Luminal.A is 0.6667 and 0.8000 The Sensitivity and Specificity for Luminal.B class is 0.5714 and 0.8571 4. The Kappa statistic: The kappa value for this model is 0.473 which states that it is a good agreement.

Macro-averaged Metrics: The per-class metrics can be averaged over all the classes resulting in macro-averaged precision, recall and F-1.

```
# macro-averaged precision
precision_rf<- c(0.5714,1.00000, 0.5714, 0.6667)
macro_precision_rf <- mean(precision_rf)</pre>
```

```
# macro-averaged recall
recall_rf <- c(0.8000, 0.33333,0.6667,0.5714)
macro_recall_rf<- mean(recall_rf)</pre>
# macro-averaged F-1
F1_rf<- c(0.6667, 0.50000, 0.6154,0.6154)
macroF1_rf <- mean(F1_rf)</pre>
macro_average_rf <-data.frame(macro_precision_rf, macro_recall_rf, macroF1_rf)</pre>
macro_average_rf
##
     macro_precision_rf macro_recall_rf macroF1_rf
## 1
               0.702375
                               0.5928575
                                           0.599375
AUC
rf_auc <- multiclass.roc(test$PAM50.mRNA, as.ordered(RF_pred))</pre>
auc(rf_auc)
## Multi-class area under the curve: 0.7388
Name_metrics <- c("Accuracy", "Precision", "Recall", "F-1", "AUC", "Kappa")
values_rf <- c(0.619, 0.702375, 0.5928575, 0.599375, 0.7388, 0.4734)
metrics_rf <- data.frame(Name_metrics, values_rf)</pre>
print (metrics_rf)
##
     Name_metrics values_rf
        Accuracy 0.6190000
## 1
## 2
        Precision 0.7023750
## 3
           Recall 0.5928575
## 4
              F-1 0.5993750
## 5
              AUC 0.7388000
## 6
            Kappa 0.4734000
```

Comparing models

By comparing accuracy, precision, recall, f-1, AUC and kappa, Both SVM and the Neural Network model works best for the given datset have approximate same accuracy, precision, recall, f-1, AUC and kappa

```
#SVM
print(metrics_svm)
```

```
## Name_metrics values_svm
## 1 Accuracy 0.8095000
## 2 Precision 0.8428500
## 3 Recall 0.8095175
## 4 F-1 0.8057750
## 5 AUC 0.7897000
## 6 Kappa 0.7399000
```

```
print(metrics_nn)
##
     Name_metrics values_nn
## 1
         Accuracy 0.8095000
## 2
        Precision 0.8541500
           Recall 0.7619075
## 3
## 4
              F-1 0.7588500
## 5
              AUC 0.8571000
## 6
            Kappa 0.7358000
#NB
print(metrics_nb)
     Name_metrics values_nb
##
## 1
         Accuracy 0.7619000
## 2
        Precision 0.7470175
## 3
           Recall 0.7392675
## 4
              F-1 0.7399175
              AUC 0.8857000
## 5
## 6
            Kappa 0.6729000
#RF
print(metrics_rf)
##
     Name_metrics values_rf
## 1
         Accuracy 0.6190000
## 2
        Precision 0.7023750
## 3
           Recall 0.5928575
              F-1 0.5993750
## 4
## 5
              AUC 0.7388000
## 6
            Kappa 0.4734000
```

Though the SVM and neural network model achieved an accuracy of 81%, the precision, recall, AUC, kappa, and f-1 scores of neural network model are higher than SVM

Evaluation with k-fold cross-validation

- k-Fold Cross Validation is done for the whole dataset
- I have used k = 10 which means 10 folds take place along with 10 repetitions
- For testing the data, I have used 3 models to test the k-fold CV
- Accuracy of each model is printed and based on the observation average accuracy is around 75-80%

K-fold Cross Validation

#Creating a train function for cross validation #We use k=10 folds with repeated validation =10

```
fitControl <- trainControl(## 10-fold CV</pre>
                           method = "repeatedcv",
                           number = 10,repeats = 10,savePredictions = TRUE, summaryFunction = multiClas
## SVM
svm_fit <- train(PAM50.mRNA ~ ., data=data_norm, trControl= fitControl, method="svmLinear")</pre>
nn_fit <- train(PAM50.mRNA ~., data= data_norm,</pre>
               method = "nnet",
               trControl = fitControl)
## # weights: 39
## initial value 104.299245
## iter 10 value 79.995263
## iter 20 value 68.794379
## iter 30 value 60.381853
## iter 40 value 59.388098
## iter 50 value 59.379352
## final value 59.379337
## converged
## # weights: 109
## initial value 116.362540
## iter 10 value 38.703189
## iter 20 value 27.949090
## iter 30 value 17.701280
## iter 40 value 14.439066
## iter 50 value 12.383683
## iter 60 value 6.794370
## iter 70 value 4.342562
## iter 80 value 4.115155
## iter 90 value 4.059620
## iter 100 value 4.025623
## final value 4.025623
## stopped after 100 iterations
## # weights: 179
## initial value 113.501953
## iter 10 value 25.731497
## iter 20 value 1.147697
## iter 30 value 0.007144
## iter 40 value 0.000410
## final value 0.000094
## converged
## # weights: 39
## initial value 106.824209
## iter 10 value 83.502515
## iter 20 value 68.645068
## iter 30 value 65.457292
## iter 40 value 61.513073
## iter 50 value 58.252483
## iter 60 value 57.896911
## final value 57.894931
## converged
## # weights: 109
```

initial value 116.266962

```
## iter 10 value 51.596366
## iter 20 value 36.012333
## iter 30 value 21.950493
## iter 40 value 17.872911
## iter 50 value 17.036263
## iter 60 value 16.922137
## iter 70 value 16.914098
## final value 16.913988
## converged
## # weights: 179
## initial value 122.764369
## iter 10 value 38.320919
## iter 20 value 20.761799
## iter 30 value 14.930962
## iter 40 value 13.658736
## iter 50 value 12.792821
## iter 60 value 12.222951
## iter 70 value 12.160783
## iter 80 value 12.155308
## iter 90 value 12.155177
## final value 12.155175
## converged
## # weights: 39
## initial value 102.999903
## iter 10 value 71.735883
## iter 20 value 70.410970
## iter 30 value 70.294498
## iter 40 value 67.538440
## iter 50 value 66.281131
## iter 60 value 66.192468
## iter 70 value 66.108589
## iter 80 value 63.075431
## iter 90 value 61.652379
## iter 100 value 60.552969
## final value 60.552969
## stopped after 100 iterations
## # weights: 109
## initial value 99.973885
## iter 10 value 46.734583
## iter 20 value 31.972769
## iter 30 value 30.881697
## iter 40 value 28.947184
## iter 50 value 27.868078
## iter 60 value 25.366362
## iter 70 value 22.809570
## iter 80 value 22.590384
## iter 90 value 19.453428
## iter 100 value 17.415759
## final value 17.415759
## stopped after 100 iterations
## # weights: 179
## initial value 124.898739
## iter 10 value 26.580909
## iter 20 value 12.932771
```

```
## iter 30 value 8.485489
## iter 40 value 5.668525
## iter 50 value 2.874250
## iter 60 value 2.468289
## iter 70 value 2.187638
## iter 80 value 2.171960
## iter 90 value 2.141243
## iter 100 value 1.600863
## final value 1.600863
## stopped after 100 iterations
## # weights: 39
## initial value 107.534552
## iter 10 value 77.408126
## iter 20 value 62.583232
## iter 30 value 62.127657
## iter 40 value 61.841436
## iter 50 value 61.220112
## iter 60 value 59.870141
## iter 70 value 57.542284
## iter 80 value 57.195943
## iter 90 value 57.091503
## iter 100 value 56.912937
## final value 56.912937
## stopped after 100 iterations
## # weights: 109
## initial value 95.978593
## iter 10 value 27.730307
## iter 20 value 11.535128
## iter 30 value 9.429867
## iter 40 value 6.068945
## iter 50 value 2.907960
## iter 60 value 0.572396
## iter 70 value 0.260605
## iter 80 value 0.129128
## iter 90 value 0.055157
## iter 100 value 0.018479
## final value 0.018479
## stopped after 100 iterations
## # weights: 179
## initial value 110.151479
## iter 10 value 29.390762
## iter 20 value 12.489817
## iter 30 value 11.199421
## iter 40 value 11.006589
## iter 50 value 9.140846
## iter 60 value 8.864893
## iter 70 value 8.427249
## iter 80 value 8.014101
## iter 90 value 7.715628
## iter 100 value 7.291841
## final value 7.291841
## stopped after 100 iterations
## # weights: 39
## initial value 99.289685
```

```
## iter 10 value 74.506952
## iter 20 value 60.928284
## iter 30 value 56.968706
## iter 40 value 54.908474
## iter 50 value 54.781351
## final value 54.781138
## converged
## # weights: 109
## initial value 137.357072
## iter 10 value 46.207196
## iter 20 value 25.596747
## iter 30 value 17.876964
## iter 40 value 15.953225
## iter 50 value 15.860954
## iter 60 value 15.754487
## iter 70 value 15.742685
## final value 15.742600
## converged
## # weights: 179
## initial value 113.028811
## iter 10 value 51.580245
## iter 20 value 18.290789
## iter 30 value 13.097499
## iter 40 value 12.212833
## iter 50 value 12.004191
## iter 60 value 11.909150
## iter 70 value 11.650546
## iter 80 value 11.634106
## iter 90 value 11.633821
## final value 11.633818
## converged
## # weights: 39
## initial value 115.982494
## iter 10 value 67.485232
## iter 20 value 52.358131
## iter 30 value 49.659574
## iter 40 value 47.370464
## iter 50 value 46.753621
## iter 60 value 46.544948
## iter 70 value 46.539404
## iter 80 value 46.497780
## iter 90 value 46.492026
## iter 100 value 45.739199
## final value 45.739199
## stopped after 100 iterations
## # weights: 109
## initial value 130.792341
## iter 10 value 55.816138
## iter 20 value 31.748520
## iter 30 value 29.668303
## iter 40 value 26.709483
## iter 50 value 24.513221
## iter 60 value 22.782926
## iter 70 value 19.023683
```

```
## iter 80 value 16.658544
## iter 90 value 15.870652
## iter 100 value 14.728344
## final value 14.728344
## stopped after 100 iterations
## # weights: 179
## initial value 104.611807
## iter 10 value 22.178543
## iter 20 value 12.448488
## iter 30 value 10.927914
## iter 40 value 9.463426
## iter 50 value 7.778376
## iter 60 value 7.736431
## iter 70 value 7.716090
## iter 80 value 7.708813
## iter 90 value 7.688731
## iter 100 value 5.733465
## final value 5.733465
## stopped after 100 iterations
## # weights: 39
## initial value 106.172496
## iter 10 value 86.872725
## iter 20 value 57.511076
## iter 30 value 56.540269
## iter 40 value 56.439194
## iter 50 value 56.396876
## iter 60 value 56.391441
## iter 70 value 56.391033
## iter 80 value 56.385723
## final value 56.385444
## converged
## # weights: 109
## initial value 104.772786
## iter 10 value 55.672629
## iter 20 value 25.962566
## iter 30 value 21.773805
## iter 40 value 18.892384
## iter 50 value 18.247111
## iter 60 value 17.458875
## iter 70 value 16.139796
## iter 80 value 15.787836
## iter 90 value 15.538131
## iter 100 value 15.277635
## final value 15.277635
## stopped after 100 iterations
## # weights: 179
## initial value 111.814271
## iter 10 value 44.354336
## iter 20 value 10.307670
## iter 30 value 6.086850
## iter 40 value 5.348093
## iter 50 value 3.497817
## iter 60 value 2.820233
## iter 70 value 2.794453
```

```
## iter 80 value 2.780624
## iter 90 value 2.750278
## iter 100 value 1.436444
## final value 1.436444
## stopped after 100 iterations
## # weights: 39
## initial value 104.974226
## iter 10 value 80.474975
## iter 20 value 71.828073
## iter 30 value 63.278804
## iter 40 value 61.888456
## iter 50 value 59.784916
## iter 60 value 59.612207
## iter 70 value 59.417354
## iter 80 value 58.905914
## iter 90 value 58.677707
## iter 100 value 58.370532
## final value 58.370532
## stopped after 100 iterations
## # weights: 109
## initial value 106.528293
## iter 10 value 53.734265
## iter 20 value 30.959814
## iter 30 value 23.095271
## iter 40 value 18.686541
## iter 50 value 17.210192
## iter 60 value 16.941136
## iter 70 value 16.866421
## iter 80 value 16.860315
## iter 90 value 16.860243
## iter 100 value 16.860240
## final value 16.860240
## stopped after 100 iterations
## # weights: 179
## initial value 106.609535
## iter 10 value 63.899599
## iter 20 value 25.022424
## iter 30 value 15.515251
## iter 40 value 13.810678
## iter 50 value 12.793296
## iter 60 value 12.484983
## iter 70 value 12.334758
## iter 80 value 12.236790
## iter 90 value 12.231608
## iter 100 value 12.231195
## final value 12.231195
## stopped after 100 iterations
## # weights: 39
## initial value 101.251697
## iter 10 value 63.047555
## iter 20 value 57.734503
## iter 30 value 56.939382
## iter 40 value 55.426947
## iter 50 value 54.959000
```

```
## iter 60 value 54.590336
## iter 70 value 54.019167
## iter 80 value 54.010700
## iter 90 value 53.827956
## iter 100 value 53.528454
## final value 53.528454
## stopped after 100 iterations
## # weights: 109
## initial value 108.635563
## iter 10 value 48.039839
## iter 20 value 37.102114
## iter 30 value 29.573146
## iter 40 value 26.880421
## iter 50 value 23.348759
## iter 60 value 18.988447
## iter 70 value 16.018596
## iter 80 value 13.116881
## iter 90 value 9.794588
## iter 100 value 7.008114
## final value 7.008114
## stopped after 100 iterations
## # weights: 179
## initial value 100.607660
## iter 10 value 27.004845
## iter 20 value 7.453921
## iter 30 value 5.302572
## iter 40 value 4.961915
## iter 50 value 4.927429
## iter 60 value 4.364980
## iter 70 value 4.289189
## iter 80 value 4.283562
## iter 90 value 4.273091
## iter 100 value 3.210138
## final value 3.210138
## stopped after 100 iterations
## # weights: 39
## initial value 100.717327
## iter 10 value 64.985780
## iter 20 value 58.567277
## iter 30 value 55.655203
## iter 40 value 53.682472
## iter 50 value 51.761930
## iter 60 value 50.823441
## iter 70 value 49.179080
## iter 80 value 48.890284
## iter 90 value 48.568349
## iter 100 value 48.209384
## final value 48.209384
## stopped after 100 iterations
## # weights: 109
## initial value 113.829643
## iter 10 value 50.596999
## iter 20 value 23.609407
## iter 30 value 18.619909
```

```
## iter 40 value 17.125002
## iter 50 value 15.399086
## iter 60 value 14.399728
## iter 70 value 6.490757
## iter 80 value 3.972695
## iter 90 value 3.905719
## iter 100 value 3.870901
## final value 3.870901
## stopped after 100 iterations
## # weights: 179
## initial value 104.998431
## iter 10 value 22.333531
## iter 20 value 13.811587
## iter 30 value 9.801557
## iter 40 value 8.154785
## iter 50 value 7.284657
## iter 60 value 5.563999
## iter 70 value 4.660962
## iter 80 value 3.850569
## iter 90 value 3.826762
## iter 100 value 3.822120
## final value 3.822120
## stopped after 100 iterations
## # weights: 39
## initial value 105.049947
## iter 10 value 81.858467
## iter 20 value 73.393758
## iter 30 value 65.001082
## iter 40 value 60.848525
## iter 50 value 59.022501
## iter 60 value 58.300252
## iter 70 value 57.186428
## iter 80 value 56.598648
## iter 90 value 56.595445
## iter 90 value 56.595444
## iter 90 value 56.595444
## final value 56.595444
## converged
## # weights: 109
## initial value 107.200565
## iter 10 value 67.677230
## iter 20 value 36.521978
## iter 30 value 22.392395
## iter 40 value 17.668750
## iter 50 value 17.246634
## iter 60 value 17.121434
## iter 70 value 17.061321
## iter 80 value 17.009409
## iter 90 value 16.442363
## iter 100 value 16.308005
## final value 16.308005
## stopped after 100 iterations
## # weights: 179
## initial value 152.069661
```

```
## iter 10 value 61.784433
## iter 20 value 27.333358
## iter 30 value 17.521918
## iter 40 value 14.049912
## iter 50 value 12.613819
## iter 60 value 12.224530
## iter 70 value 12.202828
## iter 80 value 12.202575
## final value 12.202574
## converged
## # weights: 39
## initial value 100.626011
## iter 10 value 65.509069
## iter 20 value 58.342048
## iter 30 value 54.963527
## iter 40 value 54.552171
## iter 50 value 54.531254
## iter 60 value 54.501058
## iter 70 value 54.491701
## iter 80 value 54.471611
## iter 90 value 54.470959
## iter 100 value 54.464181
## final value 54.464181
## stopped after 100 iterations
## # weights: 109
## initial value 116.438720
## iter 10 value 26.557513
## iter 20 value 15.497749
## iter 30 value 10.841419
## iter 40 value 8.225107
## iter 50 value 7.381428
## iter 60 value 7.358426
## iter 70 value 7.346893
## iter 80 value 7.343231
## iter 90 value 7.337401
## iter 100 value 7.282426
## final value 7.282426
## stopped after 100 iterations
## # weights: 179
## initial value 110.390245
## iter 10 value 21.362538
## iter 20 value 4.026865
## iter 30 value 1.539409
## iter 40 value 0.173331
## iter 50 value 0.153800
## iter 60 value 0.136163
## iter 70 value 0.127874
## iter 80 value 0.120223
## iter 90 value 0.112015
## iter 100 value 0.106384
## final value 0.106384
## stopped after 100 iterations
## # weights: 39
## initial value 98.552367
```

```
## iter 10 value 67.855438
## iter 20 value 63.206718
## iter 30 value 62.068004
## iter 40 value 60.066793
## iter 50 value 58.205317
## iter 60 value 56.389512
## iter 70 value 54.260892
## iter 80 value 54.025613
## iter 90 value 53.933159
## iter 100 value 53.914166
## final value 53.914166
## stopped after 100 iterations
## # weights: 109
## initial value 107.177071
## iter 10 value 51.032188
## iter 20 value 29.849584
## iter 30 value 17.488955
## iter 40 value 16.283049
## iter 50 value 14.894966
## iter 60 value 14.787368
## iter 70 value 14.758977
## iter 80 value 14.753727
## iter 90 value 14.752571
## iter 100 value 14.750209
## final value 14.750209
## stopped after 100 iterations
## # weights: 179
## initial value 110.349521
## iter 10 value 24.671307
## iter 20 value 9.924925
## iter 30 value 5.793526
## iter 40 value 3.025711
## iter 50 value 2.306541
## iter 60 value 1.593979
## iter 70 value 1.384801
## iter 80 value 0.026944
## iter 90 value 0.004425
## iter 100 value 0.000988
## final value 0.000988
## stopped after 100 iterations
## # weights: 39
## initial value 117.025048
## iter 10 value 69.939157
## iter 20 value 62.618639
## iter 30 value 60.193304
## iter 40 value 56.433804
## iter 50 value 55.966883
## iter 60 value 55.787166
## iter 70 value 55.643447
## final value 55.641012
## converged
## # weights: 109
## initial value 107.239992
## iter 10 value 60.210263
```

```
## iter 20 value 33.247721
## iter 30 value 20.005296
## iter 40 value 16.829320
## iter 50 value 16.611187
## iter 60 value 16.585789
## iter 70 value 16.576311
## iter 80 value 16.575989
## iter 90 value 16.575887
## iter 100 value 16.570244
## final value 16.570244
## stopped after 100 iterations
## # weights: 179
## initial value 98.070342
## iter 10 value 40.086104
## iter 20 value 20.348670
## iter 30 value 14.270639
## iter 40 value 12.530497
## iter 50 value 12.316868
## iter 60 value 12.177518
## iter 70 value 12.115303
## iter 80 value 12.104557
## final value 12.104504
## converged
## # weights: 39
## initial value 100.912213
## iter 10 value 70.384471
## iter 20 value 52.876164
## iter 30 value 46.449586
## iter 40 value 43.333689
## iter 50 value 41.598764
## iter 60 value 40.001864
## iter 70 value 38.421120
## iter 80 value 37.921159
## iter 90 value 36.138026
## iter 100 value 34.847282
## final value 34.847282
## stopped after 100 iterations
## # weights: 109
## initial value 112.716553
## iter 10 value 52.125195
## iter 20 value 39.553309
## iter 30 value 37.208279
## iter 40 value 36.981957
## iter 50 value 36.923396
## iter 60 value 32.555804
## iter 70 value 32.026935
## iter 80 value 30.925919
## iter 90 value 30.219515
## iter 100 value 29.505104
## final value 29.505104
## stopped after 100 iterations
## # weights: 179
## initial value 118.677646
## iter 10 value 28.638221
```

```
## iter 20 value 15.491842
## iter 30 value 13.136416
## iter 40 value 12.318244
## iter 50 value 12.228116
## iter 60 value 12.050837
## iter 70 value 11.857411
## iter 80 value 11.816403
## iter 90 value 11.801960
## iter 100 value 11.759114
## final value 11.759114
## stopped after 100 iterations
## # weights: 39
## initial value 99.880365
## iter 10 value 57.961515
## iter 20 value 56.762476
## iter 30 value 56.721989
## final value 56.721929
## converged
## # weights: 109
## initial value 96.754578
## iter 10 value 34.799439
## iter 20 value 21.553847
## iter 30 value 18.274166
## iter 40 value 16.807931
## iter 50 value 16.266339
## iter 60 value 15.969466
## iter 70 value 15.861306
## iter 80 value 15.832816
## iter 90 value 15.760522
## iter 100 value 14.929405
## final value 14.929405
## stopped after 100 iterations
## # weights: 179
## initial value 103.676180
## iter 10 value 33.171005
## iter 20 value 13.314321
## iter 30 value 7.562079
## iter 40 value 5.631875
## iter 50 value 5.124174
## iter 60 value 5.056457
## iter 70 value 5.030734
## iter 80 value 5.023789
## iter 90 value 4.175961
## iter 100 value 4.164328
## final value 4.164328
## stopped after 100 iterations
## # weights: 39
## initial value 109.231542
## iter 10 value 81.919137
## iter 20 value 71.954902
## iter 30 value 61.252902
## iter 40 value 59.000512
## iter 50 value 58.645676
## iter 60 value 58.209515
```

```
## iter 70 value 57.945933
## iter 80 value 57.894637
## iter 90 value 57.824625
## iter 100 value 57.793450
## final value 57.793450
## stopped after 100 iterations
## # weights: 109
## initial value 107.791837
## iter 10 value 53.681348
## iter 20 value 36.243093
## iter 30 value 28.404314
## iter 40 value 22.999579
## iter 50 value 19.119749
## iter 60 value 17.219522
## iter 70 value 16.420321
## iter 80 value 16.287253
## iter 90 value 16.258708
## iter 100 value 16.248457
## final value 16.248457
## stopped after 100 iterations
## # weights: 179
## initial value 115.324174
## iter 10 value 40.915611
## iter 20 value 24.398844
## iter 30 value 16.096842
## iter 40 value 13.897308
## iter 50 value 13.622497
## iter 60 value 12.734113
## iter 70 value 12.536133
## iter 80 value 12.485224
## iter 90 value 12.460506
## iter 100 value 12.459113
## final value 12.459113
## stopped after 100 iterations
## # weights: 39
## initial value 101.757311
## iter 10 value 68.407113
## iter 20 value 63.642736
## iter 30 value 62.402660
## iter 40 value 55.957965
## iter 50 value 53.515893
## iter 60 value 51.354498
## iter 70 value 50.607363
## iter 80 value 49.028262
## iter 90 value 48.682382
## iter 100 value 47.838888
## final value 47.838888
## stopped after 100 iterations
## # weights: 109
## initial value 103.194169
## iter 10 value 52.745907
## iter 20 value 41.613003
## iter 30 value 39.727312
## iter 40 value 37.670857
```

```
## iter 50 value 32.824262
## iter 60 value 29.672714
## iter 70 value 28.608191
## iter 80 value 27.761329
## iter 90 value 27.553337
## iter 100 value 26.904029
## final value 26.904029
## stopped after 100 iterations
## # weights: 179
## initial value 110.863530
## iter 10 value 27.436186
## iter 20 value 14.791227
## iter 30 value 10.701541
## iter 40 value 9.301095
## iter 50 value 8.630477
## iter 60 value 7.591230
## iter 70 value 4.815392
## iter 80 value 4.761449
## iter 90 value 4.754500
## iter 100 value 4.738265
## final value 4.738265
## stopped after 100 iterations
## # weights: 39
## initial value 110.959131
## iter 10 value 63.465868
## iter 20 value 58.488543
## iter 30 value 57.203294
## iter 40 value 57.110512
## iter 50 value 57.007346
## iter 60 value 56.496470
## iter 70 value 56.462252
## iter 80 value 56.443633
## iter 90 value 56.390091
## iter 100 value 52.483490
## final value 52.483490
## stopped after 100 iterations
## # weights: 109
## initial value 108.824751
## iter 10 value 47.460205
## iter 20 value 26.905718
## iter 30 value 20.341997
## iter 40 value 16.560114
## iter 50 value 15.997722
## iter 60 value 15.915869
## iter 70 value 15.900989
## iter 80 value 15.893431
## iter 90 value 15.884524
## iter 100 value 15.883344
## final value 15.883344
## stopped after 100 iterations
## # weights: 179
## initial value 111.206377
## iter 10 value 25.127204
## iter 20 value 7.266960
```

```
## iter 30 value 4.714106
## iter 40 value 1.481820
## iter 50 value 1.406999
## iter 60 value 1.393125
## iter 70 value 1.389521
## iter 80 value 1.387649
## iter 90 value 1.386772
## iter 100 value 1.386590
## final value 1.386590
## stopped after 100 iterations
## # weights: 39
## initial value 100.917487
## iter 10 value 71.981814
## iter 20 value 60.175294
## iter 30 value 58.317043
## iter 40 value 57.186037
## iter 50 value 56.272042
## iter 60 value 55.872294
## iter 70 value 55.864496
## iter 80 value 55.862892
## iter 80 value 55.862892
## final value 55.862892
## converged
## # weights: 109
## initial value 104.944030
## iter 10 value 50.983381
## iter 20 value 31.509779
## iter 30 value 24.463671
## iter 40 value 18.705393
## iter 50 value 17.301566
## iter 60 value 16.273576
## iter 70 value 16.074725
## iter 80 value 15.446332
## iter 90 value 15.324312
## iter 100 value 15.322977
## final value 15.322977
## stopped after 100 iterations
## # weights: 179
## initial value 119.735846
## iter 10 value 58.348497
## iter 20 value 26.209849
## iter 30 value 15.392135
## iter 40 value 13.540317
## iter 50 value 13.085016
## iter 60 value 12.674765
        70 value 12.249233
## iter
## iter 80 value 12.138040
## iter 90 value 12.062072
## iter 100 value 12.012855
## final value 12.012855
## stopped after 100 iterations
## # weights: 39
## initial value 106.136459
## iter 10 value 65.889177
```

```
## iter 20 value 57.929137
## iter 30 value 56.948422
## iter 40 value 56.669705
## iter 50 value 56.637367
## iter 60 value 56.634346
## iter 70 value 55.888021
## iter 80 value 54.384295
## iter 90 value 51.968513
## iter 100 value 51.953473
## final value 51.953473
## stopped after 100 iterations
## # weights: 109
## initial value 102.682068
## iter 10 value 27.112006
## iter 20 value 9.241030
## iter 30 value 5.248934
## iter 40 value 3.790089
## iter 50 value 3.768558
## iter 60 value 3.743937
## iter 70 value 3.664264
## iter 80 value 3.643307
## iter 90 value 3.630765
## iter 100 value 3.623009
## final value 3.623009
## stopped after 100 iterations
## # weights: 179
## initial value 132.329856
## iter 10 value 40.464593
## iter 20 value 7.703621
## iter 30 value 0.637454
## iter 40 value 0.282670
## iter 50 value 0.249801
## iter 60 value 0.235429
## iter 70 value 0.213995
## iter 80 value 0.199750
## iter 90 value 0.186244
## iter 100 value 0.170999
## final value 0.170999
## stopped after 100 iterations
## # weights: 39
## initial value 107.932002
## iter 10 value 82.114107
## iter 20 value 71.940654
## iter 30 value 64.243677
## iter 40 value 62.468894
## iter 50 value 61.935605
## iter 60 value 60.912114
## iter 70 value 59.416281
## iter 80 value 58.338382
## iter 90 value 58.285290
## iter 100 value 57.972039
## final value 57.972039
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 98.842008
## iter 10 value 28.955405
## iter 20 value 19.944667
## iter 30 value 17.611146
## iter 40 value 10.582223
## iter 50 value 7.941031
## iter 60 value 6.535764
## iter 70 value 6.226752
## iter 80 value 4.861934
## iter 90 value 4.823937
## iter 100 value 4.796571
## final value 4.796571
## stopped after 100 iterations
## # weights: 179
## initial value 99.691223
## iter 10 value 16.235017
## iter 20 value 6.053412
## iter 30 value 0.050231
## iter 40 value 0.001047
## final value 0.000067
## converged
## # weights: 39
## initial value 100.342119
## iter 10 value 66.064036
## iter 20 value 63.202448
## iter 30 value 61.353635
## iter 40 value 58.557329
## iter 50 value 58.329317
## iter 60 value 58.185655
## iter 70 value 57.935024
## iter 80 value 57.929060
## final value 57.929028
## converged
## # weights: 109
## initial value 114.883864
## iter 10 value 60.194812
## iter 20 value 31.552175
## iter 30 value 21.947784
## iter 40 value 17.649294
## iter 50 value 17.059606
## iter 60 value 17.042136
## iter 70 value 17.041686
## iter 70 value 17.041685
## iter 70 value 17.041685
## final value 17.041685
## converged
## # weights: 179
## initial value 135.344635
## iter 10 value 62.292829
## iter 20 value 35.416389
## iter 30 value 25.528215
## iter 40 value 17.108175
## iter 50 value 15.490671
## iter 60 value 13.508727
```

```
## iter 70 value 12.997724
## iter 80 value 12.481864
## iter 90 value 12.450390
## iter 100 value 12.449450
## final value 12.449450
## stopped after 100 iterations
## # weights: 39
## initial value 96.281811
## iter 10 value 63.609748
## iter 20 value 57.973872
## iter 30 value 55.453585
## iter 40 value 55.411477
## iter 50 value 55.345670
## iter 60 value 55.256349
## iter 70 value 55.206647
## iter 80 value 53.726007
## iter 90 value 53.687086
## iter 100 value 53.647190
## final value 53.647190
## stopped after 100 iterations
## # weights: 109
## initial value 103.866339
## iter 10 value 38.025011
## iter 20 value 27.001563
## iter 30 value 21.842538
## iter 40 value 19.261342
## iter 50 value 19.003653
## iter 60 value 18.894927
## iter 70 value 18.811324
## iter 80 value 18.659426
## iter 90 value 18.477321
## iter 100 value 18.298549
## final value 18.298549
## stopped after 100 iterations
## # weights: 179
## initial value 115.247467
## iter 10 value 19.718212
## iter 20 value 2.403281
## iter 30 value 0.186954
## iter 40 value 0.161772
## iter 50 value 0.149354
## iter 60 value 0.134385
## iter 70 value 0.126576
## iter 80 value 0.115716
## iter 90 value 0.103181
## iter 100 value 0.096828
## final value 0.096828
## stopped after 100 iterations
## # weights: 39
## initial value 108.894872
## iter 10 value 73.916422
## iter 20 value 65.266602
## iter 30 value 62.575629
## iter 40 value 61.718241
```

```
## iter 50 value 61.590989
## iter 60 value 61.578241
## iter 70 value 61.574426
## iter 80 value 61.572429
## iter 90 value 61.571914
## iter 100 value 61.571797
## final value 61.571797
## stopped after 100 iterations
## # weights: 109
## initial value 102.560674
## iter 10 value 30.690458
## iter 20 value 14.741499
## iter 30 value 13.482866
## iter 40 value 12.212443
## iter 50 value 12.124849
## iter 60 value 10.814600
## iter 70 value 10.668416
## iter 80 value 10.577347
## iter 90 value 10.563861
## iter 100 value 10.560087
## final value 10.560087
## stopped after 100 iterations
## # weights: 179
## initial value 105.150326
## iter 10 value 20.393961
## iter 20 value 5.560496
## iter 30 value 3.471349
## iter 40 value 3.297772
## iter 50 value 3.295724
## iter 60 value 2.340438
## iter 70 value 1.909851
## final value 1.909619
## converged
## # weights: 39
## initial value 110.168234
## iter 10 value 73.003745
## iter 20 value 64.755496
## iter 30 value 64.610060
## iter 40 value 64.605490
## final value 64.605478
## converged
## # weights: 109
## initial value 113.847105
## iter 10 value 44.279322
## iter 20 value 27.604761
## iter 30 value 21.034629
## iter 40 value 18.344250
## iter 50 value 16.931710
## iter 60 value 16.760816
## iter 70 value 16.754013
## iter 80 value 16.753892
## final value 16.753892
## converged
## # weights: 179
```

```
## initial value 113.759595
## iter 10 value 43.098184
## iter 20 value 20.631854
## iter 30 value 15.325356
## iter 40 value 13.401510
## iter 50 value 12.720648
## iter 60 value 12.631544
## iter 70 value 12.608085
## iter 80 value 12.558423
## iter 90 value 12.493229
## iter 100 value 12.490125
## final value 12.490125
## stopped after 100 iterations
## # weights: 39
## initial value 113.229508
## iter 10 value 72.061956
## iter 20 value 57.918778
## iter 30 value 55.499836
## iter 40 value 52.932819
## iter 50 value 52.587012
## iter 60 value 52.578704
## iter 70 value 52.534078
## iter 80 value 51.803947
## iter 90 value 51.656701
## iter 100 value 51.644981
## final value 51.644981
## stopped after 100 iterations
## # weights: 109
## initial value 107.200968
## iter 10 value 46.784343
## iter 20 value 22.831097
## iter 30 value 8.638815
## iter 40 value 3.959222
## iter 50 value 2.488978
## iter 60 value 2.289492
## iter 70 value 2.250892
## iter 80 value 2.240761
## iter 90 value 2.226854
## iter 100 value 2.219139
## final value 2.219139
## stopped after 100 iterations
## # weights: 179
## initial value 147.812819
## iter 10 value 50.119464
## iter 20 value 15.137491
## iter 30 value 4.521312
## iter 40 value 0.406948
## iter 50 value 0.160106
## iter 60 value 0.143188
## iter 70 value 0.136240
## iter 80 value 0.127416
## iter 90 value 0.119076
## iter 100 value 0.106769
## final value 0.106769
```

```
## stopped after 100 iterations
## # weights: 39
## initial value 96.754508
## iter 10 value 63.341723
## iter 20 value 52.047592
## iter 30 value 47.012724
## iter 40 value 44.223938
## iter 50 value 42.665768
## iter 60 value 42.446280
## iter 70 value 42.385498
## iter 80 value 42.312469
## iter 90 value 42.296377
## iter 100 value 42.293163
## final value 42.293163
## stopped after 100 iterations
## # weights: 109
## initial value 107.598829
## iter 10 value 40.357649
## iter 20 value 4.992283
## iter 30 value 1.932641
## iter 40 value 1.388108
## iter 50 value 1.386534
## iter 60 value 1.386490
## iter 70 value 1.386341
## final value 1.386341
## converged
## # weights: 179
## initial value 130.835202
## iter 10 value 47.799740
## iter 20 value 28.302585
## iter 30 value 25.568887
## iter 40 value 24.101009
## iter 50 value 23.653895
## iter 60 value 23.487577
## iter 70 value 19.583634
## iter 80 value 14.318355
## iter 90 value 13.953269
## iter 100 value 13.219584
## final value 13.219584
## stopped after 100 iterations
## # weights: 39
## initial value 117.625333
## iter 10 value 84.737886
## iter 20 value 76.168626
## iter 30 value 70.701832
## iter 40 value 67.513769
## iter 50 value 66.968536
## iter 60 value 66.961837
## iter 70 value 66.961178
## final value 66.961156
## converged
## # weights: 109
## initial value 114.068774
## iter 10 value 44.952273
```

```
## iter 20 value 25.753066
## iter 30 value 19.603756
## iter 40 value 16.553390
## iter 50 value 16.409820
## iter 60 value 16.403952
## final value 16.403938
## converged
## # weights: 179
## initial value 104.955856
## iter 10 value 55.690905
## iter 20 value 24.103895
## iter 30 value 16.377522
## iter 40 value 13.678623
## iter 50 value 12.943146
## iter 60 value 12.451581
## iter 70 value 12.159373
## iter 80 value 12.017555
## iter 90 value 11.987827
## iter 100 value 11.985743
## final value 11.985743
## stopped after 100 iterations
## # weights: 39
## initial value 117.782625
## iter 10 value 64.787896
## iter 20 value 56.506052
## iter 30 value 51.372711
## iter 40 value 49.728805
## iter 50 value 46.880625
## iter 60 value 46.290194
## iter 70 value 45.209927
## iter 80 value 45.199754
## iter 90 value 40.538430
## iter 100 value 38.787199
## final value 38.787199
## stopped after 100 iterations
## # weights: 109
## initial value 105.848269
## iter 10 value 18.267208
## iter 20 value 3.891816
## iter 30 value 3.804844
## iter 40 value 3.630826
## iter 50 value 3.567384
## iter 60 value 3.531214
## iter 70 value 2.412683
## iter 80 value 1.512288
## iter 90 value 0.212810
## iter 100 value 0.139011
## final value 0.139011
## stopped after 100 iterations
## # weights: 179
## initial value 119.754249
## iter 10 value 19.431984
## iter 20 value 8.523413
## iter 30 value 4.887146
```

```
## iter 40 value 3.219610
## iter 50 value 2.285572
## iter 60 value 2.098373
## iter 70 value 2.056346
## iter 80 value 2.040285
## iter 90 value 2.032331
## iter 100 value 2.020975
## final value 2.020975
## stopped after 100 iterations
## # weights: 39
## initial value 104.764480
## iter 10 value 63.246282
## iter 20 value 61.323781
## iter 30 value 60.611185
## iter 40 value 60.519722
## iter 50 value 57.711113
## iter 60 value 54.692006
## iter 70 value 52.427619
## iter 80 value 51.925607
## iter 90 value 51.238294
## iter 100 value 50.944807
## final value 50.944807
## stopped after 100 iterations
## # weights: 109
## initial value 103.804462
## iter 10 value 47.808466
## iter 20 value 27.109549
## iter 30 value 25.265803
## iter 40 value 23.525787
## iter 50 value 23.256232
## iter 60 value 21.088757
## iter 70 value 20.795420
## iter 80 value 20.501748
## iter 90 value 19.992038
## iter 100 value 14.802605
## final value 14.802605
## stopped after 100 iterations
## # weights: 179
## initial value 114.975822
## iter 10 value 23.694777
## iter 20 value 10.433985
## iter 30 value 1.263466
## iter 40 value 0.058917
## iter 50 value 0.008678
## iter 60 value 0.003870
## iter 70 value 0.001816
## iter 80 value 0.001265
## iter 90 value 0.000424
## iter 100 value 0.000275
## final value 0.000275
## stopped after 100 iterations
## # weights: 39
## initial value 120.569025
## iter 10 value 71.814694
```

```
## iter 20 value 63.688957
## iter 30 value 61.620733
## iter 40 value 58.511222
## iter 50 value 57.961639
## iter 60 value 57.444182
## iter 70 value 57.377931
## iter 80 value 57.003371
## iter 90 value 56.434279
## iter 100 value 56.196210
## final value 56.196210
## stopped after 100 iterations
## # weights: 109
## initial value 103.408545
## iter 10 value 47.612621
## iter 20 value 28.336631
## iter 30 value 22.684814
## iter 40 value 18.948529
## iter 50 value 17.592884
## iter 60 value 16.588068
## iter 70 value 16.519867
## iter 80 value 16.518014
## final value 16.518009
## converged
## # weights: 179
## initial value 104.308140
## iter 10 value 50.166380
## iter 20 value 23.371493
## iter 30 value 14.335170
## iter 40 value 13.686882
## iter 50 value 13.601760
## iter 60 value 13.250208
## iter 70 value 12.900405
## iter 80 value 12.867244
## iter 90 value 12.820714
## iter 100 value 12.756087
## final value 12.756087
## stopped after 100 iterations
## # weights: 39
## initial value 120.621421
## iter 10 value 74.875171
## iter 20 value 56.586052
## iter 30 value 51.122404
## iter 40 value 47.262118
## iter 50 value 45.469478
## iter 60 value 45.313649
## iter 70 value 44.885905
## iter 80 value 44.826018
## iter 90 value 44.731641
## iter 100 value 44.687277
## final value 44.687277
## stopped after 100 iterations
## # weights: 109
## initial value 113.421039
## iter 10 value 31.834062
```

```
## iter 20 value 17.352883
## iter 30 value 12.351665
## iter 40 value 12.045616
## iter 50 value 11.909714
## iter 60 value 11.874033
## iter 70 value 11.648950
## iter 80 value 11.573960
## iter 90 value 11.526576
## iter 100 value 11.519786
## final value 11.519786
## stopped after 100 iterations
## # weights: 179
## initial value 91.351331
## iter 10 value 17.184451
## iter 20 value 0.628226
## iter 30 value 0.246212
## iter 40 value 0.230224
## iter 50 value 0.199532
## iter 60 value 0.175362
## iter 70 value 0.159984
## iter 80 value 0.150411
## iter 90 value 0.143870
## iter 100 value 0.138255
## final value 0.138255
## stopped after 100 iterations
## # weights: 39
## initial value 110.449141
## iter 10 value 67.797707
## iter 20 value 61.818435
## iter 30 value 61.011241
## iter 40 value 60.216677
## iter 50 value 55.241866
## iter 60 value 54.554785
## iter 70 value 54.089802
## iter 80 value 54.088596
## iter 90 value 54.083244
## iter 100 value 53.774009
## final value 53.774009
## stopped after 100 iterations
## # weights: 109
## initial value 102.533147
## iter 10 value 55.427692
## iter 20 value 34.586424
## iter 30 value 25.402713
## iter 40 value 19.138679
## iter 50 value 15.418337
## iter 60 value 13.841208
## iter 70 value 13.533989
## iter 80 value 13.353667
## iter 90 value 13.286907
## iter 100 value 11.271647
## final value 11.271647
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 130.377959
## iter 10 value 32.501684
## iter 20 value 2.826998
## iter 30 value 0.052304
## iter 40 value 0.000183
## final value 0.000081
## converged
## # weights: 39
## initial value 104.728586
## iter 10 value 79.749280
## iter 20 value 67.331347
## iter 30 value 59.104183
## iter 40 value 56.931520
## iter 50 value 56.511163
## iter 60 value 56.263544
## iter 70 value 56.082933
## iter 80 value 56.010199
## final value 56.009809
## converged
## # weights: 109
## initial value 122.370198
## iter 10 value 58.603496
## iter 20 value 33.589782
## iter 30 value 21.254683
## iter 40 value 17.517895
## iter 50 value 15.869222
## iter 60 value 15.408248
## iter 70 value 15.372602
## iter 80 value 15.368923
## iter 90 value 15.368778
## final value 15.368777
## converged
## # weights: 179
## initial value 104.607661
## iter 10 value 38.735094
## iter 20 value 17.333637
## iter 30 value 13.194963
## iter 40 value 12.507821
## iter 50 value 12.387834
## iter 60 value 12.361019
## iter 70 value 12.359437
## iter 80 value 12.359250
## final value 12.359247
## converged
## # weights: 39
## initial value 101.726800
## iter 10 value 66.283439
## iter 20 value 57.303158
## iter 30 value 56.336324
## iter 40 value 55.237768
## iter 50 value 54.294321
## iter 60 value 53.968806
## iter 70 value 51.064280
## iter 80 value 50.433155
```

```
## iter 90 value 48.602883
## iter 100 value 48.401940
## final value 48.401940
## stopped after 100 iterations
## # weights: 109
## initial value 111.955568
## iter 10 value 78.479271
## iter 20 value 43.163637
## iter 30 value 27.507234
## iter 40 value 21.873629
## iter 50 value 20.352327
## iter 60 value 13.567628
## iter 70 value 13.514547
## iter 80 value 13.445732
## iter 90 value 12.894365
## iter 100 value 9.906976
## final value 9.906976
## stopped after 100 iterations
## # weights: 179
## initial value 107.997271
## iter 10 value 22.674823
## iter 20 value 6.362081
## iter 30 value 2.763221
## iter 40 value 2.717509
## iter 50 value 2.696336
## iter 60 value 2.465623
## iter 70 value 2.064416
## iter 80 value 1.260392
## iter 90 value 0.180732
## iter 100 value 0.156331
## final value 0.156331
## stopped after 100 iterations
## # weights: 39
## initial value 103.886470
## iter 10 value 67.344386
## iter 20 value 64.551048
## iter 30 value 63.754207
## iter 40 value 63.460116
## iter 50 value 63.426192
## iter 60 value 63.421501
## iter 70 value 63.421203
## iter 80 value 63.420945
## iter 90 value 63.420821
## final value 63.420814
## converged
## # weights: 109
## initial value 107.331717
## iter 10 value 34.019434
## iter 20 value 24.313956
## iter 30 value 19.528021
## iter 40 value 18.793440
## iter 50 value 15.316729
## iter 60 value 14.713690
## iter 70 value 11.631361
```

```
## iter 80 value 5.907635
## iter 90 value 4.947808
## iter 100 value 3.911174
## final value 3.911174
## stopped after 100 iterations
## # weights: 179
## initial value 119.341946
## iter 10 value 29.260662
## iter 20 value 10.545634
## iter 30 value 5.544076
## iter 40 value 3.055880
## iter 50 value 2.647555
## iter 60 value 0.096522
## iter 70 value 0.038207
## iter 80 value 0.013386
## iter 90 value 0.005843
## iter 100 value 0.001177
## final value 0.001177
## stopped after 100 iterations
## # weights: 39
## initial value 107.943692
## iter 10 value 73.184575
## iter 20 value 65.641014
## iter 30 value 63.547644
## iter 40 value 60.300469
## iter 50 value 60.049537
## iter 60 value 59.579250
## iter 70 value 59.106173
## iter 80 value 58.607341
## iter 90 value 57.949431
## iter 100 value 57.624258
## final value 57.624258
## stopped after 100 iterations
## # weights: 109
## initial value 103.475084
## iter 10 value 53.919216
## iter 20 value 29.723484
## iter 30 value 21.664634
## iter 40 value 19.160742
## iter 50 value 18.748583
## iter 60 value 18.602662
## iter 70 value 17.131767
## iter 80 value 16.901708
## iter 90 value 16.894407
## iter 100 value 16.892603
## final value 16.892603
## stopped after 100 iterations
## # weights: 179
## initial value 108.097968
## iter 10 value 48.924665
## iter 20 value 24.858059
## iter 30 value 15.815989
## iter 40 value 13.403120
## iter 50 value 12.614113
```

```
## iter 60 value 12.460090
## iter 70 value 12.450702
## iter 80 value 12.450540
## final value 12.450536
## converged
## # weights: 39
## initial value 115.454924
## iter 10 value 71.413989
## iter 20 value 62.974229
## iter 30 value 56.606915
## iter 40 value 53.905331
## iter 50 value 53.896478
## iter 60 value 53.892414
## iter 70 value 53.887586
## iter 80 value 52.767746
## iter 90 value 52.751001
## iter 100 value 52.748772
## final value 52.748772
## stopped after 100 iterations
## # weights: 109
## initial value 113.938518
## iter 10 value 35.060278
## iter 20 value 27.042466
## iter 30 value 13.422112
## iter 40 value 6.748367
## iter 50 value 5.740099
## iter 60 value 5.664929
## iter 70 value 5.614855
## iter 80 value 2.445752
## iter 90 value 2.275440
## iter 100 value 2.237137
## final value 2.237137
## stopped after 100 iterations
## # weights: 179
## initial value 118.218415
## iter 10 value 25.909847
## iter 20 value 12.803819
## iter 30 value 6.989589
## iter 40 value 6.510152
## iter 50 value 6.433263
## iter 60 value 6.398587
## iter 70 value 6.368572
## iter 80 value 5.617161
## iter 90 value 4.980381
## iter 100 value 4.918676
## final value 4.918676
## stopped after 100 iterations
## # weights: 39
## initial value 111.018778
## iter 10 value 62.708955
## iter 20 value 56.069550
## iter 30 value 54.005286
## iter 40 value 52.125736
## iter 50 value 51.234403
```

```
## iter 60 value 50.238066
## iter 70 value 49.805674
## iter 80 value 49.316203
## iter 90 value 47.410580
## iter 100 value 47.209908
## final value 47.209908
## stopped after 100 iterations
## # weights: 109
## initial value 117.907953
## iter 10 value 38.436704
## iter 20 value 24.825572
## iter 30 value 22.416180
## iter 40 value 20.491748
## iter 50 value 20.275935
## iter 60 value 20.234888
## iter
        70 value 20.200826
## iter 80 value 20.161646
## iter 90 value 20.003843
## iter 100 value 19.884209
## final value 19.884209
## stopped after 100 iterations
## # weights: 179
## initial value 105.328276
## iter 10 value 20.402097
## iter 20 value 8.646863
## iter 30 value 7.759287
## iter 40 value 6.427525
## iter 50 value 6.329033
## iter 60 value 6.314727
## iter 70 value 5.947411
## iter 80 value 5.037935
## iter 90 value 4.757455
## iter 100 value 4.685897
## final value 4.685897
## stopped after 100 iterations
## # weights: 39
## initial value 107.950372
## iter 10 value 77.465959
## iter 20 value 72.231504
## iter 30 value 66.943362
## iter 40 value 59.435617
## iter 50 value 58.790556
## iter 60 value 58.583038
## iter 70 value 58.532780
## iter 80 value 58.528483
## final value 58.528371
## converged
## # weights: 109
## initial value 104.885899
## iter 10 value 44.535130
## iter 20 value 28.851385
## iter 30 value 20.433509
## iter 40 value 17.720537
## iter 50 value 17.073965
```

```
## iter 60 value 16.949024
## iter 70 value 16.627891
## iter 80 value 15.979658
## iter 90 value 15.968381
## final value 15.968214
## converged
## # weights: 179
## initial value 135.006360
## iter 10 value 40.221471
## iter 20 value 24.276930
## iter 30 value 15.254512
## iter 40 value 13.772375
## iter 50 value 12.933977
## iter 60 value 12.535304
## iter 70 value 12.415618
## iter 80 value 12.399639
## iter 90 value 12.398837
## final value 12.398817
## converged
## # weights:
## initial value 102.252185
## iter 10 value 57.893798
## iter 20 value 54.164271
## iter 30 value 52.992441
## iter 40 value 52.401022
## iter 50 value 51.459193
## iter 60 value 50.369926
## iter 70 value 49.350939
## iter 80 value 49.162610
## iter 90 value 48.353654
## iter 100 value 47.968038
## final value 47.968038
## stopped after 100 iterations
## # weights: 109
## initial value 115.785334
## iter 10 value 24.391870
## iter 20 value 12.769583
## iter 30 value 8.410923
## iter 40 value 8.189229
## iter 50 value 8.172680
## iter 60 value 8.155788
## iter 70 value 8.118178
## iter 80 value 8.074023
## iter 90 value 8.052222
## iter 100 value 8.044779
## final value 8.044779
## stopped after 100 iterations
## # weights: 179
## initial value 134.189972
## iter 10 value 52.438019
## iter 20 value 18.773631
## iter 30 value 12.852092
## iter 40 value 9.980223
## iter 50 value 9.801038
```

```
## iter 60 value 9.648594
## iter 70 value 9.624935
## iter 80 value 9.605365
## iter 90 value 9.567770
## iter 100 value 7.218887
## final value 7.218887
## stopped after 100 iterations
## # weights: 39
## initial value 104.866968
## iter 10 value 73.693383
## iter 20 value 68.395897
## iter 30 value 67.734504
## iter 40 value 67.081068
## iter 50 value 64.541466
## iter 60 value 62.515964
## iter 70 value 62.302233
## iter 80 value 61.067547
## iter 90 value 61.040022
## iter 100 value 61.018462
## final value 61.018462
## stopped after 100 iterations
## # weights: 109
## initial value 103.240359
## iter 10 value 39.397373
## iter 20 value 28.145510
## iter 30 value 22.790912
## iter 40 value 17.300874
## iter 50 value 13.603455
## iter 60 value 10.355601
## iter 70 value 6.048703
## iter 80 value 5.482557
## iter 90 value 5.443515
## iter 100 value 5.425352
## final value 5.425352
## stopped after 100 iterations
## # weights: 179
## initial value 107.657019
## iter 10 value 17.318848
## iter 20 value 4.847240
## iter 30 value 2.824152
## iter 40 value 1.899926
## iter 50 value 1.410387
## iter 60 value 0.014240
## iter 70 value 0.005613
## iter 80 value 0.002647
## iter 90 value 0.001600
## iter 100 value 0.001183
## final value 0.001183
## stopped after 100 iterations
## # weights: 39
## initial value 115.940116
## iter 10 value 80.433108
## iter 20 value 64.166528
## iter 30 value 60.732525
```

```
## iter 40 value 59.635453
## iter 50 value 59.078226
## iter 60 value 58.995451
## iter 70 value 58.780248
## iter 80 value 58.340267
## iter 90 value 58.153291
## iter 100 value 58.142893
## final value 58.142893
## stopped after 100 iterations
## # weights: 109
## initial value 109.906160
## iter 10 value 44.732187
## iter 20 value 32.476660
## iter 30 value 22.268931
## iter 40 value 18.126868
## iter 50 value 17.171169
## iter 60 value 17.071029
## iter 70 value 16.989255
## iter 80 value 16.714127
## iter 90 value 16.706393
## final value 16.706373
## converged
## # weights: 179
## initial value 121.964959
## iter 10 value 32.843419
## iter 20 value 18.458923
## iter 30 value 13.165546
## iter 40 value 12.555487
## iter 50 value 12.510685
## iter 60 value 12.493080
## iter 70 value 12.490919
## iter 80 value 12.490658
## iter 90 value 12.490643
## final value 12.490642
## converged
## # weights: 39
## initial value 100.338530
## iter 10 value 76.169395
## iter 20 value 70.448455
## iter 30 value 54.246536
## iter 40 value 48.802907
## iter 50 value 47.175589
## iter 60 value 47.107765
## iter 70 value 47.054691
## iter 80 value 47.026156
## iter 90 value 47.017621
## iter 100 value 47.010133
## final value 47.010133
## stopped after 100 iterations
## # weights: 109
## initial value 91.979990
## iter 10 value 30.707373
## iter 20 value 21.919875
## iter 30 value 17.714452
```

```
## iter 40 value 12.384744
## iter 50 value 9.426785
## iter 60 value 6.455048
## iter 70 value 6.057448
## iter 80 value 5.762124
## iter 90 value 5.474438
## iter 100 value 4.812094
## final value 4.812094
## stopped after 100 iterations
## # weights: 179
## initial value 90.186043
## iter 10 value 12.282116
## iter 20 value 1.873845
## iter 30 value 1.507811
## iter 40 value 0.473736
## iter 50 value 0.144523
## iter 60 value 0.123729
## iter 70 value 0.118167
## iter 80 value 0.112059
## iter 90 value 0.103234
## iter 100 value 0.095990
## final value 0.095990
## stopped after 100 iterations
## # weights: 39
## initial value 101.796799
## iter 10 value 72.835712
## iter 20 value 70.430026
## iter 30 value 70.013493
## iter 40 value 69.973718
## iter 50 value 69.968970
## iter 60 value 69.968507
## final value 69.967893
## converged
## # weights: 109
## initial value 114.626117
## iter 10 value 38.527097
## iter 20 value 24.571755
## iter 30 value 21.572643
## iter 40 value 20.168706
## iter 50 value 16.688986
## iter 60 value 12.992059
## iter 70 value 12.444683
## iter 80 value 11.546264
## iter 90 value 9.769672
## iter 100 value 7.153478
## final value 7.153478
## stopped after 100 iterations
## # weights: 179
## initial value 108.547980
## iter 10 value 27.010044
## iter 20 value 3.617901
## iter 30 value 1.928378
## iter 40 value 1.911100
## iter 50 value 1.909862
```

```
## iter 60 value 1.909543
## final value 1.909543
## converged
## # weights: 39
## initial value 100.239706
## iter 10 value 78.694682
## iter 20 value 67.427987
## iter 30 value 64.626249
## iter 40 value 61.717163
## iter 50 value 60.361951
## iter 60 value 60.288245
## final value 60.288213
## converged
## # weights: 109
## initial value 106.313650
## iter 10 value 43.068440
## iter 20 value 28.983081
## iter 30 value 22.984883
## iter 40 value 17.894523
## iter 50 value 17.235365
## iter 60 value 16.915588
## iter 70 value 16.358465
## iter 80 value 15.798334
## iter 90 value 15.794051
## final value 15.794037
## converged
## # weights: 179
## initial value 117.629130
## iter 10 value 41.854606
## iter 20 value 26.693568
## iter 30 value 18.463893
## iter 40 value 13.978631
## iter 50 value 12.741489
## iter 60 value 12.604487
## iter 70 value 12.585013
## iter 80 value 12.583861
## final value 12.583839
## converged
## # weights: 39
## initial value 103.631835
## iter 10 value 65.396030
## iter 20 value 54.768146
## iter 30 value 37.966065
## iter 40 value 35.718768
## iter 50 value 35.701729
## iter 60 value 35.688947
## iter 70 value 35.678597
## iter 80 value 35.666540
## iter 90 value 35.663116
## iter 100 value 35.660045
## final value 35.660045
## stopped after 100 iterations
## # weights: 109
## initial value 117.118302
```

```
## iter 10 value 35.340095
## iter 20 value 17.220076
## iter 30 value 14.223567
## iter 40 value 14.030370
## iter 50 value 13.894558
## iter 60 value 13.647213
## iter 70 value 13.204747
## iter 80 value 13.050662
## iter 90 value 12.963808
## iter 100 value 12.927217
## final value 12.927217
## stopped after 100 iterations
## # weights: 179
## initial value 101.943250
## iter 10 value 21.042763
## iter 20 value 5.771815
## iter 30 value 3.656828
## iter
       40 value 3.471391
## iter 50 value 1.991635
## iter 60 value 1.603838
## iter 70 value 1.553675
## iter 80 value 1.546291
## iter 90 value 1.532753
## iter 100 value 1.520303
## final value 1.520303
## stopped after 100 iterations
## # weights: 39
## initial value 99.797558
## iter 10 value 65.945680
## iter 20 value 57.633580
## iter 30 value 54.165715
## iter 40 value 53.010454
## iter 50 value 51.946564
## iter 60 value 50.192761
## iter 70 value 48.545435
## iter 80 value 48.206089
## iter 90 value 48.070908
## iter 100 value 46.161606
## final value 46.161606
## stopped after 100 iterations
## # weights: 109
## initial value 97.850406
## iter 10 value 41.959992
## iter 20 value 10.897053
## iter 30 value 5.645714
## iter 40 value 5.108890
## iter 50 value 3.718183
## iter
       60 value 3.521962
## iter 70 value 3.013266
## iter 80 value 2.338570
## iter 90 value 2.261968
## iter 100 value 2.256348
## final value 2.256348
## stopped after 100 iterations
```

```
## # weights: 179
## initial value 112.764363
## iter 10 value 23.686353
## iter 20 value 3.595541
## iter 30 value 2.243603
## iter 40 value 1.967209
## iter 50 value 1.914465
## iter 60 value 1.910244
## iter 70 value 1.513489
## iter 80 value 0.009331
## iter 90 value 0.003783
## iter 100 value 0.001051
## final value 0.001051
## stopped after 100 iterations
## # weights: 39
## initial value 115.324800
## iter 10 value 67.952091
## iter 20 value 63.175048
## iter 30 value 60.835177
## iter 40 value 58.722498
## iter 50 value 56.706761
## iter 60 value 56.624996
## iter 70 value 56.624357
## final value 56.624349
## converged
## # weights: 109
## initial value 108.072009
## iter 10 value 48.128830
## iter 20 value 30.797125
## iter 30 value 23.500813
## iter 40 value 17.361479
## iter 50 value 15.796159
## iter 60 value 15.355431
## iter 70 value 14.347636
## iter 80 value 14.254360
## final value 14.254081
## converged
## # weights: 179
## initial value 125.039959
## iter 10 value 30.179099
## iter 20 value 14.638748
## iter 30 value 12.198245
## iter 40 value 11.678432
## iter 50 value 11.606010
## iter 60 value 11.599622
## iter 70 value 11.599551
## final value 11.599551
## converged
## # weights: 39
## initial value 107.627648
## iter 10 value 65.001464
## iter 20 value 61.205110
## iter 30 value 57.776204
## iter 40 value 57.720982
```

```
## iter 50 value 57.706946
## iter 60 value 57.246885
## iter 70 value 56.640930
## iter 80 value 56.374022
## iter 90 value 55.642286
## iter 100 value 55.458651
## final value 55.458651
## stopped after 100 iterations
## # weights: 109
## initial value 108.253895
## iter 10 value 29.477915
## iter 20 value 22.458155
## iter 30 value 11.034093
## iter 40 value 10.664936
## iter 50 value 6.319931
## iter 60 value 5.367667
## iter 70 value 5.292838
## iter 80 value 5.271042
## iter 90 value 5.050436
## iter 100 value 4.658994
## final value 4.658994
## stopped after 100 iterations
## # weights: 179
## initial value 126.248823
## iter 10 value 17.031883
## iter 20 value 2.573441
## iter 30 value 2.423008
## iter 40 value 2.397704
## iter 50 value 2.382752
## iter 60 value 1.513084
## iter 70 value 0.350683
## iter 80 value 0.181818
## iter 90 value 0.172458
## iter 100 value 0.165108
## final value 0.165108
## stopped after 100 iterations
## # weights: 39
## initial value 99.092512
## iter 10 value 54.520354
## iter 20 value 52.461179
## iter 30 value 51.131887
## iter 40 value 51.123527
## iter 50 value 51.122706
## final value 51.122704
## converged
## # weights: 109
## initial value 104.389699
## iter 10 value 27.861184
## iter 20 value 16.934191
## iter 30 value 13.458953
## iter 40 value 11.705940
## iter 50 value 11.395266
## iter 60 value 10.838364
## iter 70 value 10.635716
```

```
## iter 80 value 10.235036
## iter 90 value 10.176873
## iter 100 value 10.115408
## final value 10.115408
## stopped after 100 iterations
## # weights: 179
## initial value 101.802878
## iter 10 value 25.946736
## iter 20 value 5.998618
## iter 30 value 2.284677
## iter 40 value 0.292031
## iter 50 value 0.024886
## iter 60 value 0.002933
## iter 70 value 0.000841
## iter 80 value 0.000362
## iter 90 value 0.000126
## iter 90 value 0.000079
## iter 90 value 0.000078
## final value 0.000078
## converged
## # weights: 39
## initial value 106.491510
## iter 10 value 81.936443
## iter 20 value 66.829675
## iter 30 value 58.211194
## iter 40 value 57.699985
## iter 50 value 57.166752
## iter 60 value 57.081910
## final value 57.081874
## converged
## # weights: 109
## initial value 112.703198
## iter 10 value 50.955553
## iter 20 value 34.678609
## iter 30 value 21.833766
## iter 40 value 16.577879
## iter 50 value 15.680038
## iter 60 value 15.480145
## iter 70 value 15.270827
## iter 80 value 15.266449
## final value 15.266422
## converged
## # weights: 179
## initial value 110.261485
## iter 10 value 40.124388
## iter 20 value 20.354962
## iter 30 value 13.681031
## iter 40 value 12.657736
## iter 50 value 12.231811
## iter 60 value 12.161312
## iter 70 value 12.081673
## iter 80 value 12.018442
## iter 90 value 12.017153
## final value 12.017153
```

```
## converged
## # weights: 39
## initial value 115.541619
## iter 10 value 69.763455
## iter 20 value 56.761979
## iter 30 value 56.176657
## iter 40 value 55.957916
## iter 50 value 55.542646
## iter 60 value 54.789151
## iter 70 value 54.609735
## iter 80 value 54.000936
## iter 90 value 53.884704
## iter 100 value 53.738417
## final value 53.738417
## stopped after 100 iterations
## # weights: 109
## initial value 101.441106
## iter 10 value 50.608195
## iter 20 value 29.747735
## iter 30 value 23.999537
## iter 40 value 22.719029
## iter 50 value 22.692488
## iter 60 value 22.288121
## iter 70 value 21.001114
## iter 80 value 20.706163
## iter 90 value 20.662388
## iter 100 value 20.626014
## final value 20.626014
## stopped after 100 iterations
## # weights: 179
## initial value 99.230238
## iter 10 value 26.101055
## iter 20 value 4.938801
## iter 30 value 1.932110
## iter 40 value 0.654283
## iter 50 value 0.330809
## iter 60 value 0.316019
## iter 70 value 0.297295
## iter 80 value 0.287927
## iter 90 value 0.271242
## iter 100 value 0.257000
## final value 0.257000
## stopped after 100 iterations
## # weights: 39
## initial value 104.669963
## iter 10 value 66.864078
## iter 20 value 57.140708
## iter
       30 value 46.290808
## iter 40 value 42.824948
## iter 50 value 39.941510
## iter 60 value 39.464018
## iter 70 value 39.274388
## iter 80 value 39.192357
## iter 90 value 39.187777
```

```
## final value 39.187680
## converged
## # weights: 109
## initial value 102.806021
## iter 10 value 34.130442
## iter 20 value 18.861193
## iter 30 value 15.530283
## iter 40 value 14.860458
## iter 50 value 13.471909
## iter 60 value 12.538304
## iter 70 value 12.487603
## iter 80 value 12.454008
## iter 90 value 11.836627
## iter 100 value 10.357161
## final value 10.357161
## stopped after 100 iterations
## # weights: 179
## initial value 110.056506
## iter 10 value 29.548531
## iter 20 value 7.109959
## iter 30 value 0.138636
## iter 40 value 0.004803
## iter 50 value 0.000520
## final value 0.000079
## converged
## # weights: 39
## initial value 107.606755
## iter 10 value 64.890558
## iter 20 value 62.218865
## iter 30 value 60.288694
## iter 40 value 59.565045
## iter 50 value 59.470008
## iter 60 value 59.438217
## iter 70 value 59.288268
## iter 80 value 58.855666
## iter 90 value 58.387994
## iter 100 value 57.969991
## final value 57.969991
## stopped after 100 iterations
## # weights: 109
## initial value 103.328452
## iter 10 value 45.950875
## iter 20 value 27.123260
## iter 30 value 21.293860
## iter 40 value 18.066032
## iter 50 value 16.870077
## iter 60 value 16.818723
## iter 70 value 16.818030
## final value 16.818018
## converged
## # weights: 179
## initial value 113.379353
## iter 10 value 35.821643
## iter 20 value 19.753527
```

```
## iter 30 value 15.336133
## iter 40 value 13.298086
## iter 50 value 12.679920
## iter 60 value 12.258777
## iter 70 value 12.175820
## iter 80 value 11.870675
## iter 90 value 11.801158
## iter 100 value 11.799397
## final value 11.799397
## stopped after 100 iterations
## # weights: 39
## initial value 105.880894
## iter 10 value 72.160537
## iter 20 value 62.226050
## iter 30 value 60.131566
## iter 40 value 58.574585
## iter 50 value 58.315643
## iter 60 value 58.310338
## iter 70 value 57.111805
## iter 80 value 55.743800
## iter 90 value 54.941298
## iter 100 value 53.855418
## final value 53.855418
## stopped after 100 iterations
## # weights: 109
## initial value 105.445966
## iter 10 value 24.748807
## iter 20 value 8.480186
## iter 30 value 5.052165
## iter 40 value 3.145878
## iter 50 value 3.107283
## iter 60 value 3.086148
## iter 70 value 1.725250
## iter 80 value 0.248367
## iter 90 value 0.203523
## iter 100 value 0.188687
## final value 0.188687
## stopped after 100 iterations
## # weights: 179
## initial value 109.204451
## iter 10 value 39.441681
## iter 20 value 17.093497
## iter 30 value 12.699110
## iter 40 value 11.690340
## iter 50 value 11.587461
## iter 60 value 11.526514
## iter 70 value 11.380410
## iter 80 value 11.260593
## iter 90 value 10.969327
## iter 100 value 10.430343
## final value 10.430343
## stopped after 100 iterations
## # weights: 39
## initial value 98.363151
```

```
## iter 10 value 67.624509
## iter 20 value 60.633832
## iter 30 value 57.193424
## iter 40 value 56.093446
## iter 50 value 55.973291
## iter 60 value 55.956924
## iter 70 value 55.955302
## iter 80 value 55.954574
## iter 90 value 55.954275
## iter 100 value 55.954048
## final value 55.954048
## stopped after 100 iterations
## # weights: 109
## initial value 103.419247
## iter 10 value 46.156053
## iter 20 value 23.139476
## iter 30 value 13.198848
## iter 40 value 10.844448
## iter 50 value 2.336920
## iter 60 value 0.047963
## iter 70 value 0.011123
## iter 80 value 0.005462
## iter 90 value 0.003336
## iter 100 value 0.000468
## final value 0.000468
## stopped after 100 iterations
## # weights: 179
## initial value 106.351747
## iter 10 value 35.788512
## iter 20 value 21.996247
## iter 30 value 19.139947
## iter 40 value 18.227073
## iter 50 value 16.185335
## iter 60 value 14.453311
## iter 70 value 14.242484
## iter 80 value 13.906471
## iter 90 value 13.542946
## iter 100 value 12.614270
## final value 12.614270
## stopped after 100 iterations
## # weights: 39
## initial value 101.909049
## iter 10 value 83.815337
## iter 20 value 61.818776
## iter 30 value 58.439010
## iter 40 value 56.635246
## iter 50 value 55.989602
## iter 60 value 55.965725
## final value 55.965721
## converged
## # weights: 109
## initial value 113.992190
## iter 10 value 34.746678
## iter 20 value 19.581461
```

```
## iter 30 value 16.901982
## iter 40 value 16.466751
## iter 50 value 16.430641
## iter 60 value 16.430493
## iter 60 value 16.430493
## iter 60 value 16.430493
## final value 16.430493
## converged
## # weights: 179
## initial value 101.122494
## iter 10 value 40.285028
## iter 20 value 19.781269
## iter 30 value 13.669962
## iter 40 value 12.997945
## iter 50 value 12.772024
## iter 60 value 12.757873
## iter 70 value 12.753326
## iter 80 value 12.750597
## iter 90 value 12.737905
## iter 100 value 12.735693
## final value 12.735693
## stopped after 100 iterations
## # weights: 39
## initial value 98.298796
## iter 10 value 61.422001
## iter 20 value 58.642248
## iter 30 value 58.065218
## iter 40 value 58.046885
## iter 50 value 57.865301
## iter 60 value 57.552258
## iter 70 value 57.350447
## iter 80 value 57.061957
## iter 90 value 56.647747
## iter 100 value 56.521577
## final value 56.521577
## stopped after 100 iterations
## # weights: 109
## initial value 109.441374
## iter 10 value 49.270731
## iter 20 value 26.177021
## iter 30 value 21.712290
## iter 40 value 18.950270
## iter 50 value 18.532337
## iter 60 value 17.944845
## iter 70 value 17.750274
## iter 80 value 16.764428
## iter 90 value 14.358276
## iter 100 value 14.257871
## final value 14.257871
## stopped after 100 iterations
## # weights: 179
## initial value 107.599108
## iter 10 value 31.782188
## iter 20 value 11.582940
```

```
## iter 30 value 9.534893
## iter 40 value 8.602921
## iter 50 value 8.055574
## iter 60 value 6.349979
## iter 70 value 6.241851
## iter 80 value 6.199797
## iter 90 value 6.183823
## iter 100 value 6.172418
## final value 6.172418
## stopped after 100 iterations
## # weights: 39
## initial value 105.743871
## iter 10 value 54.754192
## iter 20 value 51.203334
## iter 30 value 50.923013
## iter 40 value 50.526202
## iter 50 value 49.996790
## iter 60 value 49.701227
## iter 70 value 49.551739
## iter 80 value 48.595680
## iter 90 value 48.542432
## iter 100 value 48.291619
## final value 48.291619
## stopped after 100 iterations
## # weights: 109
## initial value 98.378575
## iter 10 value 35.522758
## iter 20 value 19.777993
## iter 30 value 13.389981
## iter 40 value 12.625373
## iter 50 value 12.556115
## iter 60 value 10.750258
## iter 70 value 10.589091
## iter 80 value 10.133228
## iter 90 value 8.204474
## iter 100 value 8.166898
## final value 8.166898
## stopped after 100 iterations
## # weights: 179
## initial value 113.392088
## iter 10 value 32.077971
## iter 20 value 17.442423
## iter 30 value 13.508386
## iter 40 value 9.728473
## iter 50 value 6.937042
## iter 60 value 5.789156
## iter 70 value 4.576619
## iter 80 value 4.371211
## iter 90 value 3.133403
## iter 100 value 2.953968
## final value 2.953968
## stopped after 100 iterations
## # weights: 39
## initial value 110.298436
```

```
## iter 10 value 70.831364
## iter 20 value 61.775637
## iter 30 value 57.984538
## iter 40 value 57.599174
## final value 57.592674
## converged
## # weights: 109
## initial value 101.927428
## iter 10 value 46.247144
## iter 20 value 22.521389
## iter 30 value 17.588709
## iter 40 value 16.488760
## iter 50 value 16.460482
## iter 60 value 16.459684
## iter 70 value 16.183518
## iter 80 value 16.137454
## iter 90 value 16.137075
## final value 16.137075
## converged
## # weights: 179
## initial value 124.797322
## iter 10 value 45.810230
## iter 20 value 24.971244
## iter 30 value 17.316854
## iter 40 value 14.727914
## iter 50 value 12.960653
## iter 60 value 12.373773
## iter 70 value 12.074334
## iter 80 value 11.976357
## iter 90 value 11.968884
## iter 100 value 11.967812
## final value 11.967812
## stopped after 100 iterations
## # weights: 39
## initial value 101.612215
## iter 10 value 64.843294
## iter 20 value 57.960247
## iter 30 value 53.378408
## iter 40 value 50.908074
## iter 50 value 48.054990
## iter 60 value 48.041764
## iter 70 value 48.038991
## iter 80 value 48.035420
## iter 90 value 48.034289
## iter 100 value 48.032511
## final value 48.032511
## stopped after 100 iterations
## # weights: 109
## initial value 128.191366
## iter 10 value 55.618530
## iter 20 value 41.772728
## iter 30 value 31.440104
## iter 40 value 29.638258
## iter 50 value 26.358346
```

```
## iter 60 value 25.134567
## iter 70 value 24.992638
## iter 80 value 24.681812
## iter 90 value 24.003667
## iter 100 value 19.789404
## final value 19.789404
## stopped after 100 iterations
## # weights: 179
## initial value 100.306791
## iter 10 value 26.609017
## iter 20 value 13.561397
## iter 30 value 11.641858
## iter 40 value 9.221786
## iter 50 value 7.792442
## iter 60 value 6.398922
## iter 70 value 5.003299
## iter 80 value 4.919302
## iter 90 value 4.908787
## iter 100 value 2.475845
## final value 2.475845
## stopped after 100 iterations
## # weights: 39
## initial value 102.553626
## iter 10 value 70.148006
## iter 20 value 67.830716
## iter 30 value 67.439081
## iter 40 value 67.093759
## iter 50 value 66.560196
## iter 60 value 64.630462
## iter 70 value 64.624002
## iter 80 value 64.060551
## iter 90 value 64.000883
## iter 100 value 63.992481
## final value 63.992481
## stopped after 100 iterations
## # weights: 109
## initial value 97.286293
## iter 10 value 20.656260
## iter 20 value 11.181246
## iter 30 value 3.458362
## iter 40 value 3.367512
## iter 50 value 3.365139
## iter 60 value 3.365060
## final value 3.365060
## converged
## # weights: 179
## initial value 132.619398
## iter 10 value 22.534098
## iter 20 value 8.917627
## iter 30 value 3.969384
## iter 40 value 3.892390
## iter 50 value 3.888532
## iter 60 value 3.888307
## final value 3.888307
```

```
## converged
## # weights: 39
## initial value 109.991995
## iter 10 value 69.334335
## iter 20 value 60.102021
## iter 30 value 56.653929
## iter 40 value 56.310702
## iter 50 value 56.001484
## final value 55.998821
## converged
## # weights: 109
## initial value 107.568757
## iter 10 value 44.422833
## iter 20 value 27.715355
## iter 30 value 19.698976
## iter 40 value 16.187469
## iter 50 value 15.878375
## iter 60 value 15.710391
## iter 70 value 15.675832
## final value 15.675699
## converged
## # weights: 179
## initial value 165.953703
## iter 10 value 61.975633
## iter 20 value 20.955199
## iter 30 value 13.015305
## iter 40 value 12.076087
## iter 50 value 11.885068
## iter 60 value 11.828839
## iter 70 value 11.781080
## iter 80 value 11.776055
## iter 90 value 11.775850
## iter 100 value 11.775827
## final value 11.775827
## stopped after 100 iterations
## # weights: 39
## initial value 108.794921
## iter 10 value 71.007891
## iter 20 value 68.856002
## iter 30 value 68.210033
## iter 40 value 68.076816
## iter 50 value 67.987203
## iter 60 value 67.950347
## iter 70 value 67.938321
## iter 80 value 67.928479
## iter 90 value 67.923468
## iter 100 value 67.241571
## final value 67.241571
## stopped after 100 iterations
## # weights: 109
## initial value 107.900894
## iter 10 value 40.841989
## iter 20 value 21.071311
## iter 30 value 13.768799
```

```
## iter 40 value 12.782516
## iter 50 value 12.716668
## iter 60 value 12.250581
## iter 70 value 10.983509
## iter 80 value 10.752676
## iter 90 value 10.665352
## iter 100 value 10.636042
## final value 10.636042
## stopped after 100 iterations
## # weights: 179
## initial value 112.823429
## iter 10 value 36.415183
## iter 20 value 22.613074
## iter 30 value 19.019331
## iter 40 value 17.745161
## iter 50 value 17.402077
## iter 60 value 16.645010
## iter 70 value 15.959768
## iter 80 value 15.620836
## iter 90 value 15.604400
## iter 100 value 15.576921
## final value 15.576921
## stopped after 100 iterations
## # weights: 39
## initial value 108.629768
## iter 10 value 67.402471
## iter 20 value 58.695268
## iter 30 value 53.511530
## iter 40 value 49.684815
## iter 50 value 48.739748
## iter 60 value 48.677250
## iter 70 value 47.712642
## iter 80 value 46.707818
## iter 90 value 46.627675
## iter 100 value 46.606485
## final value 46.606485
## stopped after 100 iterations
## # weights: 109
## initial value 113.450153
## iter 10 value 46.638019
## iter 20 value 29.465079
## iter 30 value 27.548786
## iter 40 value 24.093047
## iter 50 value 22.343031
## iter 60 value 20.104562
## iter 70 value 14.460399
## iter 80 value 13.923829
## iter 90 value 12.317854
## iter 100 value 9.034738
## final value 9.034738
## stopped after 100 iterations
## # weights: 179
## initial value 113.906049
## iter 10 value 34.327061
```

```
## iter 20 value 5.716226
## iter 30 value 2.428107
## iter 40 value 1.931394
## iter 50 value 1.910801
## iter 60 value 1.909633
## iter 70 value 1.906771
## iter 80 value 1.387489
## iter 90 value 1.386314
## iter 100 value 1.386312
## final value 1.386312
## stopped after 100 iterations
## # weights: 39
## initial value 122.529973
## iter 10 value 79.396166
## iter 20 value 64.985604
## iter 30 value 62.206387
## iter 40 value 60.081526
## iter 50 value 59.708559
## final value 59.706366
## converged
## # weights: 109
## initial value 110.888675
## iter 10 value 54.470268
## iter 20 value 28.999347
## iter 30 value 20.760679
## iter 40 value 18.376026
## iter 50 value 16.044729
## iter 60 value 15.885068
## iter 70 value 15.884350
## final value 15.884347
## converged
## # weights: 179
## initial value 128.878174
## iter 10 value 56.205250
## iter 20 value 24.379722
## iter 30 value 16.468165
## iter 40 value 13.774095
## iter 50 value 13.161681
## iter 60 value 13.022688
## iter 70 value 12.972723
## iter 80 value 12.865362
## iter 90 value 12.769547
## iter 100 value 12.621778
## final value 12.621778
## stopped after 100 iterations
## # weights: 39
## initial value 112.224905
## iter 10 value 61.512202
## iter 20 value 56.892719
## iter 30 value 55.881156
## iter 40 value 55.852551
## iter 50 value 54.757422
## iter 60 value 52.780328
## iter 70 value 52.379012
```

```
## iter 80 value 52.361437
## iter 90 value 52.351697
## iter 100 value 52.348910
## final value 52.348910
## stopped after 100 iterations
## # weights: 109
## initial value 114.821735
## iter 10 value 35.898037
## iter 20 value 15.665394
## iter 30 value 13.078620
## iter 40 value 12.855718
## iter 50 value 12.697503
## iter 60 value 12.292885
## iter 70 value 12.177201
## iter 80 value 12.150601
## iter 90 value 12.114948
## iter 100 value 12.090412
## final value 12.090412
## stopped after 100 iterations
## # weights: 179
## initial value 112.748959
## iter 10 value 28.986174
## iter 20 value 8.680665
## iter 30 value 3.286974
## iter 40 value 3.007362
## iter 50 value 2.944428
## iter 60 value 2.938071
## iter 70 value 2.927772
## iter 80 value 2.870727
## iter 90 value 2.055352
## iter 100 value 2.048296
## final value 2.048296
## stopped after 100 iterations
## # weights: 39
## initial value 95.850764
## iter 10 value 62.453250
## iter 20 value 57.903188
## iter 30 value 50.615237
## iter 40 value 47.879036
## iter 50 value 47.216456
## iter 60 value 47.167493
## iter 70 value 46.902150
## iter 80 value 45.775966
## iter 90 value 45.761625
## iter 100 value 45.350463
## final value 45.350463
## stopped after 100 iterations
## # weights: 109
## initial value 115.113788
## iter 10 value 35.844141
## iter 20 value 18.404776
## iter 30 value 16.750748
## iter 40 value 16.243210
## iter 50 value 16.191514
```

```
## iter 60 value 16.188828
## iter 70 value 16.188421
## iter 80 value 16.188386
## final value 16.188378
## converged
## # weights: 179
## initial value 98.513063
## iter 10 value 11.027990
## iter 20 value 0.375485
## iter 30 value 0.003338
## final value 0.000063
## converged
## # weights: 39
## initial value 98.246345
## iter 10 value 75.637567
## iter 20 value 62.043113
## iter 30 value 61.098588
## iter 40 value 60.071421
## iter 50 value 57.564738
## iter 60 value 57.131452
## iter 70 value 56.347131
## iter 80 value 55.781382
## iter 90 value 55.752338
## final value 55.752265
## converged
## # weights: 109
## initial value 97.917261
## iter 10 value 49.815846
## iter 20 value 29.049349
## iter 30 value 18.481692
## iter 40 value 16.472866
## iter 50 value 16.343455
## iter 60 value 16.341053
## iter 70 value 16.341018
## iter 70 value 16.341018
## iter 70 value 16.341018
## final value 16.341018
## converged
## # weights: 179
## initial value 116.019447
## iter 10 value 34.449699
## iter 20 value 16.765738
## iter 30 value 12.384215
## iter 40 value 11.945279
## iter 50 value 11.793083
## iter 60 value 11.659182
## iter 70 value 11.654419
## iter 80 value 11.653949
## iter 90 value 11.653946
## final value 11.653945
## converged
## # weights: 39
## initial value 97.285276
## iter 10 value 56.132521
```

```
## iter 20 value 53.182804
## iter 30 value 52.474062
## iter 40 value 51.545756
## iter 50 value 51.254303
## iter 60 value 50.260952
## iter 70 value 47.860074
## iter 80 value 47.002980
## iter 90 value 43.680638
## iter 100 value 42.720709
## final value 42.720709
## stopped after 100 iterations
## # weights: 109
## initial value 110.367790
## iter 10 value 32.758854
## iter 20 value 24.553723
## iter 30 value 22.242126
## iter 40 value 18.200074
## iter 50 value 16.325706
## iter 60 value 14.797029
## iter 70 value 13.842307
## iter 80 value 13.463217
## iter 90 value 13.446519
## iter 100 value 13.436926
## final value 13.436926
## stopped after 100 iterations
## # weights: 179
## initial value 103.522266
## iter 10 value 24.327191
## iter 20 value 11.539824
## iter 30 value 5.156288
## iter 40 value 3.623096
## iter 50 value 3.258208
## iter 60 value 3.159337
## iter 70 value 2.602408
## iter 80 value 2.270124
## iter 90 value 2.239744
## iter 100 value 2.198266
## final value 2.198266
## stopped after 100 iterations
## # weights: 39
## initial value 104.273895
## iter 10 value 80.178196
## iter 20 value 66.431267
## iter 30 value 63.061843
## iter 40 value 59.589533
## iter 50 value 56.643400
## iter 60 value 53.768765
## iter 70 value 53.402926
## iter 80 value 53.359037
## iter 90 value 53.343772
## iter 100 value 53.340174
## final value 53.340174
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 104.200368
## iter 10 value 21.243159
## iter 20 value 14.672576
## iter 30 value 13.342334
## iter 40 value 12.862696
## iter 50 value 12.394133
## iter 60 value 12.358633
## iter 70 value 12.247942
## iter 80 value 12.200295
## iter 90 value 12.022641
## iter 100 value 11.489501
## final value 11.489501
## stopped after 100 iterations
## # weights: 179
## initial value 111.120113
## iter 10 value 37.173051
## iter 20 value 13.196762
## iter 30 value 5.858393
## iter 40 value 3.103790
## iter 50 value 0.863028
## iter 60 value 0.317633
## iter 70 value 0.115295
## iter 80 value 0.021825
## iter 90 value 0.006982
## iter 100 value 0.001611
## final value 0.001611
## stopped after 100 iterations
## # weights: 39
## initial value 104.685204
## iter 10 value 75.140028
## iter 20 value 65.893640
## iter 30 value 64.216871
## iter 40 value 61.255605
## iter 50 value 59.087264
## iter 60 value 58.293563
## iter 70 value 58.015434
## iter 80 value 57.571189
## iter 90 value 57.133244
## iter 100 value 57.114779
## final value 57.114779
## stopped after 100 iterations
## # weights: 109
## initial value 105.860992
## iter 10 value 52.608029
## iter 20 value 26.844929
## iter 30 value 18.915280
## iter 40 value 16.107803
## iter 50 value 15.697892
## iter 60 value 15.682900
## iter 70 value 15.682415
## final value 15.682415
## converged
## # weights: 179
## initial value 100.267651
```

```
## iter 10 value 26.444931
## iter 20 value 16.495048
## iter 30 value 14.037741
## iter 40 value 13.412082
## iter 50 value 12.767397
## iter 60 value 12.364658
## iter 70 value 12.289889
## iter 80 value 12.274146
## iter 90 value 11.989018
## iter 100 value 11.937742
## final value 11.937742
## stopped after 100 iterations
## # weights: 39
## initial value 109.747464
## iter 10 value 56.331905
## iter 20 value 49.041478
## iter 30 value 46.397400
## iter 40 value 45.541175
## iter 50 value 45.428697
## iter 60 value 44.929897
## iter 70 value 44.795720
## iter 80 value 44.595521
## iter 90 value 44.015823
## iter 100 value 43.945282
## final value 43.945282
## stopped after 100 iterations
## # weights: 109
## initial value 113.605104
## iter 10 value 43.466165
## iter 20 value 13.322778
## iter 30 value 12.275673
## iter 40 value 3.683061
## iter 50 value 3.018197
## iter 60 value 2.891836
## iter 70 value 2.684194
## iter 80 value 2.670181
## iter 90 value 2.530119
## iter 100 value 1.568872
## final value 1.568872
## stopped after 100 iterations
## # weights: 179
## initial value 106.179434
## iter 10 value 20.801837
## iter 20 value 2.620206
## iter 30 value 0.302520
## iter 40 value 0.250760
## iter 50 value 0.228579
## iter
       60 value 0.215545
## iter 70 value 0.203795
## iter 80 value 0.192144
## iter 90 value 0.184748
## iter 100 value 0.171009
## final value 0.171009
## stopped after 100 iterations
```

```
## # weights: 39
## initial value 100.973925
## iter 10 value 57.784520
## iter 20 value 52.654989
## iter 30 value 50.229106
## iter 40 value 50.017123
## iter 50 value 49.974248
## iter 60 value 49.956866
## iter 70 value 49.940940
## iter 80 value 49.937328
## iter 90 value 49.936365
## iter 100 value 49.925764
## final value 49.925764
## stopped after 100 iterations
## # weights: 109
## initial value 105.825365
## iter 10 value 76.407707
## iter 20 value 22.095252
## iter 30 value 18.100747
## iter 40 value 17.651635
## iter 50 value 17.234190
## iter 60 value 17.117495
## iter 70 value 16.987831
## iter 80 value 16.946128
## iter 90 value 16.910896
## iter 100 value 16.543400
## final value 16.543400
## stopped after 100 iterations
## # weights: 179
## initial value 142.362018
## iter 10 value 30.525406
## iter 20 value 15.346714
## iter 30 value 9.294052
## iter 40 value 8.462413
## iter 50 value 8.252972
## iter 60 value 7.937159
## iter 70 value 7.013596
## iter 80 value 5.452642
## iter 90 value 5.413474
## iter 100 value 5.398253
## final value 5.398253
## stopped after 100 iterations
## # weights: 39
## initial value 100.160587
## iter 10 value 76.431954
## iter 20 value 64.302241
## iter 30 value 60.977127
## iter 40 value 59.615263
## iter 50 value 58.452224
## iter 60 value 58.296665
## final value 58.296647
## converged
## # weights: 109
## initial value 103.178689
```

```
## iter 10 value 55.083940
## iter 20 value 35.274241
## iter 30 value 21.559043
## iter 40 value 16.599123
## iter 50 value 16.115686
## iter 60 value 15.885448
## iter 70 value 15.668320
## iter 80 value 15.658130
## final value 15.658085
## converged
## # weights: 179
## initial value 119.207751
## iter 10 value 35.043853
## iter 20 value 19.259180
## iter 30 value 14.215892
## iter 40 value 13.225754
## iter 50 value 12.315779
## iter 60 value 12.073245
## iter 70 value 12.054459
## iter 80 value 12.053437
## final value 12.053428
## converged
## # weights: 39
## initial value 102.870859
## iter 10 value 65.530454
## iter 20 value 57.153446
## iter 30 value 51.913187
## iter 40 value 51.479502
## iter 50 value 51.473677
## iter 60 value 51.472720
## iter 70 value 51.471604
## iter 80 value 51.463795
## iter 90 value 50.363243
## iter 100 value 46.940685
## final value 46.940685
## stopped after 100 iterations
## # weights: 109
## initial value 106.702826
## iter 10 value 39.530385
## iter 20 value 31.274918
## iter 30 value 28.143944
## iter 40 value 28.086041
## iter 50 value 28.029798
## iter 60 value 28.011869
## iter 70 value 27.902788
## iter 80 value 27.278051
## iter 90 value 22.609223
## iter 100 value 19.270466
## final value 19.270466
## stopped after 100 iterations
## # weights: 179
## initial value 131.102634
## iter 10 value 33.895810
## iter 20 value 15.953847
```

```
## iter 30 value 1.198414
## iter 40 value 0.281149
## iter 50 value 0.245686
## iter 60 value 0.220006
## iter 70 value 0.197431
## iter 80 value 0.173932
## iter 90 value 0.161048
## iter 100 value 0.145110
## final value 0.145110
## stopped after 100 iterations
## # weights: 39
## initial value 103.592126
## iter 10 value 79.208572
## iter 20 value 73.755657
## iter 30 value 72.896207
## iter 40 value 71.964296
## iter 50 value 71.126707
## iter 60 value 70.989323
## iter 70 value 70.785329
## iter 80 value 70.698559
## iter 90 value 69.205389
## iter 100 value 68.510788
## final value 68.510788
## stopped after 100 iterations
## # weights: 109
## initial value 106.964333
## iter 10 value 44.649207
## iter 20 value 21.816080
## iter 30 value 15.075309
## iter 40 value 14.089762
## iter 50 value 13.680062
## iter 60 value 13.633934
## iter 70 value 13.627799
## iter 80 value 13.627371
## iter 90 value 13.627071
## iter 100 value 13.626788
## final value 13.626788
## stopped after 100 iterations
## # weights: 179
## initial value 141.144873
## iter 10 value 54.513365
## iter 20 value 31.163049
## iter 30 value 18.840031
## iter 40 value 6.465195
## iter 50 value 0.491711
## iter 60 value 0.133729
## iter 70 value 0.035622
## iter 80 value 0.006694
## iter 90 value 0.002690
## iter 100 value 0.000853
## final value 0.000853
## stopped after 100 iterations
## # weights: 39
## initial value 100.055874
```

```
## iter 10 value 73.232858
## iter 20 value 64.678985
## iter 30 value 61.608996
## iter 40 value 56.456073
## iter 50 value 55.024790
## iter 60 value 54.999220
## final value 54.999116
## converged
## # weights: 109
## initial value 101.514721
## iter 10 value 45.969373
## iter 20 value 26.365913
## iter 30 value 21.266877
## iter 40 value 16.977484
## iter 50 value 16.230839
## iter 60 value 16.217339
## iter 70 value 16.209484
## iter 80 value 16.187465
## iter 90 value 16.169314
## iter 100 value 16.168323
## final value 16.168323
## stopped after 100 iterations
## # weights: 179
## initial value 104.303083
## iter 10 value 35.011912
## iter 20 value 19.831508
## iter 30 value 14.055075
## iter 40 value 12.933150
## iter 50 value 12.813755
## iter 60 value 12.306662
## iter 70 value 11.898152
## iter 80 value 11.837618
## iter 90 value 11.837181
## final value 11.837157
## converged
## # weights: 39
## initial value 96.801867
## iter 10 value 63.620322
## iter 20 value 57.892841
## iter 30 value 56.838873
## iter 40 value 56.219903
## iter 50 value 53.626096
## iter 60 value 49.140950
## iter 70 value 44.668155
## iter 80 value 40.548887
## iter 90 value 36.706260
## iter 100 value 36.109958
## final value 36.109958
## stopped after 100 iterations
## # weights: 109
## initial value 109.479868
## iter 10 value 32.572522
## iter 20 value 19.541014
## iter 30 value 18.301975
```

```
## iter 40 value 18.104545
## iter 50 value 18.074769
## iter 60 value 18.059927
## iter 70 value 17.614017
## iter 80 value 17.574021
## iter 90 value 16.286495
## iter 100 value 15.855222
## final value 15.855222
## stopped after 100 iterations
## # weights: 179
## initial value 111.920541
## iter 10 value 37.578512
## iter 20 value 12.645302
## iter 30 value 11.388523
## iter 40 value 9.620623
## iter 50 value 9.098781
## iter 60 value 8.992492
## iter 70 value 8.891099
## iter 80 value 8.697270
## iter 90 value 8.129159
## iter 100 value 4.823805
## final value 4.823805
## stopped after 100 iterations
## # weights: 39
## initial value 111.713833
## iter 10 value 79.246710
## iter 20 value 57.860666
## iter 30 value 56.787143
## iter 40 value 56.250428
## iter 50 value 55.686524
## iter 60 value 54.387649
## iter 70 value 53.317710
## iter 80 value 53.035131
## iter 90 value 51.536488
## iter 100 value 51.109565
## final value 51.109565
## stopped after 100 iterations
## # weights: 109
## initial value 120.647719
## iter 10 value 63.132511
## iter 20 value 36.602502
## iter 30 value 26.648292
## iter 40 value 23.688316
## iter 50 value 21.633203
## iter 60 value 21.402477
## iter 70 value 21.036864
## iter 80 value 20.353243
## iter 90 value 19.863080
## iter 100 value 18.247658
## final value 18.247658
## stopped after 100 iterations
## # weights: 179
## initial value 104.277310
## iter 10 value 23.973221
```

```
## iter 20 value 4.155488
## iter 30 value 2.297378
## iter 40 value 2.250008
## iter 50 value 2.249362
## final value 2.249356
## converged
## # weights: 39
## initial value 102.263714
## iter 10 value 72.966869
## iter 20 value 63.939617
## iter 30 value 57.829239
## iter 40 value 56.970629
## iter 50 value 56.742280
## iter 60 value 56.326926
## iter 70 value 55.708456
## iter 80 value 55.686515
## final value 55.686510
## converged
## # weights: 109
## initial value 110.621111
## iter 10 value 42.607705
## iter 20 value 22.965918
## iter 30 value 17.436947
## iter 40 value 16.868241
## iter 50 value 16.831300
## iter 60 value 16.829190
## iter 70 value 16.829109
## final value 16.829108
## converged
## # weights: 179
## initial value 136.559433
## iter 10 value 47.009197
## iter 20 value 26.503933
## iter 30 value 17.023482
## iter 40 value 13.653284
## iter 50 value 12.781369
## iter 60 value 12.629725
## iter 70 value 12.604546
## iter 80 value 12.602509
## iter 90 value 12.579541
## iter 100 value 12.572201
## final value 12.572201
## stopped after 100 iterations
## # weights: 39
## initial value 106.054116
## iter 10 value 68.605511
## iter 20 value 62.390504
## iter 30 value 60.442573
## iter 40 value 59.435690
## iter 50 value 54.401535
## iter 60 value 50.280244
## iter 70 value 48.810406
## iter 80 value 48.149396
## iter 90 value 47.956578
```

```
## iter 100 value 47.579645
## final value 47.579645
## stopped after 100 iterations
## # weights: 109
## initial value 112.321504
## iter 10 value 47.353748
## iter 20 value 26.897734
## iter 30 value 15.043379
## iter 40 value 12.644526
## iter 50 value 11.953823
## iter 60 value 11.673357
## iter 70 value 11.259692
## iter 80 value 10.991315
## iter 90 value 9.303511
## iter 100 value 9.209382
## final value 9.209382
## stopped after 100 iterations
## # weights: 179
## initial value 102.179906
## iter 10 value 18.568774
## iter 20 value 3.508090
## iter 30 value 0.655172
## iter 40 value 0.185853
## iter 50 value 0.167920
## iter 60 value 0.159749
## iter 70 value 0.144850
## iter 80 value 0.132991
## iter 90 value 0.117110
## iter 100 value 0.106324
## final value 0.106324
## stopped after 100 iterations
## # weights: 39
## initial value 111.480998
## iter 10 value 78.242925
## iter 20 value 73.444508
## iter 30 value 69.506263
## iter 40 value 68.014797
## iter 50 value 67.849233
## iter 60 value 67.828279
## iter 70 value 67.825893
## iter 80 value 67.825610
## iter 90 value 67.825386
## final value 67.825338
## converged
## # weights: 109
## initial value 114.175458
## iter 10 value 50.689604
## iter 20 value 37.328890
## iter 30 value 31.590767
## iter 40 value 31.457850
## iter 50 value 31.296538
## iter 60 value 31.213510
## iter 70 value 31.197732
## iter 80 value 31.146990
```

```
## iter 90 value 31.111700
## iter 100 value 31.104484
## final value 31.104484
## stopped after 100 iterations
## # weights: 179
## initial value 136.031181
## iter 10 value 29.466597
## iter 20 value 15.446634
## iter 30 value 11.987605
## iter 40 value 9.698625
## iter 50 value 7.126129
## iter 60 value 6.087501
## iter 70 value 5.762533
## iter 80 value 5.549954
## iter 90 value 5.427731
## iter 100 value 4.661472
## final value 4.661472
## stopped after 100 iterations
## # weights: 39
## initial value 108.610709
## iter 10 value 92.176523
## iter 20 value 70.321851
## iter 30 value 62.516397
## iter 40 value 61.273200
## iter 50 value 61.262374
## final value 61.262364
## converged
## # weights: 109
## initial value 142.176533
## iter 10 value 75.274168
## iter 20 value 47.143641
## iter 30 value 34.925998
## iter 40 value 22.069397
## iter 50 value 17.370080
## iter 60 value 16.839116
## iter 70 value 16.697302
## iter 80 value 16.679119
## iter 90 value 16.678242
## iter 100 value 16.678227
## final value 16.678227
## stopped after 100 iterations
## # weights: 179
## initial value 114.377077
## iter 10 value 39.244631
## iter 20 value 18.233864
## iter 30 value 15.055457
## iter 40 value 13.744873
## iter 50 value 13.089238
## iter 60 value 12.959814
## iter 70 value 12.677389
## iter 80 value 12.619670
## iter 90 value 12.618037
## final value 12.618011
## converged
```

```
## # weights: 39
## initial value 107.724661
## iter 10 value 71.430965
## iter 20 value 66.711827
## iter 30 value 65.020372
## iter 40 value 64.229349
## iter 50 value 64.141627
## iter 60 value 61.321387
## iter 70 value 60.498816
## iter 80 value 60.093404
## iter 90 value 59.423075
## iter 100 value 59.098253
## final value 59.098253
## stopped after 100 iterations
## # weights: 109
## initial value 108.904984
## iter 10 value 45.818045
## iter 20 value 27.374777
## iter 30 value 15.843102
## iter 40 value 13.368509
## iter 50 value 11.419267
## iter 60 value 11.068134
## iter 70 value 11.043306
## iter 80 value 11.002143
## iter 90 value 10.786482
## iter 100 value 10.781322
## final value 10.781322
## stopped after 100 iterations
## # weights: 179
## initial value 105.547978
## iter 10 value 24.652374
## iter 20 value 1.721313
## iter 30 value 0.317535
## iter 40 value 0.284089
## iter 50 value 0.245125
## iter 60 value 0.226665
## iter 70 value 0.197506
## iter 80 value 0.179818
## iter 90 value 0.160245
## iter 100 value 0.142763
## final value 0.142763
## stopped after 100 iterations
## # weights: 39
## initial value 120.645515
## iter 10 value 61.637485
## iter 20 value 53.123121
## iter 30 value 48.896885
## iter
       40 value 48.715487
## iter 50 value 48.714829
## iter 60 value 48.714706
## iter 60 value 48.714706
## iter 60 value 48.714706
## final value 48.714706
## converged
```

```
## # weights: 109
## initial value 106.861698
## iter 10 value 22.601379
## iter 20 value 10.206481
## iter 30 value 9.499900
## iter 40 value 9.433939
## iter 50 value 9.424377
## iter 60 value 9.366633
## iter 70 value 9.326111
## iter 80 value 9.274295
## iter 90 value 9.232818
## iter 100 value 9.212769
## final value 9.212769
## stopped after 100 iterations
## # weights: 179
## initial value 98.938722
## iter 10 value 17.703871
## iter 20 value 9.831781
## iter 30 value 6.229459
## iter 40 value 5.584642
## iter 50 value 5.555802
## iter 60 value 3.320211
## iter 70 value 3.293532
## iter 80 value 3.166623
## iter 90 value 3.147293
## iter 100 value 3.143740
## final value 3.143740
## stopped after 100 iterations
## # weights: 39
## initial value 109.335008
## iter 10 value 79.676130
## iter 20 value 66.874308
## iter 30 value 62.342386
## iter 40 value 61.042051
## iter 50 value 59.267353
## iter 60 value 58.817616
## iter 70 value 58.812712
## final value 58.812628
## converged
## # weights: 109
## initial value 145.111699
## iter 10 value 86.303677
## iter 20 value 48.857213
## iter 30 value 25.180625
## iter 40 value 17.647507
## iter 50 value 16.928808
## iter 60 value 16.404409
## iter 70 value 16.274914
## iter 80 value 16.247700
## iter 90 value 16.237129
## iter 100 value 16.233919
## final value 16.233919
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 120.650479
## iter 10 value 35.934519
## iter 20 value 18.102816
## iter 30 value 14.347485
## iter 40 value 12.788833
## iter 50 value 12.095441
## iter 60 value 11.909278
## iter 70 value 11.804437
## iter 80 value 11.793172
## iter 90 value 11.792811
## iter 100 value 11.792798
## final value 11.792798
## stopped after 100 iterations
## # weights: 39
## initial value 105.058638
## iter 10 value 78.685765
## iter 20 value 61.916126
## iter 30 value 59.164554
## iter 40 value 59.156549
## iter 50 value 59.109079
## iter 60 value 58.005356
## iter 70 value 57.790018
## iter 80 value 57.758283
## iter 90 value 57.725367
## iter 100 value 55.973502
## final value 55.973502
## stopped after 100 iterations
## # weights: 109
## initial value 100.919183
## iter 10 value 46.992471
## iter 20 value 8.827525
## iter 30 value 8.555688
## iter 40 value 8.535719
## iter 50 value 8.516062
## iter 60 value 8.502384
## iter 70 value 8.491016
## iter 80 value 8.483104
## iter 90 value 8.477313
## iter 100 value 8.471856
## final value 8.471856
## stopped after 100 iterations
## # weights: 179
## initial value 103.582615
## iter 10 value 28.582405
## iter 20 value 9.640441
## iter 30 value 6.065557
## iter 40 value 5.002357
## iter 50 value 4.290514
## iter 60 value 4.179204
## iter 70 value 4.111332
## iter 80 value 4.068451
## iter 90 value 4.041857
## iter 100 value 4.032621
## final value 4.032621
```

```
## stopped after 100 iterations
## # weights: 39
## initial value 101.782962
## iter 10 value 63.538181
## iter 20 value 57.534621
## iter 30 value 52.627096
## iter 40 value 51.794453
## iter 50 value 51.261330
## iter 60 value 51.258879
## iter 70 value 50.643786
## iter 80 value 50.086664
## iter 90 value 50.083176
## iter 100 value 50.080439
## final value 50.080439
## stopped after 100 iterations
## # weights: 109
## initial value 129.743763
## iter 10 value 37.640433
## iter 20 value 25.819287
## iter 30 value 21.980830
## iter 40 value 20.449987
## iter 50 value 19.446402
## iter 60 value 18.715112
## iter 70 value 17.642275
## iter 80 value 14.590593
## iter 90 value 14.309223
## iter 100 value 14.074484
## final value 14.074484
## stopped after 100 iterations
## # weights: 179
## initial value 105.748942
## iter 10 value 52.283717
## iter 20 value 12.942157
## iter 30 value 9.968846
## iter 40 value 8.587177
## iter 50 value 7.526196
## iter 60 value 5.213605
## iter 70 value 1.773347
## iter 80 value 0.122413
## iter 90 value 0.045164
## iter 100 value 0.016591
## final value 0.016591
## stopped after 100 iterations
## # weights: 39
## initial value 107.914127
## iter 10 value 75.450615
## iter 20 value 62.906812
## iter 30 value 60.135136
## iter 40 value 57.707300
## iter 50 value 57.538580
## iter 60 value 57.025217
## iter 70 value 56.121082
## iter 80 value 55.252240
## iter 90 value 54.211762
```

```
## iter 100 value 54.207165
## final value 54.207165
## stopped after 100 iterations
## # weights: 109
## initial value 112.021578
## iter 10 value 53.765593
## iter 20 value 29.344986
## iter 30 value 19.989816
## iter 40 value 16.226424
## iter 50 value 15.447024
## iter 60 value 15.249666
## iter 70 value 15.247884
## iter 80 value 15.247825
## final value 15.247824
## converged
## # weights: 179
## initial value 100.916432
## iter 10 value 29.115380
## iter 20 value 15.897803
## iter 30 value 13.455035
## iter 40 value 12.771820
## iter 50 value 12.035579
## iter 60 value 11.874778
## iter 70 value 11.860929
## iter 80 value 11.860628
## final value 11.860625
## converged
## # weights: 39
## initial value 109.489056
## iter 10 value 79.237084
## iter 20 value 56.231497
## iter 30 value 53.489126
## iter 40 value 51.955535
## iter 50 value 51.353479
## iter 60 value 50.026939
## iter 70 value 49.197584
## iter 80 value 49.014764
## iter 90 value 48.677395
## iter 100 value 47.770908
## final value 47.770908
## stopped after 100 iterations
## # weights: 109
## initial value 114.920171
## iter 10 value 35.067807
## iter 20 value 22.209082
## iter 30 value 21.535110
## iter 40 value 18.112931
## iter 50 value 17.587471
## iter 60 value 17.548903
## iter 70 value 17.539604
## iter 80 value 17.525455
## iter 90 value 17.512473
## iter 100 value 17.500743
## final value 17.500743
```

```
## stopped after 100 iterations
## # weights: 179
## initial value 114.293964
## iter 10 value 20.980641
## iter 20 value 7.444786
## iter 30 value 5.627587
## iter 40 value 5.029160
## iter 50 value 2.918532
## iter 60 value 2.632264
## iter 70 value 2.579031
## iter 80 value 0.671726
## iter 90 value 0.321176
## iter 100 value 0.287104
## final value 0.287104
## stopped after 100 iterations
## # weights: 39
## initial value 104.998400
## iter 10 value 65.387418
## iter 20 value 55.477505
## iter 30 value 52.459363
## iter 40 value 49.830521
## iter 50 value 47.659870
## iter 60 value 46.665998
## iter 70 value 46.366990
## iter 80 value 46.350113
## iter 90 value 46.330118
## iter 100 value 45.724295
## final value 45.724295
## stopped after 100 iterations
## # weights: 109
## initial value 102.743289
## iter 10 value 30.690013
## iter 20 value 20.372395
## iter 30 value 18.373636
## iter 40 value 15.410071
## iter 50 value 14.657767
## iter 60 value 8.497372
## iter 70 value 7.810610
## iter 80 value 7.086149
## iter 90 value 7.039267
## iter 100 value 7.030767
## final value 7.030767
## stopped after 100 iterations
## # weights: 179
## initial value 113.419119
## iter 10 value 15.849269
## iter 20 value 0.460053
## iter 30 value 0.002249
## final value 0.000091
## converged
## # weights: 39
## initial value 109.775895
## iter 10 value 82.341835
## iter 20 value 72.463490
```

```
## iter 30 value 68.560593
## iter 40 value 63.323000
## iter 50 value 59.900773
## iter 60 value 56.980065
## iter 70 value 56.732388
## final value 56.732046
## converged
## # weights: 109
## initial value 107.624634
## iter 10 value 60.097713
## iter 20 value 31.943296
## iter 30 value 20.570084
## iter 40 value 17.380692
## iter 50 value 17.148130
## iter 60 value 17.143968
## final value 17.143850
## converged
## # weights: 179
## initial value 113.279239
## iter 10 value 31.886952
## iter 20 value 17.598580
## iter 30 value 14.606289
## iter 40 value 12.420506
## iter 50 value 12.121150
## iter 60 value 11.972272
## iter 70 value 11.961330
## iter 80 value 11.960746
## iter 90 value 11.960684
## final value 11.960682
## converged
## # weights: 39
## initial value 104.642163
## iter 10 value 80.124375
## iter 20 value 68.035844
## iter 30 value 63.527875
## iter 40 value 61.474930
## iter 50 value 58.870838
## iter 60 value 58.119046
## iter 70 value 58.111727
## iter 80 value 58.081951
## iter 90 value 58.059267
## iter 100 value 58.056581
## final value 58.056581
## stopped after 100 iterations
## # weights: 109
## initial value 118.713302
## iter 10 value 35.495847
## iter 20 value 23.053432
## iter 30 value 17.678575
## iter 40 value 13.383633
## iter 50 value 6.403850
## iter 60 value 5.774992
## iter 70 value 5.147307
## iter 80 value 5.109554
```

```
## iter 90 value 5.074664
## iter 100 value 5.051312
## final value 5.051312
## stopped after 100 iterations
## # weights: 179
## initial value 136.136052
## iter 10 value 46.161538
## iter 20 value 32.615980
## iter 30 value 28.986056
## iter 40 value 27.673266
## iter 50 value 24.785597
## iter 60 value 22.698249
## iter 70 value 21.600201
## iter 80 value 21.481798
## iter 90 value 21.318503
## iter 100 value 21.205923
## final value 21.205923
## stopped after 100 iterations
## # weights: 39
## initial value 113.350016
## iter 10 value 70.729232
## iter 20 value 62.349199
## iter 30 value 59.609717
## iter 40 value 59.014316
## iter 50 value 57.435928
## iter 60 value 54.055970
## iter 70 value 53.599324
## iter 80 value 53.577901
## iter 90 value 53.560959
## iter 100 value 53.490420
## final value 53.490420
## stopped after 100 iterations
## # weights: 109
## initial value 104.944288
## iter 10 value 30.882796
## iter 20 value 11.783503
## iter 30 value 8.486227
## iter 40 value 7.040081
## iter 50 value 7.038859
## iter 60 value 6.989847
## iter 70 value 6.968175
## iter 80 value 6.968151
## iter 90 value 6.968106
## iter 100 value 6.967826
## final value 6.967826
## stopped after 100 iterations
## # weights: 179
## initial value 112.800098
## iter 10 value 26.388570
## iter 20 value 5.599708
## iter 30 value 1.552520
## iter 40 value 1.399780
## iter 50 value 1.386405
## iter 60 value 1.386349
```

```
## iter 70 value 1.386321
## iter 80 value 1.386313
## final value 1.386312
## converged
## # weights: 39
## initial value 112.669379
## iter 10 value 93.971481
## iter 20 value 77.797487
## iter 30 value 66.077188
## iter 40 value 62.868417
## iter 50 value 60.977595
## iter 60 value 59.630071
## iter 70 value 59.126414
## iter 80 value 58.868170
## iter 90 value 58.682021
## iter 100 value 58.462834
## final value 58.462834
## stopped after 100 iterations
## # weights: 109
## initial value 97.018452
## iter 10 value 55.793926
## iter 20 value 30.563067
## iter 30 value 19.111092
## iter 40 value 17.362120
## iter 50 value 17.196488
## iter 60 value 17.187790
## iter 70 value 17.187614
## final value 17.187614
## converged
## # weights: 179
## initial value 110.573094
## iter 10 value 41.750471
## iter 20 value 22.404796
## iter 30 value 14.663886
## iter 40 value 12.767774
## iter 50 value 12.070395
## iter 60 value 11.977927
## iter 70 value 11.968827
## iter 80 value 11.968221
## final value 11.968218
## converged
## # weights: 39
## initial value 102.668271
## iter 10 value 76.222737
## iter 20 value 71.120357
## iter 30 value 70.002522
## iter 40 value 66.179119
## iter 50 value 63.429924
## iter 60 value 62.301242
## iter 70 value 61.205275
## iter 80 value 59.872349
## iter 90 value 59.411723
## iter 100 value 59.345851
## final value 59.345851
```

```
## stopped after 100 iterations
## # weights: 109
## initial value 106.120849
## iter 10 value 32.403149
## iter 20 value 18.474979
## iter 30 value 13.088859
## iter 40 value 11.447226
## iter 50 value 8.970656
## iter 60 value 8.864178
## iter 70 value 8.804195
## iter 80 value 8.642453
## iter 90 value 7.001615
## iter 100 value 4.944231
## final value 4.944231
## stopped after 100 iterations
## # weights: 179
## initial value 106.316868
## iter 10 value 23.770381
## iter 20 value 11.783803
## iter 30 value 9.676904
## iter 40 value 8.118733
## iter 50 value 7.545715
## iter 60 value 7.511021
## iter 70 value 7.446200
## iter 80 value 7.423943
## iter 90 value 7.412165
## iter 100 value 7.341176
## final value 7.341176
## stopped after 100 iterations
## # weights: 39
## initial value 105.344985
## iter 10 value 60.114813
## iter 20 value 52.056150
## iter 30 value 48.603365
## iter 40 value 47.673760
## iter 50 value 47.586850
## iter 60 value 47.504725
## iter 70 value 47.190988
## iter 80 value 46.176918
## iter 90 value 46.047269
## iter 100 value 45.007590
## final value 45.007590
## stopped after 100 iterations
## # weights: 109
## initial value 100.276078
## iter 10 value 43.295442
## iter 20 value 25.915349
## iter 30 value 23.194969
## iter 40 value 21.158903
## iter 50 value 19.133764
## iter 60 value 17.733614
## iter 70 value 14.300281
## iter 80 value 7.969243
## iter 90 value 6.214430
```

```
## iter 100 value 5.944171
## final value 5.944171
## stopped after 100 iterations
## # weights: 179
## initial value 97.761042
## iter 10 value 17.154700
## iter 20 value 7.829067
## iter 30 value 3.227378
## iter 40 value 2.747947
## iter 50 value 0.046889
## iter 60 value 0.015875
## iter 70 value 0.006553
## iter 80 value 0.003962
## iter 90 value 0.002762
## iter 100 value 0.001081
## final value 0.001081
## stopped after 100 iterations
## # weights: 39
## initial value 101.671479
## iter 10 value 70.382295
## iter 20 value 62.066615
## iter 30 value 59.764257
## iter 40 value 59.088796
## iter 50 value 58.916054
## iter 60 value 58.834693
## iter 70 value 58.832265
## final value 58.832239
## converged
## # weights: 109
## initial value 107.998837
## iter 10 value 60.476972
## iter 20 value 45.654734
## iter 30 value 31.672688
## iter 40 value 20.794969
## iter 50 value 18.011520
## iter 60 value 16.053407
## iter 70 value 15.741586
## iter 80 value 15.728957
## iter 90 value 15.728822
## final value 15.728820
## converged
## # weights: 179
## initial value 123.720915
## iter 10 value 40.468597
## iter 20 value 25.862052
## iter 30 value 17.573230
## iter 40 value 14.579384
## iter 50 value 13.731864
## iter 60 value 13.452821
## iter 70 value 12.803875
## iter 80 value 12.559815
## iter 90 value 12.364204
## iter 100 value 12.345274
## final value 12.345274
```

```
## stopped after 100 iterations
## # weights: 39
## initial value 104.100554
## iter 10 value 66.890695
## iter 20 value 56.156042
## iter 30 value 51.173877
## iter 40 value 46.797223
## iter 50 value 44.345010
## iter 60 value 44.239260
## iter 70 value 43.372173
## iter 80 value 43.302501
## iter 90 value 42.916799
## iter 100 value 42.882276
## final value 42.882276
## stopped after 100 iterations
## # weights: 109
## initial value 110.092708
## iter 10 value 44.928603
## iter 20 value 32.047071
## iter 30 value 28.337515
## iter 40 value 27.179591
## iter 50 value 26.412786
## iter 60 value 24.532276
## iter 70 value 24.468793
## iter 80 value 22.672416
## iter 90 value 21.936651
## iter 100 value 21.866777
## final value 21.866777
## stopped after 100 iterations
## # weights: 179
## initial value 117.480129
## iter 10 value 26.056083
## iter 20 value 14.870582
## iter 30 value 10.580739
## iter 40 value 10.144600
## iter 50 value 9.985516
## iter 60 value 9.963580
## iter 70 value 9.921848
## iter 80 value 9.782611
## iter 90 value 8.005353
## iter 100 value 6.558581
## final value 6.558581
## stopped after 100 iterations
## # weights: 39
## initial value 98.626120
## iter 10 value 79.910268
## iter 20 value 68.317892
## iter 30 value 65.890290
## iter 40 value 62.895747
## iter 50 value 59.481191
## iter 60 value 57.686874
## iter 70 value 57.410912
## iter 80 value 57.292111
## iter 90 value 57.276122
```

```
## iter 100 value 57.271064
## final value 57.271064
## stopped after 100 iterations
## # weights: 109
## initial value 144.941972
## iter 10 value 77.123106
## iter 20 value 24.284521
## iter 30 value 19.662510
## iter 40 value 18.675218
## iter 50 value 15.764644
## iter 60 value 15.015242
## iter 70 value 14.918402
## iter 80 value 14.830300
## iter 90 value 14.784867
## iter 100 value 14.764151
## final value 14.764151
## stopped after 100 iterations
## # weights: 179
## initial value 104.057813
## iter 10 value 15.322900
## iter 20 value 2.267731
## iter 30 value 1.440125
## iter 40 value 1.386743
## final value 1.386299
## converged
## # weights: 39
## initial value 103.571185
## iter 10 value 70.173304
## iter 20 value 63.246199
## iter 30 value 60.504459
## iter 40 value 60.162146
## iter 50 value 59.895116
## iter 60 value 59.401383
## iter 70 value 58.845643
## final value 58.844214
## converged
## # weights: 109
## initial value 114.942385
## iter 10 value 47.235792
## iter 20 value 31.410793
## iter 30 value 23.149571
## iter 40 value 18.383528
## iter 50 value 16.339475
## iter 60 value 16.044604
## iter 70 value 15.984868
## iter 80 value 15.739692
## iter 90 value 15.726611
## iter 100 value 15.726518
## final value 15.726518
## stopped after 100 iterations
## # weights: 179
## initial value 114.203579
## iter 10 value 30.198653
## iter 20 value 17.327694
```

```
## iter 30 value 13.825932
## iter 40 value 12.743544
## iter 50 value 12.296795
## iter 60 value 12.202201
## iter 70 value 12.167712
## iter 80 value 12.164789
## iter 90 value 12.164686
## final value 12.164678
## converged
## # weights: 39
## initial value 105.847642
## iter 10 value 81.982261
## iter 20 value 64.064822
## iter 30 value 63.955642
## iter 40 value 63.954360
## iter 50 value 63.948536
## iter 60 value 63.917226
## iter 70 value 59.457800
## iter 80 value 58.373205
## iter 90 value 58.039035
## iter 100 value 57.125163
## final value 57.125163
## stopped after 100 iterations
## # weights: 109
## initial value 106.936039
## iter 10 value 37.802223
## iter 20 value 19.992552
## iter 30 value 11.378917
## iter 40 value 10.923480
## iter 50 value 6.559896
## iter 60 value 5.622549
## iter 70 value 5.569587
## iter 80 value 5.503688
## iter 90 value 0.296169
## iter 100 value 0.243515
## final value 0.243515
## stopped after 100 iterations
## # weights: 179
## initial value 126.812508
## iter 10 value 19.841783
## iter 20 value 0.682379
## iter 30 value 0.171284
## iter 40 value 0.151431
## iter 50 value 0.141022
## iter 60 value 0.126759
## iter 70 value 0.120292
## iter 80 value 0.113210
## iter 90 value 0.103905
## iter 100 value 0.100037
## final value 0.100037
## stopped after 100 iterations
## # weights: 39
## initial value 102.331851
## iter 10 value 77.118741
```

```
## iter 20 value 75.645531
## iter 30 value 74.549658
## iter 40 value 73.721185
## iter 50 value 73.186026
## iter 60 value 71.937590
## iter 70 value 68.867118
## iter 80 value 65.978751
## iter 90 value 65.755511
## iter 100 value 65.744870
## final value 65.744870
## stopped after 100 iterations
## # weights: 109
## initial value 108.028265
## iter 10 value 49.317289
## iter 20 value 35.085843
## iter 30 value 27.572547
## iter 40 value 24.487502
## iter 50 value 19.450828
## iter 60 value 13.490997
## iter 70 value 12.074980
## iter 80 value 11.645751
## iter 90 value 11.542920
## iter 100 value 11.367370
## final value 11.367370
## stopped after 100 iterations
## # weights: 179
## initial value 104.751551
## iter 10 value 29.066701
## iter 20 value 7.801708
## iter 30 value 6.297922
## iter 40 value 6.172064
## iter 50 value 5.233750
## iter 60 value 4.792694
## iter 70 value 4.782133
## iter 80 value 4.781743
## iter 90 value 4.781042
## iter 100 value 4.780859
## final value 4.780859
## stopped after 100 iterations
## # weights: 39
## initial value 103.743076
## iter 10 value 76.126151
## iter 20 value 63.917704
## iter 30 value 60.665179
## iter 40 value 59.454249
## iter 50 value 57.937774
## iter 60 value 57.913853
## final value 57.913845
## converged
## # weights: 109
## initial value 104.302171
## iter 10 value 48.039656
## iter 20 value 31.112935
## iter 30 value 23.007944
```

```
## iter 40 value 17.853326
## iter 50 value 17.175604
## iter 60 value 17.133322
## iter 70 value 17.132918
## final value 17.132917
## converged
## # weights: 179
## initial value 127.097661
## iter 10 value 48.790091
## iter 20 value 24.244372
## iter 30 value 14.552219
## iter 40 value 12.402271
## iter 50 value 12.250747
## iter 60 value 12.235741
## iter 70 value 12.233540
## iter 80 value 12.233358
## final value 12.233354
## converged
## # weights: 39
## initial value 101.214908
## iter 10 value 57.856495
## iter 20 value 49.857730
## iter 30 value 45.087983
## iter 40 value 42.571795
## iter 50 value 41.809917
## iter 60 value 41.786536
## iter 70 value 41.766615
## iter 80 value 41.749285
## iter 90 value 41.726401
## iter 100 value 41.717331
## final value 41.717331
## stopped after 100 iterations
## # weights: 109
## initial value 114.460738
## iter 10 value 26.376054
## iter 20 value 14.035897
## iter 30 value 5.341266
## iter 40 value 4.259323
## iter 50 value 4.086854
## iter 60 value 1.732333
## iter 70 value 0.225387
## iter 80 value 0.194067
## iter 90 value 0.178984
## iter 100 value 0.164494
## final value 0.164494
## stopped after 100 iterations
## # weights: 179
## initial value 100.455778
## iter 10 value 19.326568
## iter 20 value 2.203497
## iter 30 value 2.046491
## iter 40 value 2.032757
## iter 50 value 1.496368
## iter 60 value 0.157461
```

```
## iter 70 value 0.113411
## iter 80 value 0.107761
## iter 90 value 0.104948
## iter 100 value 0.098663
## final value 0.098663
## stopped after 100 iterations
## # weights: 39
## initial value 112.864063
## iter 10 value 62.462087
## iter 20 value 55.510725
## iter 30 value 52.142977
## iter 40 value 50.505459
## iter 50 value 48.869881
## iter 60 value 48.618819
## iter 70 value 47.891296
## iter 80 value 46.751051
## iter 90 value 45.436151
## iter 100 value 45.029035
## final value 45.029035
## stopped after 100 iterations
## # weights: 109
## initial value 118.942175
## iter 10 value 29.365561
## iter 20 value 13.913647
## iter 30 value 11.398030
## iter 40 value 9.533676
## iter 50 value 9.114137
## iter 60 value 9.023862
## iter 70 value 8.885840
## iter 80 value 8.826751
## iter 90 value 8.616269
## iter 100 value 8.140201
## final value 8.140201
## stopped after 100 iterations
## # weights: 179
## initial value 114.418584
## iter 10 value 27.335317
## iter 20 value 1.596331
## iter 30 value 0.059301
## iter 40 value 0.003534
## iter 50 value 0.000353
## iter 60 value 0.000242
## final value 0.000083
## converged
## # weights: 39
## initial value 100.799018
## iter 10 value 69.652839
## iter 20 value 62.652819
## iter 30 value 60.119425
## iter 40 value 59.945466
## iter 50 value 59.475821
## iter 60 value 58.757868
## iter 70 value 58.432540
## iter 80 value 58.424767
```

```
## final value 58.424747
## converged
## # weights: 109
## initial value 103.169973
## iter 10 value 55.363697
## iter 20 value 28.392282
## iter 30 value 20.062653
## iter 40 value 18.759972
## iter 50 value 18.545062
## iter 60 value 18.255249
## iter 70 value 17.149311
## iter 80 value 16.855155
## iter 90 value 16.027953
## iter 100 value 15.929266
## final value 15.929266
## stopped after 100 iterations
## # weights: 179
## initial value 111.943346
## iter 10 value 49.714763
## iter 20 value 28.826984
## iter 30 value 16.145633
## iter 40 value 13.411749
## iter 50 value 12.947011
## iter 60 value 12.718059
## iter 70 value 12.537811
## iter 80 value 12.517755
## iter 90 value 12.516124
## iter 100 value 12.515708
## final value 12.515708
## stopped after 100 iterations
## # weights: 39
## initial value 103.759376
## iter 10 value 66.053519
## iter 20 value 57.723898
## iter 30 value 57.338882
## iter 40 value 56.280778
## iter 50 value 56.245932
## iter 60 value 56.226694
## iter 70 value 55.121251
## iter 80 value 53.917501
## iter 90 value 52.696280
## iter 100 value 52.632124
## final value 52.632124
## stopped after 100 iterations
## # weights: 109
## initial value 118.250672
## iter 10 value 33.272180
## iter 20 value 19.501960
## iter 30 value 14.308341
## iter 40 value 12.284254
## iter 50 value 10.582490
## iter 60 value 10.508805
## iter 70 value 10.421194
## iter 80 value 8.035947
```

```
## iter 90 value 7.806725
## iter 100 value 7.793478
## final value 7.793478
## stopped after 100 iterations
## # weights: 179
## initial value 148.202269
## iter 10 value 56.558341
## iter 20 value 26.479386
## iter 30 value 19.184836
## iter 40 value 16.191693
## iter 50 value 13.420992
## iter 60 value 12.697412
## iter 70 value 11.523216
## iter 80 value 9.176712
## iter 90 value 7.463419
## iter 100 value 5.642809
## final value 5.642809
## stopped after 100 iterations
## # weights: 39
## initial value 102.418690
## iter 10 value 71.055217
## iter 20 value 67.158590
## iter 30 value 66.769307
## iter 40 value 64.429575
## iter 50 value 58.258543
## iter 60 value 57.853237
## iter 70 value 56.986039
## iter 80 value 55.328486
## iter 90 value 54.955551
## iter 100 value 54.930910
## final value 54.930910
## stopped after 100 iterations
## # weights: 109
## initial value 107.352464
## iter 10 value 47.984754
## iter 20 value 33.645589
## iter 30 value 16.722743
## iter 40 value 14.057835
## iter 50 value 13.439273
## iter 60 value 12.480599
## iter 70 value 8.984974
## iter 80 value 7.746744
## iter 90 value 5.408461
## iter 100 value 5.285482
## final value 5.285482
## stopped after 100 iterations
## # weights: 179
## initial value 108.400527
## iter 10 value 29.485720
## iter 20 value 13.479316
## iter 30 value 6.086225
## iter 40 value 2.366902
## iter 50 value 1.943584
## iter 60 value 1.925195
```

```
## iter 70 value 1.918077
## iter 80 value 1.911607
## iter 90 value 1.910946
## iter 100 value 1.910351
## final value 1.910351
## stopped after 100 iterations
## # weights: 39
## initial value 100.498558
## iter 10 value 72.863891
## iter 20 value 65.038302
## iter 30 value 62.552814
## iter 40 value 59.521106
## iter 50 value 58.275421
## iter 60 value 57.626380
## iter 70 value 57.549974
## final value 57.549638
## converged
## # weights: 109
## initial value 116.104385
## iter 10 value 37.909599
## iter 20 value 26.356494
## iter 30 value 21.022665
## iter 40 value 17.201416
## iter 50 value 16.109427
## iter 60 value 16.071303
## final value 16.071091
## converged
## # weights: 179
## initial value 104.956466
## iter 10 value 36.318322
## iter 20 value 20.596644
## iter 30 value 14.937412
## iter 40 value 13.056861
## iter 50 value 12.407184
## iter 60 value 12.159840
## iter 70 value 11.921136
## iter 80 value 11.794267
## iter 90 value 11.777995
## iter 100 value 11.777649
## final value 11.777649
## stopped after 100 iterations
## # weights: 39
## initial value 101.454706
## iter 10 value 83.182233
## iter 20 value 65.771691
## iter 30 value 65.682662
## iter 40 value 64.496918
## iter 50 value 63.703936
## iter 60 value 63.678678
## iter 70 value 63.670216
## iter 80 value 63.664699
## iter 90 value 63.660486
## iter 100 value 63.142506
## final value 63.142506
```

```
## stopped after 100 iterations
## # weights: 109
## initial value 107.794895
## iter 10 value 39.211013
## iter 20 value 17.750283
## iter 30 value 14.292229
## iter 40 value 13.848316
## iter 50 value 13.789629
## iter 60 value 13.739810
## iter 70 value 13.630570
## iter 80 value 13.455568
## iter 90 value 10.859825
## iter 100 value 7.800682
## final value 7.800682
## stopped after 100 iterations
## # weights: 179
## initial value 103.415146
## iter 10 value 24.953071
## iter 20 value 6.832968
## iter 30 value 4.254160
## iter 40 value 3.998683
## iter 50 value 3.900428
## iter 60 value 3.819728
## iter 70 value 3.667355
## iter 80 value 3.623497
## iter 90 value 3.448991
## iter 100 value 3.207144
## final value 3.207144
## stopped after 100 iterations
## # weights: 39
## initial value 103.877719
## iter 10 value 64.147785
## iter 20 value 51.995981
## iter 30 value 51.241590
## iter 40 value 51.212248
## iter 50 value 51.176162
## iter 60 value 50.635572
## iter 70 value 50.487205
## final value 50.487088
## converged
## # weights: 109
## initial value 104.283856
## iter 10 value 32.515271
## iter 20 value 12.151861
## iter 30 value 9.025931
## iter 40 value 8.576796
## iter 50 value 8.055914
## iter
       60 value 7.802891
## iter 70 value 7.287348
## iter 80 value 7.229932
## iter 90 value 7.218098
## iter 100 value 7.209661
## final value 7.209661
## stopped after 100 iterations
```

```
## # weights: 179
## initial value 144.903137
## iter 10 value 44.076925
## iter 20 value 17.596644
## iter 30 value 15.087289
## iter 40 value 14.563269
## iter 50 value 10.337919
## iter 60 value 6.468374
## iter 70 value 5.629895
## iter 80 value 5.545960
## iter 90 value 5.545622
## iter 100 value 5.545402
## final value 5.545402
## stopped after 100 iterations
## # weights: 39
## initial value 101.783259
## iter 10 value 83.271780
## iter 20 value 63.992951
## iter 30 value 58.957729
## iter 40 value 58.767410
## iter 50 value 58.764643
## final value 58.764475
## converged
## # weights: 109
## initial value 100.139730
## iter 10 value 43.931145
## iter 20 value 23.671386
## iter 30 value 17.530501
## iter 40 value 16.050389
## iter 50 value 15.651794
## iter 60 value 15.617358
## final value 15.617284
## converged
## # weights: 179
## initial value 102.106877
## iter 10 value 34.941892
## iter 20 value 17.666120
## iter 30 value 13.932899
## iter 40 value 13.055006
## iter 50 value 12.705135
## iter 60 value 12.620219
## iter 70 value 12.591342
## iter 80 value 12.582013
## iter 90 value 12.581567
## final value 12.581563
## converged
## # weights: 39
## initial value 103.317784
## iter 10 value 76.721712
## iter 20 value 61.428351
## iter 30 value 58.273222
## iter 40 value 56.857703
## iter 50 value 54.590488
## iter 60 value 53.882256
```

```
## iter 70 value 53.662349
## iter 80 value 53.332954
## iter 90 value 52.778608
## iter 100 value 52.734339
## final value 52.734339
## stopped after 100 iterations
## # weights: 109
## initial value 129.941677
## iter 10 value 60.195476
## iter 20 value 27.532384
## iter 30 value 10.376122
## iter 40 value 5.788687
## iter 50 value 4.397872
## iter 60 value 3.877862
## iter 70 value 3.761546
## iter 80 value 3.743188
## iter 90 value 3.721730
## iter 100 value 3.685009
## final value 3.685009
## stopped after 100 iterations
## # weights: 179
## initial value 111.463558
## iter 10 value 18.731761
## iter 20 value 7.272737
## iter 30 value 4.765955
## iter 40 value 3.170226
## iter 50 value 2.994089
## iter 60 value 2.974798
## iter 70 value 2.967666
## iter 80 value 1.626395
## iter 90 value 1.569963
## iter 100 value 0.207179
## final value 0.207179
## stopped after 100 iterations
## # weights: 39
## initial value 100.126293
## iter 10 value 68.405075
## iter 20 value 55.765061
## iter 30 value 55.032503
## iter 40 value 51.629556
## iter 50 value 47.032138
## iter 60 value 42.576347
## iter 70 value 41.546640
## iter 80 value 41.468662
## iter 90 value 41.435321
## iter 100 value 41.415153
## final value 41.415153
## stopped after 100 iterations
## # weights: 109
## initial value 106.303020
## iter 10 value 48.377013
## iter 20 value 22.074398
## iter 30 value 17.603824
## iter 40 value 17.158453
```

```
## iter 50 value 17.116159
## iter 60 value 17.102256
## iter 70 value 17.092050
## iter 80 value 17.088361
## iter 90 value 16.883673
## iter 100 value 16.865443
## final value 16.865443
## stopped after 100 iterations
## # weights: 179
## initial value 105.622444
## iter 10 value 22.276828
## iter 20 value 11.035662
## iter 30 value 7.575096
## iter 40 value 7.254733
## iter 50 value 7.188539
## iter 60 value 6.943245
## iter 70 value 5.324949
## iter 80 value 2.551888
## iter 90 value 0.290725
## iter 100 value 0.036278
## final value 0.036278
## stopped after 100 iterations
## # weights: 39
## initial value 103.204839
## iter 10 value 70.888866
## iter 20 value 64.854832
## iter 30 value 60.505710
## iter 40 value 59.323505
## iter 50 value 59.110164
## iter 60 value 59.051548
## iter 70 value 59.035015
## iter 70 value 59.035015
## iter 70 value 59.035015
## final value 59.035015
## converged
## # weights: 109
## initial value 128.849614
## iter 10 value 61.938800
## iter 20 value 36.552475
## iter 30 value 22.405648
## iter 40 value 17.283416
## iter 50 value 16.586665
## iter 60 value 16.515742
## iter 70 value 16.477041
## iter 80 value 16.466998
## iter 90 value 16.466202
## final value 16.466183
## converged
## # weights: 179
## initial value 110.271610
## iter 10 value 50.180700
## iter 20 value 31.554709
## iter 30 value 17.315662
## iter 40 value 12.944208
```

```
## iter 50 value 12.251342
## iter 60 value 12.159385
## iter 70 value 12.150538
## iter 80 value 12.149933
## iter 90 value 12.149916
## final value 12.149915
## converged
## # weights: 39
## initial value 116.089303
## iter 10 value 67.748410
## iter 20 value 61.994831
## iter 30 value 59.615635
## iter 40 value 58.915516
## iter 50 value 58.520884
## iter 60 value 58.472508
## iter 70 value 58.370575
## iter 80 value 58.324050
## iter 90 value 58.128299
## iter 100 value 57.032730
## final value 57.032730
## stopped after 100 iterations
## # weights: 109
## initial value 115.247747
## iter 10 value 54.195659
## iter 20 value 26.147525
## iter 30 value 11.149365
## iter 40 value 7.042787
## iter 50 value 5.315439
## iter 60 value 5.068685
## iter 70 value 2.995200
## iter 80 value 2.918379
## iter 90 value 0.724086
## iter 100 value 0.227656
## final value 0.227656
## stopped after 100 iterations
## # weights: 179
## initial value 120.202035
## iter 10 value 24.849869
## iter 20 value 5.699435
## iter 30 value 2.140725
## iter 40 value 0.298732
## iter 50 value 0.233885
## iter 60 value 0.218877
## iter 70 value 0.201590
## iter 80 value 0.185795
## iter 90 value 0.166678
## iter 100 value 0.154123
## final value 0.154123
## stopped after 100 iterations
## # weights: 39
## initial value 111.245106
## iter 10 value 66.224925
## iter 20 value 58.223940
## iter 30 value 56.860679
```

```
## iter 40 value 56.397640
## iter 50 value 55.970450
## iter 60 value 55.966223
## iter 70 value 55.957568
## iter 80 value 55.952434
## iter 90 value 55.949370
## iter 100 value 54.594722
## final value 54.594722
## stopped after 100 iterations
## # weights: 109
## initial value 123.392432
## iter 10 value 45.416071
## iter 20 value 36.055904
## iter 30 value 31.993854
## iter 40 value 30.507479
## iter 50 value 29.519126
## iter 60 value 28.094554
## iter 70 value 25.689313
## iter 80 value 25.554940
## iter 90 value 24.832351
## iter 100 value 23.627305
## final value 23.627305
## stopped after 100 iterations
## # weights: 179
## initial value 105.709171
## iter 10 value 43.103155
## iter 20 value 14.170640
## iter 30 value 12.900424
## iter 40 value 12.833486
## iter 50 value 12.823268
## iter 60 value 12.821913
## final value 12.821723
## converged
## # weights: 39
## initial value 105.265412
## iter 10 value 81.937971
## iter 20 value 70.282207
## iter 30 value 62.482980
## iter 40 value 59.963098
## iter 50 value 59.667168
## final value 59.663345
## converged
## # weights: 109
## initial value 105.580125
## iter 10 value 63.714006
## iter 20 value 35.025028
## iter 30 value 21.801435
## iter
       40 value 17.655637
## iter 50 value 16.452754
## iter 60 value 16.089910
## iter 70 value 15.963306
## iter 80 value 15.961979
## final value 15.961975
## converged
```

```
## # weights: 179
## initial value 114.650738
## iter 10 value 41.052686
## iter 20 value 21.329416
## iter 30 value 14.684476
## iter 40 value 13.090797
## iter 50 value 12.767833
## iter 60 value 12.689669
## iter 70 value 12.600563
## iter 80 value 12.503758
## iter 90 value 12.409280
## iter 100 value 12.210580
## final value 12.210580
## stopped after 100 iterations
## # weights: 39
## initial value 115.811519
## iter 10 value 79.337877
## iter 20 value 58.080930
## iter 30 value 57.255951
## iter 40 value 57.123048
## iter 50 value 56.103605
## iter 60 value 55.394270
## iter 70 value 53.976233
## iter 80 value 53.766231
## iter 90 value 53.249351
## iter 100 value 52.797949
## final value 52.797949
## stopped after 100 iterations
## # weights: 109
## initial value 117.923969
## iter 10 value 24.646291
## iter 20 value 14.026214
## iter 30 value 9.177271
## iter 40 value 5.924326
## iter 50 value 5.483767
## iter 60 value 5.135246
## iter 70 value 4.958604
## iter 80 value 4.928025
## iter 90 value 4.906942
## iter 100 value 4.661015
## final value 4.661015
## stopped after 100 iterations
## # weights: 179
## initial value 116.128278
## iter 10 value 42.986851
## iter 20 value 14.446667
## iter 30 value 11.459803
## iter
       40 value 8.716805
## iter 50 value 8.694895
## iter 60 value 8.657438
## iter 70 value 8.621765
## iter 80 value 8.574368
## iter 90 value 8.551392
## iter 100 value 8.529180
```

```
## final value 8.529180
## stopped after 100 iterations
## # weights: 39
## initial value 107.340911
## iter 10 value 72.660866
## iter 20 value 66.396492
## iter 30 value 65.011520
## iter 40 value 64.354710
## iter 50 value 64.190403
## iter 60 value 64.042831
## iter 70 value 62.320274
## iter 80 value 62.147468
## iter 90 value 62.050015
## iter 100 value 61.795247
## final value 61.795247
## stopped after 100 iterations
## # weights: 109
## initial value 103.874866
## iter 10 value 23.723727
## iter 20 value 18.175229
## iter 30 value 16.711179
## iter 40 value 16.525947
## iter 50 value 15.656980
## iter 60 value 15.653166
## iter 70 value 15.652339
## iter 80 value 11.944017
## iter 90 value 11.854378
## iter 100 value 11.849322
## final value 11.849322
## stopped after 100 iterations
## # weights: 179
## initial value 113.038569
## iter 10 value 32.378972
## iter 20 value 8.421653
## iter 30 value 4.424333
## iter 40 value 3.355625
## iter 50 value 3.324319
## iter 60 value 3.139607
## iter 70 value 3.034021
## iter 80 value 3.014223
## final value 3.014210
## converged
## # weights: 39
## initial value 99.041836
## iter 10 value 67.109224
## iter 20 value 60.501539
## iter 30 value 59.480042
## iter 40 value 59.323734
## iter 50 value 59.118990
## iter 60 value 59.041670
## iter 70 value 59.004644
## iter 80 value 59.000747
## iter 90 value 59.000643
## iter 90 value 59.000642
```

```
## iter 90 value 59.000642
## final value 59.000642
## converged
## # weights: 109
## initial value 101.656206
## iter 10 value 54.337202
## iter 20 value 35.092868
## iter 30 value 20.243122
## iter 40 value 17.059275
## iter 50 value 16.739259
## iter 60 value 16.728669
## iter 70 value 16.728506
## final value 16.728505
## converged
## # weights: 179
## initial value 123.961804
## iter 10 value 43.347079
## iter 20 value 23.015554
## iter 30 value 15.211703
## iter 40 value 13.614319
## iter 50 value 13.066238
## iter 60 value 12.610179
## iter 70 value 12.497111
## iter 80 value 12.479448
## iter 90 value 12.479088
## final value 12.479084
## converged
## # weights: 39
## initial value 108.017094
## iter 10 value 65.099989
## iter 20 value 59.685023
## iter 30 value 57.108309
## iter 40 value 56.280465
## iter 50 value 55.785627
## iter 60 value 55.453078
## iter 70 value 55.036672
## iter 80 value 53.334857
## iter 90 value 52.551259
## iter 100 value 52.147789
## final value 52.147789
## stopped after 100 iterations
## # weights: 109
## initial value 116.672782
## iter 10 value 41.651438
## iter 20 value 27.238467
## iter 30 value 23.052093
## iter 40 value 19.017861
## iter 50 value 17.616272
## iter 60 value 12.500732
## iter 70 value 11.516223
## iter 80 value 10.583411
## iter 90 value 10.446586
## iter 100 value 10.261042
## final value 10.261042
```

```
## stopped after 100 iterations
## # weights: 179
## initial value 94.282572
## iter 10 value 22.961099
## iter 20 value 7.777698
## iter 30 value 0.681871
## iter 40 value 0.224789
## iter 50 value 0.192345
## iter 60 value 0.176125
## iter 70 value 0.165881
## iter 80 value 0.152975
## iter 90 value 0.143282
## iter 100 value 0.134875
## final value 0.134875
## stopped after 100 iterations
## # weights: 39
## initial value 110.021643
## iter 10 value 68.212301
## iter 20 value 59.550413
## iter 30 value 59.481303
## iter 40 value 59.392364
## iter 50 value 59.143030
## iter 60 value 57.941421
## iter 70 value 57.452159
## iter 80 value 57.352490
## iter 90 value 57.315204
## iter 100 value 50.270674
## final value 50.270674
## stopped after 100 iterations
## # weights: 109
## initial value 102.449494
## iter 10 value 44.123670
## iter 20 value 39.494611
## iter 30 value 36.662272
## iter 40 value 36.340838
## iter 50 value 36.287414
## iter 60 value 29.409074
## iter 70 value 26.102623
## iter 80 value 23.830512
## iter 90 value 23.335620
## iter 100 value 22.892945
## final value 22.892945
## stopped after 100 iterations
## # weights: 179
## initial value 109.368167
## iter 10 value 32.193699
## iter 20 value 7.439406
## iter 30 value 6.206705
## iter 40 value 6.078528
## iter 50 value 5.561726
## iter 60 value 5.463710
## iter 70 value 5.024072
## iter 80 value 5.023187
## iter 90 value 5.022420
```

```
## iter 100 value 5.022298
## final value 5.022298
## stopped after 100 iterations
## # weights: 39
## initial value 109.439745
## iter 10 value 73.986853
## iter 20 value 63.028384
## iter 30 value 62.211043
## iter 40 value 60.641053
## iter 50 value 58.404440
## iter 60 value 58.245921
## final value 58.245131
## converged
## # weights: 109
## initial value 107.880325
## iter 10 value 53.370218
## iter 20 value 26.541850
## iter 30 value 18.213279
## iter 40 value 16.622744
## iter 50 value 15.745120
## iter 60 value 15.589815
## iter 70 value 15.571255
## iter 80 value 15.570163
## final value 15.570148
## converged
## # weights: 179
## initial value 111.497857
## iter 10 value 43.211401
## iter 20 value 19.413065
## iter 30 value 13.084488
## iter 40 value 12.228906
## iter 50 value 11.910745
## iter 60 value 11.877340
## iter 70 value 11.811154
## iter 80 value 11.695662
## iter 90 value 11.690528
## iter 100 value 11.690448
## final value 11.690448
## stopped after 100 iterations
## # weights: 39
## initial value 103.125773
## iter 10 value 71.451501
## iter 20 value 61.829624
## iter 30 value 56.122091
## iter 40 value 53.219752
## iter 50 value 52.470908
## iter 60 value 52.453926
## iter 70 value 52.418181
## iter 80 value 52.376890
## iter 90 value 52.374963
## iter 100 value 52.367397
## final value 52.367397
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 99.332307
## iter 10 value 37.815508
## iter 20 value 21.659869
## iter 30 value 14.800785
## iter 40 value 10.486945
## iter 50 value 9.263875
## iter 60 value 8.688339
## iter 70 value 8.304383
## iter 80 value 7.822464
## iter 90 value 7.776147
## iter 100 value 7.460304
## final value 7.460304
## stopped after 100 iterations
## # weights: 179
## initial value 121.814606
## iter 10 value 38.978621
## iter 20 value 13.983112
## iter 30 value 6.622315
## iter 40 value 4.770961
## iter 50 value 3.860275
## iter 60 value 3.551931
## iter 70 value 3.525396
## iter 80 value 3.501125
## iter 90 value 3.465392
## iter 100 value 3.421909
## final value 3.421909
## stopped after 100 iterations
## # weights: 39
## initial value 102.174818
## iter 10 value 67.613526
## iter 20 value 59.602468
## iter 30 value 56.104884
## iter 40 value 52.593574
## iter 50 value 50.422135
## iter 60 value 48.705385
## iter 70 value 46.473854
## iter 80 value 45.935970
## iter 90 value 44.882264
## iter 100 value 42.778872
## final value 42.778872
## stopped after 100 iterations
## # weights: 109
## initial value 123.849203
## iter 10 value 44.407363
## iter 20 value 27.265906
## iter 30 value 25.611426
## iter 40 value 24.241634
## iter 50 value 23.730479
## iter 60 value 23.511143
## iter 70 value 23.332134
## iter 80 value 23.039610
## iter 90 value 23.029608
## iter 100 value 23.028797
## final value 23.028797
```

```
## stopped after 100 iterations
## # weights: 179
## initial value 102.813888
## iter 10 value 10.676891
## iter 20 value 3.874461
## iter 30 value 0.197353
## iter 40 value 0.022392
## iter 50 value 0.006858
## iter 60 value 0.003255
## iter 70 value 0.002284
## iter 80 value 0.000879
## iter 90 value 0.000760
## iter 100 value 0.000644
## final value 0.000644
## stopped after 100 iterations
## # weights: 39
## initial value 114.315697
## iter 10 value 76.631101
## iter 20 value 63.977421
## iter 30 value 60.801777
## iter 40 value 58.862362
## iter 50 value 58.805680
## final value 58.805662
## converged
## # weights: 109
## initial value 106.364714
## iter 10 value 51.340241
## iter 20 value 31.023406
## iter 30 value 23.868594
## iter 40 value 21.237593
## iter 50 value 17.254159
## iter 60 value 15.957460
## iter 70 value 15.783566
## iter 80 value 15.782541
## final value 15.782540
## converged
## # weights: 179
## initial value 118.194570
## iter 10 value 49.684520
## iter 20 value 26.817583
## iter 30 value 18.452851
## iter 40 value 16.126790
## iter 50 value 15.198806
## iter 60 value 13.817211
## iter 70 value 12.730359
## iter 80 value 12.569357
## iter 90 value 12.555857
## iter 100 value 12.555210
## final value 12.555210
## stopped after 100 iterations
## # weights: 39
## initial value 104.093704
## iter 10 value 87.064566
## iter 20 value 73.034495
```

```
## iter 30 value 64.654843
## iter 40 value 62.123316
## iter 50 value 60.579700
## iter 60 value 56.961239
## iter 70 value 56.298580
## iter 80 value 56.235444
## iter 90 value 56.112641
## iter 100 value 55.676192
## final value 55.676192
## stopped after 100 iterations
## # weights: 109
## initial value 105.599813
## iter 10 value 33.900488
## iter 20 value 20.542418
## iter 30 value 16.328696
## iter 40 value 12.349470
## iter 50 value 11.675740
## iter 60 value 11.176623
## iter 70 value 10.955957
## iter 80 value 10.919434
## iter 90 value 10.482479
## iter 100 value 10.197780
## final value 10.197780
## stopped after 100 iterations
## # weights: 179
## initial value 121.186413
## iter 10 value 30.571532
## iter 20 value 10.703225
## iter 30 value 6.003909
## iter 40 value 4.975763
## iter 50 value 4.935455
## iter 60 value 4.291221
## iter 70 value 2.088793
## iter 80 value 2.057112
## iter 90 value 0.267156
## iter 100 value 0.168442
## final value 0.168442
## stopped after 100 iterations
## # weights: 39
## initial value 110.877185
## iter 10 value 76.407135
## iter 20 value 72.348556
## iter 30 value 69.966259
## iter 40 value 69.547092
## iter 50 value 69.498840
## iter 60 value 69.495193
## iter 70 value 69.493092
## iter 80 value 69.492085
## iter 90 value 69.491702
## iter 100 value 69.491507
## final value 69.491507
## stopped after 100 iterations
## # weights: 109
## initial value 105.196367
```

```
## iter 10 value 37.023872
## iter 20 value 17.167431
## iter 30 value 9.115900
## iter 40 value 8.242091
## iter 50 value 7.888245
## iter 60 value 7.738698
## iter 70 value 7.644838
## iter 80 value 6.979559
## iter 90 value 6.708106
## iter 100 value 6.190028
## final value 6.190028
## stopped after 100 iterations
## # weights: 179
## initial value 106.516563
## iter 10 value 20.375429
## iter 20 value 5.266175
## iter 30 value 3.049322
## iter 40 value 0.101922
## iter 50 value 0.018552
## iter 60 value 0.006890
## iter 70 value 0.001468
## iter 80 value 0.000607
## iter 90 value 0.000303
## iter 100 value 0.000223
## final value 0.000223
## stopped after 100 iterations
## # weights: 39
## initial value 101.404151
## iter 10 value 77.673161
## iter 20 value 64.204767
## iter 30 value 60.219105
## iter 40 value 58.820615
## iter 50 value 56.904625
## iter 60 value 56.388151
## iter 70 value 56.328674
## final value 56.328205
## converged
## # weights: 109
## initial value 101.292210
## iter 10 value 50.777953
## iter 20 value 29.976999
## iter 30 value 22.164987
## iter 40 value 17.258583
## iter 50 value 17.016999
## iter 60 value 16.955515
## iter 70 value 16.954735
## iter 70 value 16.954735
## iter 70 value 16.954735
## final value 16.954735
## converged
## # weights: 179
## initial value 104.055174
## iter 10 value 47.348492
## iter 20 value 23.581423
```

```
## iter 30 value 15.088989
## iter 40 value 13.470877
## iter 50 value 12.780743
## iter 60 value 12.646527
## iter 70 value 12.619079
## iter 80 value 12.615867
## iter 90 value 12.615342
## final value 12.615286
## converged
## # weights: 39
## initial value 113.953935
## iter 10 value 74.348849
## iter 20 value 64.698372
## iter 30 value 60.667436
## iter 40 value 55.076760
## iter 50 value 54.340462
## iter 60 value 53.604334
## iter 70 value 52.059013
## iter 80 value 51.691377
## iter 90 value 51.024556
## iter 100 value 49.925475
## final value 49.925475
## stopped after 100 iterations
## # weights: 109
## initial value 105.875971
## iter 10 value 30.724684
## iter 20 value 20.858453
## iter 30 value 20.198774
## iter 40 value 20.147281
## iter 50 value 20.127714
## iter 60 value 19.977439
## iter 70 value 18.689845
## iter 80 value 18.636513
## iter 90 value 18.600274
## iter 100 value 14.984050
## final value 14.984050
## stopped after 100 iterations
## # weights: 179
## initial value 113.916567
## iter 10 value 22.879010
## iter 20 value 19.433213
## iter 30 value 16.111486
## iter 40 value 14.549640
## iter 50 value 14.292218
## iter 60 value 14.033920
## iter 70 value 13.727135
## iter 80 value 13.401281
## iter 90 value 13.208178
## iter 100 value 13.044966
## final value 13.044966
## stopped after 100 iterations
## # weights: 39
## initial value 101.336606
## iter 10 value 58.713996
```

```
## iter 20 value 53.202679
## iter 30 value 50.970440
## iter 40 value 46.339869
## iter 50 value 45.662357
## iter 60 value 45.654508
## iter 70 value 45.654307
## iter 80 value 45.654286
## iter 80 value 45.654286
## iter 80 value 45.654286
## final value 45.654286
## converged
## # weights: 109
## initial value 113.403331
## iter 10 value 34.088838
## iter 20 value 16.824174
## iter 30 value 9.719780
## iter 40 value 7.304369
## iter 50 value 4.207232
## iter 60 value 3.889979
## iter 70 value 3.739514
## iter 80 value 3.640510
## iter 90 value 3.387003
## iter 100 value 3.362435
## final value 3.362435
## stopped after 100 iterations
## # weights: 179
## initial value 102.553090
## iter 10 value 30.479127
## iter 20 value 20.209764
## iter 30 value 14.850936
## iter 40 value 12.200090
## iter 50 value 9.978704
## iter 60 value 9.364917
## iter 70 value 9.270627
## iter 80 value 9.179702
## iter 90 value 8.941796
## iter 100 value 8.651415
## final value 8.651415
## stopped after 100 iterations
## # weights: 39
## initial value 123.515370
## iter 10 value 83.153644
## iter 20 value 70.419076
## iter 30 value 64.038411
## iter 40 value 62.308386
## iter 50 value 59.468454
## iter 60 value 58.712067
## iter 70 value 58.678013
## final value 58.677908
## converged
## # weights: 109
## initial value 109.985406
## iter 10 value 51.749223
## iter 20 value 35.665648
```

```
## iter 30 value 25.664766
## iter 40 value 21.912126
## iter 50 value 19.258182
## iter 60 value 17.923731
## iter 70 value 17.399123
## iter 80 value 17.054677
## iter 90 value 17.008316
## iter 100 value 16.999392
## final value 16.999392
## stopped after 100 iterations
## # weights: 179
## initial value 109.986602
## iter 10 value 40.930350
## iter 20 value 23.371774
## iter 30 value 15.478120
## iter 40 value 12.919731
## iter 50 value 12.301632
## iter 60 value 12.187109
## iter 70 value 12.177108
## iter 80 value 12.174443
## iter 90 value 12.174280
## final value 12.174272
## converged
## # weights: 39
## initial value 106.896869
## iter 10 value 75.647908
## iter 20 value 62.928572
## iter 30 value 60.338873
## iter 40 value 56.850236
## iter 50 value 54.587161
## iter 60 value 52.077018
## iter 70 value 49.267671
## iter 80 value 46.899490
## iter 90 value 46.375950
## iter 100 value 46.048647
## final value 46.048647
## stopped after 100 iterations
## # weights: 109
## initial value 104.227153
## iter 10 value 49.408174
## iter 20 value 33.398251
## iter 30 value 26.351524
## iter 40 value 21.805631
## iter 50 value 20.833091
## iter 60 value 20.257008
        70 value 20.185132
## iter
## iter 80 value 20.135941
## iter 90 value 20.071287
## iter 100 value 20.016161
## final value 20.016161
## stopped after 100 iterations
## # weights: 179
## initial value 132.382320
## iter 10 value 67.797836
```

```
## iter 20 value 45.232275
## iter 30 value 22.550630
## iter 40 value 11.521243
## iter 50 value 9.066594
## iter 60 value 8.286383
## iter 70 value 7.339770
## iter 80 value 7.113598
## iter 90 value 7.090570
## iter 100 value 7.044105
## final value 7.044105
## stopped after 100 iterations
## # weights: 39
## initial value 106.255009
## iter 10 value 66.829748
## iter 20 value 61.989661
## iter 30 value 60.600386
## iter 40 value 59.364435
## iter 50 value 58.857610
## iter 60 value 58.583840
## iter 70 value 56.685758
## iter 80 value 55.652771
## iter 90 value 54.303317
## iter 100 value 54.088777
## final value 54.088777
## stopped after 100 iterations
## # weights: 109
## initial value 95.237205
## iter 10 value 46.263256
## iter 20 value 14.889634
## iter 30 value 9.748428
## iter 40 value 7.481966
## iter 50 value 6.447576
## iter 60 value 6.247395
## iter 70 value 6.246110
## iter 80 value 6.244189
## iter 90 value 6.243411
## iter 100 value 6.243190
## final value 6.243190
## stopped after 100 iterations
## # weights: 179
## initial value 103.675430
## iter 10 value 29.037170
## iter 20 value 10.186919
## iter 30 value 5.119725
## iter 40 value 2.403088
## iter 50 value 0.084057
## iter 60 value 0.012795
## iter
       70 value 0.001608
## iter 80 value 0.000367
## iter 90 value 0.000108
## iter 90 value 0.000093
## iter 90 value 0.000093
## final value 0.000093
## converged
```

```
## # weights: 39
## initial value 101.282811
## iter 10 value 68.911659
## iter 20 value 65.579028
## iter 30 value 63.183165
## iter 40 value 60.489143
## iter 50 value 58.956900
## iter 60 value 58.877260
## iter 70 value 58.853650
## iter 80 value 58.814921
## iter 90 value 58.812273
## final value 58.812269
## converged
## # weights: 109
## initial value 115.216886
## iter 10 value 50.025580
## iter 20 value 28.517549
## iter 30 value 20.338125
## iter 40 value 16.609221
## iter 50 value 16.169486
## iter 60 value 16.134992
## iter 70 value 16.134684
## final value 16.134684
## converged
## # weights: 179
## initial value 116.085321
## iter 10 value 30.954048
## iter 20 value 18.705041
## iter 30 value 15.362583
## iter 40 value 13.987375
## iter 50 value 12.235154
## iter 60 value 11.987891
## iter 70 value 11.935045
## iter 80 value 11.889550
## iter 90 value 11.814101
## iter 100 value 11.698297
## final value 11.698297
## stopped after 100 iterations
## # weights: 39
## initial value 110.172818
## iter 10 value 66.795218
## iter 20 value 56.627134
## iter 30 value 55.113840
## iter 40 value 53.916655
## iter 50 value 53.513457
## iter 60 value 53.429708
## iter 70 value 53.424925
## iter 80 value 53.409361
## iter 90 value 53.323718
## iter 100 value 52.026737
## final value 52.026737
## stopped after 100 iterations
## # weights: 109
## initial value 105.943214
```

```
## iter 10 value 61.305445
## iter 20 value 46.566526
## iter 30 value 45.255248
## iter 40 value 44.839251
## iter 50 value 44.365837
## iter 60 value 43.802265
## iter 70 value 42.893714
## iter 80 value 42.582211
## iter 90 value 41.538761
## iter 100 value 41.511259
## final value 41.511259
## stopped after 100 iterations
## # weights: 179
## initial value 114.208948
## iter 10 value 13.133927
## iter 20 value 0.377683
## iter 30 value 0.206190
## iter 40 value 0.182578
## iter 50 value 0.161110
## iter 60 value 0.137105
## iter 70 value 0.126345
## iter 80 value 0.118637
## iter 90 value 0.111652
## iter 100 value 0.104741
## final value 0.104741
## stopped after 100 iterations
## # weights: 39
## initial value 97.141722
## iter 10 value 63.657723
## iter 20 value 56.257900
## iter 30 value 50.666105
## iter 40 value 43.475527
## iter 50 value 39.802194
## iter 60 value 38.635474
## iter 70 value 38.446019
## iter 80 value 38.401404
## iter 90 value 38.375354
## iter 100 value 38.361117
## final value 38.361117
## stopped after 100 iterations
## # weights: 109
## initial value 109.996305
## iter 10 value 22.052461
## iter 20 value 17.123695
## iter 30 value 16.218858
## iter 40 value 15.686875
## iter 50 value 14.917486
## iter
       60 value 14.028244
## iter 70 value 12.434591
## iter 80 value 12.181066
## iter 90 value 10.251471
## iter 100 value 10.249067
## final value 10.249067
## stopped after 100 iterations
```

```
## # weights: 179
## initial value 130.458739
## iter 10 value 47.501540
## iter 20 value 23.194329
## iter 30 value 17.354249
## iter 40 value 16.618484
## iter 50 value 14.646119
## iter 60 value 14.325470
## iter 70 value 14.278478
## iter 80 value 14.225292
## iter 90 value 14.137936
## iter 100 value 14.071084
## final value 14.071084
## stopped after 100 iterations
## # weights: 39
## initial value 108.177681
## iter 10 value 72.116776
## iter 20 value 64.391265
## iter 30 value 60.224527
## iter 40 value 57.103019
## iter 50 value 55.732328
## iter 60 value 55.658225
## final value 55.657593
## converged
## # weights: 109
## initial value 106.246967
## iter 10 value 40.889581
## iter 20 value 27.603289
## iter 30 value 17.852300
## iter 40 value 16.264181
## iter 50 value 16.007933
## iter 60 value 15.766623
## iter 70 value 15.270261
## iter 80 value 14.501568
## iter 90 value 14.486939
## final value 14.486893
## converged
## # weights: 179
## initial value 100.280449
## iter 10 value 44.011581
## iter 20 value 22.418312
## iter 30 value 15.536459
## iter 40 value 13.070783
## iter 50 value 12.258427
## iter 60 value 11.768666
## iter 70 value 11.601077
## iter 80 value 11.450492
## iter 90 value 11.432543
## iter 100 value 11.409719
## final value 11.409719
## stopped after 100 iterations
## # weights: 39
## initial value 100.902533
## iter 10 value 69.495457
```

```
## iter 20 value 56.489412
## iter 30 value 52.155389
## iter 40 value 49.536678
## iter 50 value 47.961094
## iter 60 value 46.255032
## iter 70 value 45.697391
## iter 80 value 45.109502
## iter 90 value 44.840550
## iter 100 value 43.631604
## final value 43.631604
## stopped after 100 iterations
## # weights: 109
## initial value 105.241944
## iter 10 value 18.004443
## iter 20 value 13.521945
## iter 30 value 11.639554
## iter 40 value 11.310311
## iter 50 value 11.252484
## iter 60 value 11.203879
## iter 70 value 11.133833
## iter 80 value 11.101025
## iter 90 value 11.046984
## iter 100 value 11.000440
## final value 11.000440
## stopped after 100 iterations
## # weights: 179
## initial value 97.749766
## iter 10 value 25.020107
## iter 20 value 10.201067
## iter 30 value 8.053348
## iter 40 value 7.415986
## iter 50 value 6.991348
## iter 60 value 2.739974
## iter 70 value 2.432719
## iter 80 value 0.218547
## iter 90 value 0.202425
## iter 100 value 0.180217
## final value 0.180217
## stopped after 100 iterations
## # weights: 39
## initial value 102.074074
## iter 10 value 76.071231
## iter 20 value 71.110618
## iter 30 value 69.010607
## iter 40 value 68.620046
## iter 50 value 68.575700
## iter 60 value 68.571629
## iter 70 value 68.570057
## iter 80 value 68.569106
## iter 90 value 68.568971
## final value 68.568960
## converged
## # weights: 109
## initial value 107.852459
```

```
## iter 10 value 59.224926
## iter 20 value 39.232537
## iter 30 value 36.123722
## iter 40 value 33.782371
## iter 50 value 33.393436
## iter 60 value 32.708590
## iter 70 value 32.305549
## iter 80 value 31.866234
## iter 90 value 31.567093
## iter 100 value 31.051610
## final value 31.051610
## stopped after 100 iterations
## # weights: 179
## initial value 122.246607
## iter 10 value 26.359998
## iter 20 value 16.447792
## iter 30 value 15.652143
## iter 40 value 15.491780
## iter 50 value 14.761786
## iter 60 value 14.753042
## iter 70 value 14.729595
## iter 80 value 12.965682
## iter 90 value 12.848481
## iter 100 value 12.845797
## final value 12.845797
## stopped after 100 iterations
## # weights: 39
## initial value 117.616004
## iter 10 value 70.091623
## iter 20 value 61.919012
## iter 30 value 60.350121
## iter 40 value 59.883131
## iter 50 value 59.557645
## iter 60 value 58.750542
## final value 58.738443
## converged
## # weights: 109
## initial value 110.271079
## iter 10 value 57.231152
## iter 20 value 29.598527
## iter 30 value 19.980459
## iter 40 value 17.459025
## iter 50 value 17.018922
## iter 60 value 16.885743
## iter 70 value 16.687703
## iter 80 value 16.614984
## iter 90 value 16.613148
## iter 90 value 16.613148
## iter 90 value 16.613148
## final value 16.613148
## converged
## # weights: 179
## initial value 108.148610
## iter 10 value 60.864418
```

```
## iter 20 value 26.107739
## iter 30 value 14.821464
## iter 40 value 12.936570
## iter 50 value 12.488586
## iter 60 value 12.399275
## iter 70 value 12.387366
## iter 80 value 12.384074
## iter 90 value 12.384002
## final value 12.384000
## converged
## # weights: 39
## initial value 99.672219
## iter 10 value 66.668741
## iter 20 value 56.541865
## iter 30 value 49.647541
## iter 40 value 46.983637
## iter 50 value 46.436520
## iter 60 value 45.291575
## iter 70 value 45.036242
## iter 80 value 45.023579
## iter 90 value 44.410058
## iter 100 value 42.803474
## final value 42.803474
## stopped after 100 iterations
## # weights: 109
## initial value 106.520235
## iter 10 value 39.520266
## iter 20 value 23.039909
## iter 30 value 15.670569
## iter 40 value 12.580631
## iter 50 value 9.704793
## iter 60 value 8.520717
## iter 70 value 8.176335
## iter 80 value 8.106534
## iter 90 value 7.807544
## iter 100 value 7.434241
## final value 7.434241
## stopped after 100 iterations
## # weights: 179
## initial value 94.639251
## iter 10 value 26.432558
## iter 20 value 11.694838
## iter 30 value 3.620085
## iter 40 value 2.811757
## iter 50 value 0.331177
## iter 60 value 0.255240
## iter 70 value 0.232505
## iter 80 value 0.219530
## iter 90 value 0.207084
## iter 100 value 0.187687
## final value 0.187687
## stopped after 100 iterations
## # weights: 39
## initial value 101.955099
```

```
## iter 10 value 58.547455
## iter 20 value 56.096916
## iter 30 value 55.576029
## iter 40 value 54.694568
## iter 50 value 54.276457
## iter 60 value 54.085698
## iter 70 value 54.055668
## iter 80 value 54.048491
## iter 90 value 54.047021
## final value 54.047018
## converged
## # weights: 109
## initial value 113.799929
## iter 10 value 50.605766
## iter 20 value 39.075051
## iter 30 value 15.877394
## iter 40 value 6.633256
## iter 50 value 5.607312
## iter 60 value 5.581716
## iter 70 value 5.581393
## iter 80 value 5.581202
## final value 5.581199
## converged
## # weights: 179
## initial value 102.433560
## iter 10 value 27.630212
## iter 20 value 8.169679
## iter 30 value 4.015954
## iter 40 value 3.823980
## iter 50 value 3.819553
## iter 60 value 3.819135
## iter 70 value 3.819110
## iter 80 value 3.819096
## final value 3.819095
## converged
## # weights: 39
## initial value 104.894651
## iter 10 value 67.222319
## iter 20 value 60.939577
## iter 30 value 58.827110
## iter 40 value 58.535081
## iter 50 value 58.276064
## iter 60 value 58.231354
## iter 70 value 57.967052
## final value 57.964173
## converged
## # weights: 109
## initial value 103.171180
## iter 10 value 53.155911
## iter 20 value 31.112623
## iter 30 value 19.107940
## iter 40 value 16.822459
## iter 50 value 16.317978
## iter 60 value 16.198758
```

```
## iter 70 value 15.755214
## iter 80 value 15.643074
## final value 15.642186
## converged
## # weights: 179
## initial value 126.716014
## iter 10 value 37.553336
## iter 20 value 18.808806
## iter 30 value 14.720767
## iter 40 value 12.824745
## iter 50 value 12.308859
## iter 60 value 11.965359
## iter 70 value 11.946654
## iter 80 value 11.945785
## final value 11.945778
## converged
## # weights: 39
## initial value 106.506137
## iter 10 value 63.483543
## iter 20 value 58.251297
## iter 30 value 56.306057
## iter 40 value 56.193543
## iter 50 value 55.944139
## iter 60 value 55.099116
## iter 70 value 54.762976
## iter 80 value 54.051169
## iter 90 value 53.387555
## iter 100 value 53.237653
## final value 53.237653
## stopped after 100 iterations
## # weights: 109
## initial value 100.743883
## iter 10 value 43.836289
## iter 20 value 28.798707
## iter 30 value 11.310145
## iter 40 value 6.674740
## iter 50 value 6.303535
## iter 60 value 6.007749
## iter 70 value 4.920568
## iter 80 value 4.839844
## iter 90 value 4.343286
## iter 100 value 4.327982
## final value 4.327982
## stopped after 100 iterations
## # weights: 179
## initial value 108.255633
## iter 10 value 37.536592
## iter 20 value 14.056476
## iter 30 value 10.499864
## iter 40 value 9.325585
## iter 50 value 9.181399
## iter 60 value 9.132712
## iter 70 value 9.108987
## iter 80 value 9.085355
```

```
## iter 90 value 9.001335
## iter 100 value 8.989747
## final value 8.989747
## stopped after 100 iterations
## # weights: 39
## initial value 113.983098
## iter 10 value 87.441127
## iter 20 value 79.352612
## iter 30 value 78.066712
## iter 40 value 76.671679
## iter 50 value 74.432980
## iter 60 value 73.952666
## iter 70 value 73.838108
## iter 80 value 73.819343
## iter 90 value 73.813846
## iter 100 value 73.812058
## final value 73.812058
## stopped after 100 iterations
## # weights: 109
## initial value 107.889333
## iter 10 value 40.854062
## iter 20 value 25.007905
## iter 30 value 19.239009
## iter 40 value 18.638793
## iter 50 value 18.518591
## iter 60 value 18.311120
## iter 70 value 17.976175
## iter 80 value 17.963642
## iter 90 value 17.959310
## iter 100 value 17.948484
## final value 17.948484
## stopped after 100 iterations
## # weights: 179
## initial value 113.977287
## iter 10 value 48.286552
## iter 20 value 20.015456
## iter 30 value 12.575581
## iter 40 value 7.679967
## iter 50 value 5.951121
## iter 60 value 4.206847
## iter 70 value 4.181080
## iter 80 value 4.172679
## iter 90 value 4.165677
## iter 100 value 4.164398
## final value 4.164398
## stopped after 100 iterations
## # weights: 39
## initial value 104.295824
## iter 10 value 75.210877
## iter 20 value 63.048352
## iter 30 value 58.488840
## iter 40 value 57.414507
## iter 50 value 57.394265
## final value 57.394232
```

```
## converged
## # weights: 109
## initial value 115.381280
## iter 10 value 48.512886
## iter 20 value 40.417664
## iter 30 value 33.541293
## iter 40 value 24.289052
## iter 50 value 18.641956
## iter 60 value 17.361724
## iter 70 value 17.056913
## iter 80 value 16.948316
## iter 90 value 16.947875
## final value 16.947873
## converged
## # weights: 179
## initial value 110.424543
## iter 10 value 31.157217
## iter 20 value 17.932690
## iter 30 value 14.527502
## iter 40 value 13.163494
## iter 50 value 12.950152
## iter 60 value 12.634455
## iter 70 value 12.527285
## iter 80 value 12.366793
## iter 90 value 12.354188
## iter 100 value 12.351244
## final value 12.351244
## stopped after 100 iterations
## # weights: 39
## initial value 104.133840
## iter 10 value 74.176563
## iter 20 value 62.253660
## iter 30 value 60.173528
## iter 40 value 57.933405
## iter 50 value 56.549344
## iter 60 value 52.659860
## iter 70 value 51.759683
## iter 80 value 51.599303
## iter 90 value 49.616761
## iter 100 value 49.371199
## final value 49.371199
## stopped after 100 iterations
## # weights: 109
## initial value 94.492022
## iter 10 value 37.476186
## iter 20 value 28.270609
## iter 30 value 26.324287
## iter
       40 value 24.945171
## iter 50 value 23.881297
## iter 60 value 23.683917
## iter 70 value 23.582756
## iter 80 value 23.558394
## iter 90 value 23.492785
## iter 100 value 23.404812
```

```
## final value 23.404812
## stopped after 100 iterations
## # weights: 179
## initial value 103.884269
## iter 10 value 26.365540
## iter 20 value 7.869652
## iter 30 value 3.692560
## iter 40 value 3.132682
## iter 50 value 2.959503
## iter 60 value 2.929386
## iter 70 value 2.909333
## iter 80 value 2.426242
## iter 90 value 2.099608
## iter 100 value 2.074604
## final value 2.074604
## stopped after 100 iterations
## # weights: 39
## initial value 106.952758
## iter 10 value 72.637910
## iter 20 value 62.074466
## iter 30 value 58.996568
## iter 40 value 58.164663
## iter 50 value 58.095806
## iter 60 value 56.929747
## iter 70 value 55.880715
## iter 80 value 55.867418
## iter 90 value 55.839945
## iter 100 value 55.825475
## final value 55.825475
## stopped after 100 iterations
## # weights: 109
## initial value 105.912788
## iter 10 value 48.835434
## iter 20 value 25.377404
## iter 30 value 17.550656
## iter 40 value 12.536803
## iter 50 value 10.418910
## iter 60 value 9.738811
## iter 70 value 9.701508
## iter 80 value 9.591408
## iter 90 value 9.534155
## iter 100 value 9.446017
## final value 9.446017
## stopped after 100 iterations
## # weights: 179
## initial value 106.309196
## iter 10 value 26.936341
## iter 20 value 7.188706
## iter 30 value 2.016982
## iter 40 value 1.726872
## iter 50 value 1.394768
## iter 60 value 1.391318
## iter 70 value 1.388821
## iter 80 value 1.387579
```

```
## iter 90 value 1.387342
## iter 100 value 1.386729
## final value 1.386729
## stopped after 100 iterations
## # weights: 39
## initial value 106.758200
## iter 10 value 68.060533
## iter 20 value 58.827192
## iter 30 value 57.217470
## iter 40 value 56.801505
## iter 50 value 56.488998
## iter 60 value 56.361462
## iter 70 value 56.355441
## final value 56.355432
## converged
## # weights: 109
## initial value 104.347575
## iter 10 value 35.032481
## iter 20 value 19.578919
## iter 30 value 17.474573
## iter 40 value 17.185165
## iter 50 value 16.163480
## iter 60 value 15.653149
## iter 70 value 15.468152
## iter 80 value 14.948157
## iter 90 value 14.942889
## final value 14.942880
## converged
## # weights: 179
## initial value 116.380740
## iter 10 value 37.487523
## iter 20 value 17.297219
## iter 30 value 13.261150
## iter 40 value 11.801239
## iter 50 value 11.587774
## iter 60 value 11.573982
## iter 70 value 11.573726
## final value 11.573721
## converged
## # weights: 39
## initial value 110.591086
## iter 10 value 79.701509
## iter 20 value 66.051565
## iter 30 value 63.311775
## iter 40 value 60.188671
## iter 50 value 57.665481
## iter 60 value 56.028626
## iter 70 value 53.021714
## iter 80 value 52.584545
## iter 90 value 50.528819
## iter 100 value 49.187377
## final value 49.187377
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 111.225988
## iter 10 value 55.032260
## iter 20 value 31.243901
## iter 30 value 21.242487
## iter 40 value 20.853761
## iter 50 value 20.505548
## iter 60 value 20.263619
## iter 70 value 19.955041
## iter 80 value 19.906153
## iter 90 value 19.825165
## iter 100 value 19.805037
## final value 19.805037
## stopped after 100 iterations
## # weights: 179
## initial value 108.085848
## iter 10 value 43.059980
## iter 20 value 4.622010
## iter 30 value 0.277884
## iter 40 value 0.233860
## iter 50 value 0.202374
## iter 60 value 0.177374
## iter 70 value 0.154542
## iter 80 value 0.146888
## iter 90 value 0.133103
## iter 100 value 0.128138
## final value 0.128138
## stopped after 100 iterations
## # weights: 39
## initial value 111.614371
## iter 10 value 67.853567
## iter 20 value 56.814323
## iter 30 value 53.091755
## iter 40 value 53.069044
## iter 50 value 53.067391
## iter 60 value 53.067199
## iter 70 value 53.067139
## final value 53.067093
## converged
## # weights: 109
## initial value 114.080231
## iter 10 value 58.374017
## iter 20 value 42.452250
## iter 30 value 29.675842
## iter 40 value 25.644864
## iter 50 value 25.331905
## iter 60 value 21.333729
## iter 70 value 20.307710
## iter 80 value 20.270711
## iter 90 value 20.264074
## iter 100 value 20.263155
## final value 20.263155
## stopped after 100 iterations
## # weights: 179
## initial value 105.586240
```

```
## iter 10 value 18.053246
## iter 20 value 5.637791
## iter 30 value 0.318545
## iter 40 value 0.009110
## iter 50 value 0.000690
## iter 60 value 0.000199
## iter 70 value 0.000102
## final value 0.000100
## converged
## # weights: 39
## initial value 103.785528
## iter 10 value 67.873462
## iter 20 value 63.564800
## iter 30 value 61.075913
## iter 40 value 60.650053
## iter 50 value 60.141771
## iter 60 value 58.677670
## iter 70 value 58.158505
## iter 80 value 57.980523
## iter 90 value 57.961032
## iter 90 value 57.961032
## iter 90 value 57.961032
## final value 57.961032
## converged
## # weights: 109
## initial value 116.856886
## iter 10 value 47.658898
## iter 20 value 31.980101
## iter 30 value 20.532882
## iter 40 value 18.388903
## iter 50 value 17.760540
## iter 60 value 17.506323
## iter 70 value 16.928798
## iter 80 value 16.607733
## iter 90 value 16.529480
## iter 100 value 16.513202
## final value 16.513202
## stopped after 100 iterations
## # weights: 179
## initial value 125.097607
## iter 10 value 54.197389
## iter 20 value 30.854924
## iter 30 value 21.855689
## iter 40 value 16.674901
## iter 50 value 14.312230
## iter 60 value 12.945907
## iter 70 value 12.813269
## iter 80 value 12.806684
## final value 12.806602
## converged
## # weights: 39
## initial value 109.454281
## iter 10 value 68.489554
## iter 20 value 58.051799
```

```
## iter 30 value 53.950597
## iter 40 value 50.492012
## iter 50 value 50.267121
## iter 60 value 50.249982
## iter 70 value 50.183703
## iter 80 value 50.142698
## iter 90 value 50.139174
## iter 100 value 50.124259
## final value 50.124259
## stopped after 100 iterations
## # weights: 109
## initial value 104.158381
## iter 10 value 29.158508
## iter 20 value 15.290944
## iter 30 value 13.104837
## iter 40 value 12.508605
## iter 50 value 12.405232
## iter 60 value 12.358646
## iter 70 value 12.026993
## iter 80 value 11.813131
## iter 90 value 11.798816
## iter 100 value 11.765708
## final value 11.765708
## stopped after 100 iterations
## # weights: 179
## initial value 126.750437
## iter 10 value 62.013794
## iter 20 value 56.182839
## iter 30 value 53.208551
## iter 40 value 52.784011
## iter 50 value 52.647743
## iter 60 value 52.524650
## iter 70 value 52.113067
## iter 80 value 51.703608
## iter 90 value 50.046333
## iter 100 value 48.482687
## final value 48.482687
## stopped after 100 iterations
## # weights: 39
## initial value 102.783425
## iter 10 value 63.566598
## iter 20 value 59.943813
## iter 30 value 59.019441
## iter 40 value 58.120556
## iter 50 value 52.672595
## iter 60 value 51.869770
## iter 70 value 50.866956
## iter 80 value 50.633918
## iter 90 value 49.237394
## iter 100 value 49.172407
## final value 49.172407
## stopped after 100 iterations
## # weights: 109
## initial value 101.225242
```

```
## iter 10 value 56.818837
## iter 20 value 30.455290
## iter 30 value 17.238070
## iter 40 value 8.748559
## iter 50 value 8.702947
## iter 60 value 8.599904
## iter 70 value 8.040919
## iter 80 value 7.187679
## iter 90 value 6.759235
## iter 100 value 6.757061
## final value 6.757061
## stopped after 100 iterations
## # weights: 179
## initial value 138.253773
## iter 10 value 47.350613
## iter 20 value 32.195540
## iter 30 value 18.590606
## iter
       40 value 17.754454
## iter 50 value 17.452344
## iter 60 value 17.294295
## iter 70 value 11.162064
## iter 80 value 7.278059
## iter 90 value 6.718116
## iter 100 value 6.681113
## final value 6.681113
## stopped after 100 iterations
## # weights: 39
## initial value 105.595506
## iter 10 value 70.007793
## iter 20 value 63.735123
## iter 30 value 60.656865
## iter 40 value 58.436173
## iter 50 value 58.376796
## final value 58.376740
## converged
## # weights: 109
## initial value 104.408190
## iter 10 value 54.812636
## iter 20 value 33.437008
## iter 30 value 26.143288
## iter 40 value 19.021726
## iter 50 value 16.400660
## iter 60 value 15.749509
## iter 70 value 15.735730
## iter 80 value 15.734676
## final value 15.734656
## converged
## # weights: 179
## initial value 144.306136
## iter 10 value 67.928940
## iter 20 value 34.866184
## iter 30 value 21.767936
## iter 40 value 15.933002
## iter 50 value 13.188490
```

```
## iter 60 value 11.904032
## iter 70 value 11.687760
## iter 80 value 11.676570
## iter 90 value 11.676338
## final value 11.676334
## converged
## # weights: 39
## initial value 100.249235
## iter 10 value 79.316363
## iter 20 value 54.146303
## iter 30 value 51.926509
## iter 40 value 50.587084
## iter 50 value 50.489573
## iter 60 value 50.474210
## iter 70 value 50.458626
## iter 80 value 50.450965
## iter 90 value 50.444014
## iter 100 value 50.426712
## final value 50.426712
## stopped after 100 iterations
## # weights: 109
## initial value 102.457680
## iter 10 value 30.173731
## iter 20 value 23.577067
## iter 30 value 19.575446
## iter 40 value 18.388200
## iter 50 value 18.086073
## iter 60 value 17.781574
## iter 70 value 17.762747
## iter 80 value 17.523431
## iter 90 value 17.007868
## iter 100 value 16.837788
## final value 16.837788
## stopped after 100 iterations
## # weights: 179
## initial value 107.763843
## iter 10 value 12.222938
## iter 20 value 4.833238
## iter 30 value 2.547389
## iter 40 value 2.506760
## iter 50 value 2.465933
## iter 60 value 2.446618
## iter 70 value 2.429134
## iter 80 value 2.414251
## iter 90 value 2.039411
## iter 100 value 0.211096
## final value 0.211096
## stopped after 100 iterations
## # weights: 39
## initial value 104.472068
## iter 10 value 65.134966
## iter 20 value 58.039746
## iter 30 value 51.529865
## iter 40 value 46.482814
```

```
## iter 50 value 45.560272
## iter 60 value 45.233478
## iter 70 value 45.166909
## iter 80 value 45.128088
## iter 90 value 45.123443
## iter 100 value 45.121269
## final value 45.121269
## stopped after 100 iterations
## # weights: 109
## initial value 98.200797
## iter 10 value 22.463152
## iter 20 value 12.801040
## iter 30 value 8.665100
## iter 40 value 8.131949
## iter 50 value 6.254631
## iter 60 value 3.792582
## iter 70 value 3.665075
## iter 80 value 3.583041
## iter 90 value 3.501858
## iter 100 value 3.476684
## final value 3.476684
## stopped after 100 iterations
## # weights: 179
## initial value 106.456868
## iter 10 value 19.267289
## iter 20 value 4.937744
## iter 30 value 2.268550
## iter 40 value 2.250703
## iter 50 value 2.249501
## iter 60 value 2.249382
## iter 70 value 2.249351
## final value 2.249341
## converged
## # weights: 39
## initial value 103.966156
## iter 10 value 67.868376
## iter 20 value 63.303665
## iter 30 value 61.969825
## iter 40 value 61.070824
## iter 50 value 60.274006
## iter 60 value 58.627562
## iter 70 value 57.704211
## iter 80 value 57.312106
## iter 90 value 57.223031
## iter 100 value 57.126860
## final value 57.126860
## stopped after 100 iterations
## # weights: 109
## initial value 103.489716
## iter 10 value 56.695186
## iter 20 value 34.519226
## iter 30 value 21.327647
## iter 40 value 19.725156
## iter 50 value 19.147502
```

```
## iter 60 value 18.786956
## iter 70 value 16.128017
## iter 80 value 16.006829
## iter 90 value 16.004455
## final value 16.004440
## converged
## # weights: 179
## initial value 127.364733
## iter 10 value 55.537825
## iter 20 value 35.271283
## iter 30 value 19.268067
## iter 40 value 16.335774
## iter 50 value 14.685336
## iter 60 value 12.903358
## iter 70 value 12.752868
## iter 80 value 12.577489
## iter 90 value 12.198395
## iter 100 value 12.115448
## final value 12.115448
## stopped after 100 iterations
## # weights: 39
## initial value 109.354850
## iter 10 value 67.274959
## iter 20 value 61.111110
## iter 30 value 59.295381
## iter 40 value 58.688705
## iter 50 value 58.659685
## iter 60 value 58.177136
## iter 70 value 57.532969
## iter 80 value 57.032554
## iter 90 value 56.404495
## iter 100 value 55.235029
## final value 55.235029
## stopped after 100 iterations
## # weights: 109
## initial value 106.031369
## iter 10 value 33.048027
## iter 20 value 18.252925
## iter 30 value 13.226675
## iter 40 value 6.668048
## iter 50 value 5.953257
## iter 60 value 5.894459
## iter 70 value 5.854195
## iter 80 value 5.835521
## iter 90 value 4.823775
## iter 100 value 3.728382
## final value 3.728382
## stopped after 100 iterations
## # weights: 179
## initial value 113.749683
## iter 10 value 37.107881
## iter 20 value 9.860052
## iter 30 value 4.308719
## iter 40 value 3.625436
```

```
## iter 50 value 3.538512
## iter 60 value 1.664658
## iter 70 value 1.603890
## iter 80 value 1.589111
## iter 90 value 1.566951
## iter 100 value 0.461612
## final value 0.461612
## stopped after 100 iterations
## # weights: 39
## initial value 110.703885
## iter 10 value 80.488171
## iter 20 value 58.409982
## iter 30 value 55.031395
## iter 40 value 54.701676
## iter 50 value 54.433406
## iter 60 value 54.427639
## iter 70 value 54.390842
## iter 80 value 53.572547
## iter 90 value 53.355908
## iter 100 value 52.636314
## final value 52.636314
## stopped after 100 iterations
## # weights: 109
## initial value 112.777157
## iter 10 value 35.777908
## iter 20 value 14.676191
## iter 30 value 8.127686
## iter 40 value 6.662476
## iter 50 value 6.064842
## iter 60 value 5.332021
## iter 70 value 3.351237
## iter 80 value 3.299404
## iter 90 value 3.296430
## iter 100 value 3.295842
## final value 3.295842
## stopped after 100 iterations
## # weights: 179
## initial value 114.786911
## iter 10 value 37.376245
## iter 20 value 31.019746
## iter 30 value 30.268802
## iter 40 value 29.571426
## iter 50 value 28.070073
## iter 60 value 27.156619
## iter 70 value 26.394960
## iter 80 value 25.183531
## iter 90 value 24.321861
## iter 100 value 24.315693
## final value 24.315693
## stopped after 100 iterations
## # weights: 39
## initial value 102.442172
## iter 10 value 77.450999
## iter 20 value 71.361642
```

```
## iter 30 value 65.028247
## iter 40 value 59.644186
## iter 50 value 58.821794
## iter 60 value 58.733068
## iter 70 value 58.293790
## iter 80 value 58.185515
## iter 90 value 58.183982
## iter 90 value 58.183981
## iter 90 value 58.183981
## final value 58.183981
## converged
## # weights: 109
## initial value 117.453113
## iter 10 value 51.873481
## iter 20 value 30.692088
## iter 30 value 21.786803
## iter 40 value 17.152029
## iter 50 value 16.807359
## iter 60 value 16.761771
## iter 70 value 16.760704
## final value 16.760703
## converged
## # weights: 179
## initial value 97.760439
## iter 10 value 40.313885
## iter 20 value 18.795398
## iter 30 value 14.296333
## iter 40 value 12.979477
## iter 50 value 12.715987
## iter 60 value 12.525710
## iter 70 value 12.495610
## iter 80 value 12.494957
## final value 12.494950
## converged
## # weights: 39
## initial value 106.384231
## iter 10 value 86.192459
## iter 20 value 64.956486
## iter 30 value 63.513229
## iter 40 value 62.674825
## iter 50 value 62.053816
## iter 60 value 62.050996
## iter 70 value 62.036129
## iter 80 value 62.029184
## iter 90 value 62.025271
## iter 100 value 60.741352
## final value 60.741352
## stopped after 100 iterations
## # weights: 109
## initial value 105.295497
## iter 10 value 49.988130
## iter 20 value 35.420163
## iter 30 value 11.650889
## iter 40 value 2.601046
```

```
## iter 50 value 0.811794
## iter 60 value 0.701504
## iter 70 value 0.628957
## iter 80 value 0.517797
## iter 90 value 0.478542
## iter 100 value 0.439401
## final value 0.439401
## stopped after 100 iterations
## # weights: 179
## initial value 96.208628
## iter 10 value 22.444190
## iter 20 value 6.249756
## iter 30 value 4.750947
## iter 40 value 4.134210
## iter 50 value 3.953896
## iter 60 value 3.583218
## iter 70 value 3.565410
## iter 80 value 3.527629
## iter 90 value 3.495411
## iter 100 value 3.286140
## final value 3.286140
## stopped after 100 iterations
## # weights: 39
## initial value 99.113616
## iter 10 value 72.921939
## iter 20 value 65.653703
## iter 30 value 65.557822
## iter 40 value 65.356098
## iter 50 value 64.570840
## iter 60 value 64.547269
## iter 70 value 64.545236
## iter 80 value 64.540920
## iter 90 value 64.540150
## iter 90 value 64.540149
## iter 90 value 64.540149
## final value 64.540149
## converged
## # weights: 109
## initial value 100.265026
## iter 10 value 32.744203
## iter 20 value 9.347204
## iter 30 value 6.459242
## iter 40 value 6.445520
## iter 50 value 6.435164
## iter 60 value 6.108671
## final value 6.108652
## converged
## # weights: 179
## initial value 107.617677
## iter 10 value 20.094854
## iter 20 value 4.197975
## iter 30 value 2.568063
## iter 40 value 2.503684
## iter 50 value 2.502390
```

```
## iter 60 value 2.502017
## iter 70 value 2.502013
## iter 70 value 2.502013
## iter 70 value 2.502013
## final value 2.502013
## converged
## # weights: 39
## initial value 101.922734
## iter 10 value 67.704235
## iter 20 value 60.938457
## iter 30 value 60.304342
## iter 40 value 57.859377
## iter 50 value 57.228095
## iter 60 value 57.071092
## iter 70 value 56.961622
## iter 80 value 56.853297
## iter 90 value 56.844005
## iter 100 value 56.121219
## final value 56.121219
## stopped after 100 iterations
## # weights: 109
## initial value 114.608044
## iter 10 value 48.681598
## iter 20 value 27.725466
## iter 30 value 20.921050
## iter 40 value 17.400910
## iter 50 value 16.832208
## iter 60 value 16.247031
## iter 70 value 15.821144
## iter 80 value 15.776211
## iter 90 value 15.775778
## final value 15.775777
## converged
## # weights: 179
## initial value 108.577470
## iter 10 value 45.492580
## iter 20 value 19.062566
## iter 30 value 13.571778
## iter 40 value 12.130881
## iter 50 value 11.818093
## iter 60 value 11.708886
## iter 70 value 11.704983
## iter 80 value 11.704270
## iter 90 value 11.704177
## final value 11.704174
## converged
## # weights: 39
## initial value 118.384405
## iter 10 value 63.061188
## iter 20 value 56.866506
## iter 30 value 52.918963
## iter 40 value 48.782648
## iter 50 value 47.685378
## iter 60 value 46.647754
```

```
## iter 70 value 46.547462
## iter 80 value 46.498774
## iter 90 value 46.468622
## iter 100 value 46.441623
## final value 46.441623
## stopped after 100 iterations
## # weights: 109
## initial value 100.523905
## iter 10 value 49.667985
## iter 20 value 32.272071
## iter 30 value 28.077417
## iter 40 value 24.766138
## iter 50 value 22.748340
## iter 60 value 17.034307
## iter 70 value 16.156846
## iter 80 value 14.613581
## iter 90 value 13.986368
## iter 100 value 12.983057
## final value 12.983057
## stopped after 100 iterations
## # weights: 179
## initial value 110.598998
## iter 10 value 16.468345
## iter 20 value 6.948654
## iter 30 value 5.736880
## iter 40 value 5.412081
## iter 50 value 2.780738
## iter 60 value 2.073711
## iter 70 value 2.064831
## iter 80 value 2.039632
## iter 90 value 2.028936
## iter 100 value 2.015563
## final value 2.015563
## stopped after 100 iterations
## # weights: 39
## initial value 107.272666
## iter 10 value 82.800546
## iter 20 value 60.398541
## iter 30 value 54.001731
## iter 40 value 50.788222
## iter 50 value 44.272933
## iter 60 value 41.151978
## iter 70 value 40.323128
## iter 80 value 39.398607
## iter 90 value 38.842088
## iter 100 value 38.661066
## final value 38.661066
## stopped after 100 iterations
## # weights: 109
## initial value 100.410174
## iter 10 value 50.415440
## iter 20 value 32.138839
## iter 30 value 24.851638
## iter 40 value 19.395535
```

```
## iter 50 value 16.168833
## iter 60 value 14.631217
## iter 70 value 14.210980
## iter 80 value 12.372002
## iter 90 value 12.245508
## iter 100 value 12.077968
## final value 12.077968
## stopped after 100 iterations
## # weights: 179
## initial value 118.142310
## iter 10 value 30.004293
## iter 20 value 7.019661
## iter 30 value 5.051398
## iter 40 value 3.789894
## iter 50 value 3.669587
## iter 60 value 3.648646
## iter 70 value 1.503231
## iter 80 value 1.415385
## iter 90 value 1.389252
## iter 100 value 1.387698
## final value 1.387698
## stopped after 100 iterations
## # weights: 39
## initial value 97.008143
## iter 10 value 72.247719
## iter 20 value 62.286870
## iter 30 value 59.569785
## iter 40 value 59.020275
## iter 50 value 57.718665
## iter 60 value 57.447456
## iter 70 value 57.445067
## final value 57.445063
## converged
## # weights: 109
## initial value 124.568072
## iter 10 value 62.224584
## iter 20 value 37.463481
## iter 30 value 23.894316
## iter 40 value 18.082286
## iter 50 value 17.401659
## iter 60 value 17.057359
## iter 70 value 16.524935
## iter 80 value 16.427424
## iter 90 value 16.422662
## iter 100 value 16.421379
## final value 16.421379
## stopped after 100 iterations
## # weights: 179
## initial value 114.851308
## iter 10 value 53.512838
## iter 20 value 25.344277
## iter 30 value 14.305521
## iter 40 value 13.132389
## iter 50 value 12.439838
```

```
## iter 60 value 12.248077
## iter 70 value 12.170287
## iter 80 value 12.157901
## iter 90 value 12.154942
## iter 100 value 12.154343
## final value 12.154343
## stopped after 100 iterations
## # weights: 39
## initial value 102.543893
## iter 10 value 67.145239
## iter 20 value 59.782776
## iter 30 value 57.414551
## iter 40 value 56.409475
## iter 50 value 55.935739
## iter 60 value 55.707111
## iter 70 value 54.681224
## iter 80 value 54.117416
## iter 90 value 53.978357
## iter 100 value 53.874841
## final value 53.874841
## stopped after 100 iterations
## # weights: 109
## initial value 109.860170
## iter 10 value 39.899647
## iter 20 value 16.202797
## iter 30 value 11.704230
## iter 40 value 11.233258
## iter 50 value 11.019408
## iter 60 value 10.904546
## iter 70 value 10.880054
## iter 80 value 6.812379
## iter 90 value 6.619221
## iter 100 value 6.581932
## final value 6.581932
## stopped after 100 iterations
## # weights: 179
## initial value 112.124048
## iter 10 value 24.134817
## iter 20 value 9.305798
## iter 30 value 6.521452
## iter 40 value 6.345559
## iter 50 value 6.243299
## iter 60 value 5.923184
## iter 70 value 5.867447
## iter 80 value 5.829981
## iter 90 value 5.761198
## iter 100 value 5.163069
## final value 5.163069
## stopped after 100 iterations
## # weights: 39
## initial value 110.634912
## iter 10 value 68.182465
## iter 20 value 56.377025
## iter 30 value 52.504603
```

```
## iter 40 value 52.058771
## iter 50 value 52.026436
## iter 60 value 52.021981
## iter 70 value 50.963324
## iter 80 value 50.921062
## iter 90 value 50.920852
## iter 100 value 50.920761
## final value 50.920761
## stopped after 100 iterations
## # weights: 109
## initial value 113.358653
## iter 10 value 38.396240
## iter 20 value 14.639965
## iter 30 value 10.721241
## iter 40 value 7.707064
## iter 50 value 7.699491
## iter 60 value 7.544705
## iter 70 value 7.092438
## iter 80 value 6.215390
## iter 90 value 6.214786
## iter 100 value 6.214691
## final value 6.214691
## stopped after 100 iterations
## # weights: 179
## initial value 107.994989
## iter 10 value 13.797898
## iter 20 value 2.024231
## iter 30 value 1.396996
## iter 40 value 1.393423
## iter 50 value 1.390854
## iter 60 value 1.389382
## iter 70 value 1.388439
## iter 80 value 1.387100
## iter 90 value 1.386784
## iter 100 value 1.385549
## final value 1.385549
## stopped after 100 iterations
## # weights: 39
## initial value 114.138293
## iter 10 value 68.792427
## iter 20 value 65.723801
## iter 30 value 64.272385
## iter 40 value 62.146904
## iter 50 value 60.602517
## iter 60 value 60.202512
## iter 70 value 60.199203
## final value 60.199192
## converged
## # weights: 109
## initial value 105.992930
## iter 10 value 53.257733
## iter 20 value 32.496133
## iter 30 value 22.352638
## iter 40 value 17.205258
```

```
## iter 50 value 16.997841
## iter 60 value 16.760333
## iter 70 value 16.740466
## final value 16.740360
## converged
## # weights: 179
## initial value 105.287521
## iter 10 value 46.482191
## iter 20 value 20.111151
## iter 30 value 14.254839
## iter 40 value 13.770123
## iter 50 value 13.301561
## iter 60 value 12.872573
## iter 70 value 12.610167
## iter 80 value 12.587757
## iter 90 value 12.581930
## iter 100 value 12.580992
## final value 12.580992
## stopped after 100 iterations
## # weights: 39
## initial value 100.356291
## iter 10 value 63.710645
## iter 20 value 59.115044
## iter 30 value 57.022947
## iter 40 value 52.301288
## iter 50 value 50.694020
## iter 60 value 50.668459
## iter 70 value 50.654261
## iter 80 value 50.650035
## iter 90 value 50.638297
## iter 100 value 50.609486
## final value 50.609486
## stopped after 100 iterations
## # weights: 109
## initial value 114.680992
## iter 10 value 32.293269
## iter 20 value 19.073418
## iter 30 value 12.460581
## iter 40 value 11.133690
## iter 50 value 10.905723
## iter 60 value 9.751756
## iter 70 value 8.308948
## iter 80 value 6.792270
## iter 90 value 6.630285
## iter 100 value 4.685427
## final value 4.685427
## stopped after 100 iterations
## # weights: 179
## initial value 107.312952
## iter 10 value 46.890948
## iter 20 value 26.889055
## iter 30 value 10.683660
## iter 40 value 7.244474
## iter 50 value 4.646879
```

```
## iter 60 value 3.150502
## iter 70 value 3.090480
## iter 80 value 2.910468
## iter 90 value 2.601603
## iter 100 value 2.470557
## final value 2.470557
## stopped after 100 iterations
## # weights: 39
## initial value 101.962831
## iter 10 value 61.687403
## iter 20 value 49.229500
## iter 30 value 47.237994
## iter 40 value 43.575805
## iter 50 value 42.249117
## iter 60 value 41.237781
## iter 70 value 40.361657
## iter 80 value 40.248338
## iter 90 value 40.098623
## iter 100 value 39.221909
## final value 39.221909
## stopped after 100 iterations
## # weights: 109
## initial value 121.703694
## iter 10 value 34.681561
## iter 20 value 18.935310
## iter 30 value 13.556597
## iter 40 value 9.527864
## iter 50 value 8.961804
## iter 60 value 7.128828
## iter 70 value 6.990757
## iter 80 value 6.856523
## iter 90 value 6.824144
## iter 100 value 4.347021
## final value 4.347021
## stopped after 100 iterations
## # weights: 179
## initial value 107.825514
## iter 10 value 19.419906
## iter 20 value 4.785055
## iter 30 value 3.048851
## iter 40 value 2.986383
## iter 50 value 2.874263
## iter 60 value 2.872770
## iter 70 value 2.872438
## iter 80 value 2.871881
## iter 90 value 2.871440
## iter 100 value 2.870835
## final value 2.870835
## stopped after 100 iterations
## # weights: 39
## initial value 103.157614
## iter 10 value 79.007399
## iter 20 value 63.890503
## iter 30 value 57.914178
```

```
## iter 40 value 57.381019
## iter 50 value 57.274709
## iter 60 value 57.274391
## iter 70 value 57.273781
## iter 80 value 57.225736
## iter 90 value 57.010591
## iter 100 value 56.997700
## final value 56.997700
## stopped after 100 iterations
## # weights: 109
## initial value 129.036942
## iter 10 value 90.194555
## iter 20 value 53.933354
## iter 30 value 31.591529
## iter 40 value 23.068586
## iter 50 value 21.534505
## iter 60 value 19.908995
## iter 70 value 19.499698
## iter 80 value 17.482025
## iter 90 value 17.081348
## iter 100 value 16.889400
## final value 16.889400
## stopped after 100 iterations
## # weights: 179
## initial value 117.924008
## iter 10 value 40.720678
## iter 20 value 19.860861
## iter 30 value 14.092558
## iter 40 value 13.261610
## iter 50 value 12.914291
## iter 60 value 12.834171
## iter 70 value 12.536180
## iter 80 value 12.278690
## iter 90 value 12.145686
## iter 100 value 12.130243
## final value 12.130243
## stopped after 100 iterations
## # weights: 39
## initial value 105.515112
## iter 10 value 64.161986
## iter 20 value 58.085071
## iter 30 value 55.082193
## iter 40 value 53.278238
## iter 50 value 52.878081
## iter 60 value 52.718081
## iter 70 value 51.289594
## iter 80 value 51.239049
## iter 90 value 50.490691
## iter 100 value 50.476307
## final value 50.476307
## stopped after 100 iterations
## # weights: 109
## initial value 107.663158
## iter 10 value 30.395845
```

```
## iter 20 value 16.338048
## iter 30 value 10.750987
## iter 40 value 9.989274
## iter 50 value 9.895840
## iter 60 value 9.882751
## iter 70 value 9.872846
## iter 80 value 6.462267
## iter 90 value 6.131068
## iter 100 value 6.067874
## final value 6.067874
## stopped after 100 iterations
## # weights: 179
## initial value 108.517010
## iter 10 value 17.789334
## iter 20 value 3.311656
## iter 30 value 2.230421
## iter 40 value 2.189777
## iter 50 value 2.159703
## iter 60 value 2.131966
## iter 70 value 2.106793
## iter 80 value 2.091404
## iter 90 value 2.076319
## iter 100 value 2.065186
## final value 2.065186
## stopped after 100 iterations
## # weights: 39
## initial value 108.925966
## iter 10 value 72.821162
## iter 20 value 61.877328
## iter 30 value 57.854829
## iter 40 value 56.727041
## iter 50 value 53.732413
## iter 60 value 52.935870
## iter 70 value 52.856231
## iter 80 value 52.832364
## iter 90 value 52.819567
## iter 100 value 52.813237
## final value 52.813237
## stopped after 100 iterations
## # weights: 109
## initial value 106.074296
## iter 10 value 49.431645
## iter 20 value 29.766269
## iter 30 value 26.193697
## iter 40 value 25.362006
## iter 50 value 22.372311
## iter 60 value 21.124044
## iter 70 value 20.134364
## iter 80 value 19.631727
## iter 90 value 19.582088
## iter 100 value 19.536727
## final value 19.536727
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 94.748098
## iter 10 value 18.101053
## iter 20 value 5.988878
## iter 30 value 2.600707
## iter 40 value 2.520008
## iter 50 value 2.503619
## iter 60 value 2.502046
## iter 70 value 2.502038
## final value 2.502015
## converged
## # weights: 39
## initial value 100.336601
## iter 10 value 77.676649
## iter 20 value 65.451032
## iter 30 value 60.880861
## iter 40 value 59.277350
## iter 50 value 58.744266
## iter 60 value 58.620230
## final value 58.619144
## converged
## # weights: 109
## initial value 105.408318
## iter 10 value 60.048458
## iter 20 value 40.369959
## iter 30 value 22.700958
## iter 40 value 19.381992
## iter 50 value 17.120847
## iter 60 value 15.918277
## iter 70 value 15.797804
## iter 80 value 15.797072
## final value 15.797070
## converged
## # weights: 179
## initial value 108.402575
## iter 10 value 46.928294
## iter 20 value 22.986062
## iter 30 value 14.016179
## iter 40 value 12.312959
## iter 50 value 12.193953
## iter 60 value 12.067436
## iter 70 value 12.032158
## iter 80 value 12.026064
## iter 90 value 12.025564
## iter 100 value 12.025479
## final value 12.025479
## stopped after 100 iterations
## # weights: 39
## initial value 106.397803
## iter 10 value 74.343234
## iter 20 value 69.506666
## iter 30 value 69.422174
## iter 40 value 69.295797
## iter 50 value 68.019008
## iter 60 value 67.694889
```

```
## iter 70 value 67.367767
## iter 80 value 67.334361
## iter 90 value 67.272162
## iter 100 value 67.232091
## final value 67.232091
## stopped after 100 iterations
## # weights: 109
## initial value 99.335807
## iter 10 value 36.136154
## iter 20 value 22.185656
## iter 30 value 11.520518
## iter 40 value 11.246073
## iter 50 value 10.694692
## iter 60 value 9.918221
## iter 70 value 9.873059
## iter 80 value 9.855063
## iter 90 value 9.830773
## iter 100 value 9.211932
## final value 9.211932
## stopped after 100 iterations
## # weights: 179
## initial value 103.370245
## iter 10 value 17.600029
## iter 20 value 0.818697
## iter 30 value 0.209391
## iter 40 value 0.185497
## iter 50 value 0.167523
## iter 60 value 0.152191
## iter 70 value 0.142294
## iter 80 value 0.134822
## iter 90 value 0.126952
## iter 100 value 0.116865
## final value 0.116865
## stopped after 100 iterations
## # weights: 39
## initial value 118.386989
## iter 10 value 70.272344
## iter 20 value 63.872142
## iter 30 value 61.895551
## iter 40 value 59.146511
## iter 50 value 58.699191
## iter 60 value 58.658716
## iter 70 value 58.453669
## iter 80 value 56.263080
## iter 90 value 54.110197
## iter 100 value 53.226084
## final value 53.226084
## stopped after 100 iterations
## # weights: 109
## initial value 124.802183
## iter 10 value 66.778583
## iter 20 value 44.573250
## iter 30 value 36.478326
## iter 40 value 35.292182
```

```
## iter 50 value 34.896069
## iter 60 value 33.868053
## iter 70 value 33.196597
## iter 80 value 30.605350
## iter 90 value 29.558017
## iter 100 value 27.928307
## final value 27.928307
## stopped after 100 iterations
## # weights: 179
## initial value 117.588671
## iter 10 value 27.697363
## iter 20 value 13.736849
## iter 30 value 10.406576
## iter 40 value 7.989163
## iter 50 value 7.711119
## iter 60 value 6.906016
## iter 70 value 6.821782
## iter 80 value 6.806915
## iter 90 value 6.797513
## iter 100 value 6.797121
## final value 6.797121
## stopped after 100 iterations
## # weights: 39
## initial value 103.102643
## iter 10 value 76.090873
## iter 20 value 68.032499
## iter 30 value 63.185168
## iter 40 value 56.705649
## iter 50 value 55.388693
## iter 60 value 55.286144
## iter 70 value 55.283618
## final value 55.283611
## converged
## # weights: 109
## initial value 102.712415
## iter 10 value 53.974792
## iter 20 value 33.187841
## iter 30 value 20.496184
## iter 40 value 16.985953
## iter 50 value 16.768001
## iter 60 value 16.717314
## iter 70 value 16.713962
## final value 16.713956
## converged
## # weights: 179
## initial value 117.899601
## iter 10 value 42.196434
## iter 20 value 21.850921
## iter 30 value 16.212846
## iter 40 value 13.845541
## iter 50 value 13.620207
## iter 60 value 13.570854
## iter 70 value 13.508730
## iter 80 value 13.508508
```

```
## iter 80 value 13.508508
## iter 80 value 13.508508
## final value 13.508508
## converged
## # weights: 39
## initial value 98.204453
## iter 10 value 66.925777
## iter 20 value 62.607699
## iter 30 value 60.918802
## iter 40 value 59.954112
## iter 50 value 57.280458
## iter 60 value 54.676078
## iter 70 value 52.782712
## iter 80 value 50.191143
## iter 90 value 50.087293
## iter 100 value 50.061071
## final value 50.061071
## stopped after 100 iterations
## # weights: 109
## initial value 105.241956
## iter 10 value 33.231360
## iter 20 value 12.857632
## iter 30 value 12.093625
## iter 40 value 11.859150
## iter 50 value 11.594309
## iter 60 value 11.401037
## iter 70 value 10.979020
## iter 80 value 10.574763
## iter 90 value 10.165482
## iter 100 value 10.058526
## final value 10.058526
## stopped after 100 iterations
## # weights: 179
## initial value 102.444457
## iter 10 value 31.587908
## iter 20 value 11.404162
## iter 30 value 8.266415
## iter 40 value 6.359106
## iter 50 value 3.739210
## iter 60 value 2.219378
## iter 70 value 1.224901
## iter 80 value 0.445671
## iter 90 value 0.363417
## iter 100 value 0.340866
## final value 0.340866
## stopped after 100 iterations
## # weights: 39
## initial value 98.605406
## iter 10 value 57.210617
## iter 20 value 52.106117
## iter 30 value 46.303226
## iter 40 value 41.827345
## iter 50 value 40.113776
## iter 60 value 40.058649
```

```
## iter 70 value 39.924372
## iter 80 value 39.252612
## iter 90 value 39.239602
## iter 100 value 39.238164
## final value 39.238164
## stopped after 100 iterations
## # weights: 109
## initial value 106.906179
## iter 10 value 54.206539
## iter 20 value 21.737477
## iter 30 value 12.047800
## iter 40 value 6.739788
## iter 50 value 4.526163
## iter 60 value 3.875355
## iter 70 value 3.691428
## iter 80 value 3.080167
## iter 90 value 2.676606
## iter 100 value 2.600660
## final value 2.600660
## stopped after 100 iterations
## # weights: 179
## initial value 112.221956
## iter 10 value 46.916463
## iter 20 value 40.083478
## iter 30 value 38.383812
## iter 40 value 36.751141
## iter 50 value 33.822764
## iter 60 value 32.927865
## iter 70 value 31.263953
## iter 80 value 25.110353
## iter 90 value 15.158420
## iter 100 value 11.506238
## final value 11.506238
## stopped after 100 iterations
## # weights: 39
## initial value 107.494255
## iter 10 value 82.514884
## iter 20 value 67.480589
## iter 30 value 59.585368
## iter 40 value 56.144077
## iter 50 value 55.831848
## iter 60 value 55.823251
## final value 55.823237
## converged
## # weights: 109
## initial value 113.692212
## iter 10 value 37.322898
## iter 20 value 23.714628
## iter 30 value 17.691700
## iter 40 value 16.320294
## iter 50 value 15.954147
## iter 60 value 15.930955
## iter 70 value 15.929962
## iter 80 value 15.929845
```

```
## final value 15.929841
## converged
## # weights: 179
## initial value 118.587377
## iter 10 value 32.762435
## iter 20 value 16.464925
## iter 30 value 13.269555
## iter 40 value 12.109244
## iter 50 value 11.565279
## iter 60 value 11.388141
## iter 70 value 11.359954
## iter 80 value 11.358902
## final value 11.358894
## converged
## # weights: 39
## initial value 101.267206
## iter 10 value 66.116622
## iter 20 value 60.371914
## iter 30 value 60.075559
## iter 40 value 58.924149
## iter 50 value 57.124919
## iter 60 value 57.090024
## iter 70 value 56.411338
## iter 80 value 55.576740
## iter 90 value 55.573734
## iter 100 value 55.291518
## final value 55.291518
## stopped after 100 iterations
## # weights: 109
## initial value 113.510219
## iter 10 value 19.640264
## iter 20 value 14.613993
## iter 30 value 14.115658
## iter 40 value 14.045824
## iter 50 value 14.005903
## iter 60 value 13.979677
## iter 70 value 13.963240
## iter 80 value 13.911020
## iter 90 value 12.171198
## iter 100 value 11.978153
## final value 11.978153
## stopped after 100 iterations
## # weights: 179
## initial value 119.859131
## iter 10 value 20.329641
## iter 20 value 3.312294
## iter 30 value 0.480520
## iter
       40 value 0.320359
## iter 50 value 0.263547
## iter 60 value 0.205837
## iter 70 value 0.185068
## iter 80 value 0.155443
## iter 90 value 0.139788
## iter 100 value 0.123685
```

```
## final value 0.123685
## stopped after 100 iterations
## # weights: 39
## initial value 101.812966
## iter 10 value 76.264657
## iter 20 value 65.880441
## iter 30 value 63.269674
## iter 40 value 62.040211
## iter 50 value 60.670281
## iter 60 value 58.861143
## iter 70 value 56.116332
## iter 80 value 54.090716
## iter 90 value 53.300410
## iter 100 value 52.969216
## final value 52.969216
## stopped after 100 iterations
## # weights: 109
## initial value 112.752655
## iter 10 value 56.817290
## iter 20 value 47.929750
## iter 30 value 43.441166
## iter 40 value 40.165341
## iter 50 value 39.680246
## iter 60 value 39.069894
## iter 70 value 38.901033
## iter 80 value 38.321340
## iter 90 value 38.206297
## iter 100 value 38.188447
## final value 38.188447
## stopped after 100 iterations
## # weights: 179
## initial value 102.980192
## iter 10 value 28.642033
## iter 20 value 7.009092
## iter 30 value 2.326874
## iter 40 value 1.920285
## iter 50 value 1.910080
## iter 60 value 1.909703
## iter 70 value 1.909601
## iter 80 value 1.909547
## iter 90 value 1.909543
## final value 1.909543
## converged
## # weights: 39
## initial value 108.018927
## iter 10 value 76.035250
## iter 20 value 70.182287
## iter 30 value 67.485350
## iter 40 value 64.979569
## iter 50 value 64.693036
## final value 64.691271
## converged
## # weights: 109
## initial value 110.894363
```

```
## iter 10 value 58.984149
## iter 20 value 34.711324
## iter 30 value 18.901614
## iter 40 value 16.341994
## iter 50 value 16.147388
## iter 60 value 16.139168
## iter 70 value 16.139106
## final value 16.139106
## converged
## # weights: 179
## initial value 110.622216
## iter 10 value 45.906126
## iter 20 value 18.942056
## iter 30 value 13.363754
## iter 40 value 12.971982
## iter 50 value 12.732662
## iter 60 value 12.534867
## iter 70 value 12.325362
## iter 80 value 12.154454
## iter 90 value 12.088094
## iter 100 value 11.968421
## final value 11.968421
## stopped after 100 iterations
## # weights: 39
## initial value 123.917267
## iter 10 value 59.369905
## iter 20 value 52.682245
## iter 30 value 51.632216
## iter 40 value 50.580369
## iter 50 value 49.876959
## iter 60 value 49.408513
## iter 70 value 49.205991
## iter 80 value 48.987359
## iter 90 value 46.115187
## iter 100 value 45.996281
## final value 45.996281
## stopped after 100 iterations
## # weights: 109
## initial value 123.525921
## iter 10 value 36.797560
## iter 20 value 14.696041
## iter 30 value 10.193233
## iter 40 value 7.424524
## iter 50 value 6.384277
## iter 60 value 6.321771
## iter 70 value 6.309662
## iter 80 value 6.299372
## iter 90 value 6.288918
## iter 100 value 6.275993
## final value 6.275993
## stopped after 100 iterations
## # weights: 179
## initial value 119.617561
## iter 10 value 10.701262
```

```
## iter 20 value 3.743690
## iter 30 value 3.512921
## iter 40 value 3.498513
## iter 50 value 3.473922
## iter 60 value 3.012627
## iter 70 value 0.283244
## iter 80 value 0.165764
## iter 90 value 0.156894
## iter 100 value 0.146648
## final value 0.146648
## stopped after 100 iterations
## # weights: 39
## initial value 101.343288
## iter 10 value 68.317580
## iter 20 value 61.609850
## iter 30 value 60.310510
## iter 40 value 58.972025
## iter 50 value 58.260087
## iter 60 value 56.263435
## iter 70 value 55.072481
## iter 80 value 53.620025
## iter 90 value 47.219951
## iter 100 value 45.520513
## final value 45.520513
## stopped after 100 iterations
## # weights: 109
## initial value 104.532899
## iter 10 value 35.538587
## iter 20 value 21.006508
## iter 30 value 18.979016
## iter 40 value 17.966696
## iter 50 value 17.896766
## iter 60 value 17.884063
## iter 70 value 17.846440
## iter 80 value 17.828289
## iter 90 value 17.816807
## iter 100 value 17.805054
## final value 17.805054
## stopped after 100 iterations
## # weights: 179
## initial value 119.528535
## iter 10 value 31.080294
## iter 20 value 12.244218
## iter 30 value 8.298214
## iter 40 value 6.722363
## iter 50 value 3.319663
## iter 60 value 0.371075
## iter 70 value 0.080728
## iter 80 value 0.008454
## iter 90 value 0.002261
## iter 100 value 0.000759
## final value 0.000759
## stopped after 100 iterations
## # weights: 39
```

```
## initial value 103.907486
## iter 10 value 78.223618
## iter 20 value 67.520408
## iter 30 value 61.734353
## iter 40 value 60.454840
## iter 50 value 59.330377
## iter 60 value 59.231196
## final value 59.230475
## converged
## # weights: 109
## initial value 100.840781
## iter 10 value 64.015531
## iter 20 value 36.767201
## iter 30 value 21.924690
## iter 40 value 18.761530
## iter 50 value 18.154767
## iter 60 value 17.188100
## iter 70 value 16.781006
## iter 80 value 16.707305
## iter 90 value 16.705089
## iter 100 value 16.704761
## final value 16.704761
## stopped after 100 iterations
## # weights: 179
## initial value 140.442939
## iter 10 value 65.830523
## iter 20 value 34.889380
## iter 30 value 19.035935
## iter 40 value 13.909923
## iter 50 value 12.793348
## iter 60 value 12.179300
## iter 70 value 12.051163
## iter 80 value 12.044959
## iter 90 value 12.044676
## final value 12.044675
## converged
## # weights: 39
## initial value 104.845205
## iter 10 value 68.660742
## iter 20 value 66.842425
## iter 30 value 66.341481
## iter 40 value 65.556342
## iter 50 value 64.031416
## iter 60 value 63.760341
## iter 70 value 63.007520
## iter 80 value 59.758458
## iter 90 value 58.836889
## iter 100 value 57.158233
## final value 57.158233
## stopped after 100 iterations
## # weights: 109
## initial value 113.894407
## iter 10 value 34.451271
## iter 20 value 21.274297
```

```
## iter 30 value 19.963719
## iter 40 value 18.595310
## iter 50 value 18.331691
## iter 60 value 16.834385
## iter 70 value 14.696309
## iter 80 value 13.042089
## iter 90 value 12.921985
## iter 100 value 12.671066
## final value 12.671066
## stopped after 100 iterations
## # weights: 179
## initial value 102.311419
## iter 10 value 28.973385
## iter 20 value 10.704907
## iter 30 value 6.771526
## iter 40 value 6.551950
## iter 50 value 6.507542
## iter 60 value 6.469068
## iter 70 value 6.436236
## iter 80 value 2.923132
## iter 90 value 2.611452
## iter 100 value 2.501461
## final value 2.501461
## stopped after 100 iterations
## # weights: 39
## initial value 108.875625
## iter 10 value 69.392753
## iter 20 value 58.759166
## iter 30 value 53.945300
## iter 40 value 50.712878
## iter 50 value 49.909091
## iter 60 value 49.811994
## iter 70 value 49.799894
## iter 80 value 49.796642
## iter 90 value 49.793880
## iter 100 value 49.793101
## final value 49.793101
## stopped after 100 iterations
## # weights: 109
## initial value 109.152712
## iter 10 value 51.267875
## iter 20 value 27.524327
## iter 30 value 24.488584
## iter 40 value 18.540232
## iter 50 value 17.193639
## iter 60 value 13.980719
## iter 70 value 13.089069
## iter 80 value 13.075474
## iter 90 value 11.481078
## iter 100 value 9.438517
## final value 9.438517
## stopped after 100 iterations
## # weights: 179
## initial value 101.701451
```

```
## iter 10 value 13.721538
## iter 20 value 3.396775
## iter 30 value 0.056290
## iter 40 value 0.004440
## iter 50 value 0.000872
## iter 60 value 0.000669
## iter 70 value 0.000135
## final value 0.000098
## converged
## # weights: 39
## initial value 111.190148
## iter 10 value 76.789437
## iter 20 value 61.225167
## iter 30 value 58.633997
## iter 40 value 57.665096
## iter 50 value 57.512190
## iter 60 value 57.506280
## final value 57.506275
## converged
## # weights: 109
## initial value 108.894507
## iter 10 value 53.697986
## iter 20 value 32.308356
## iter 30 value 19.677219
## iter 40 value 17.305521
## iter 50 value 17.273404
## iter 60 value 17.215712
## iter 70 value 17.196846
## final value 17.196768
## converged
## # weights: 179
## initial value 101.106225
## iter 10 value 34.196886
## iter 20 value 23.177652
## iter 30 value 15.454677
## iter 40 value 12.687252
## iter 50 value 12.105685
## iter 60 value 12.067431
## iter 70 value 12.066501
## final value 12.066480
## converged
## # weights: 39
## initial value 109.381752
## iter 10 value 72.643184
## iter 20 value 56.692340
## iter 30 value 55.597493
## iter 40 value 54.487736
## iter 50 value 51.299320
## iter 60 value 50.347516
## iter 70 value 46.912685
## iter 80 value 45.439339
## iter 90 value 45.050521
## iter 100 value 44.822432
## final value 44.822432
```

```
## stopped after 100 iterations
## # weights: 109
## initial value 109.943663
## iter 10 value 24.163936
## iter 20 value 17.127179
## iter 30 value 13.224191
## iter 40 value 11.606982
## iter 50 value 11.501437
## iter 60 value 11.244020
## iter 70 value 11.109880
## iter 80 value 10.837208
## iter 90 value 10.815140
## iter 100 value 10.395669
## final value 10.395669
## stopped after 100 iterations
## # weights: 179
## initial value 121.821713
## iter 10 value 32.238282
## iter 20 value 12.867662
## iter 30 value 10.981383
## iter 40 value 10.736350
## iter 50 value 10.689823
## iter 60 value 10.489605
## iter 70 value 8.674168
## iter 80 value 8.029012
## iter 90 value 7.941209
## iter 100 value 7.480538
## final value 7.480538
## stopped after 100 iterations
## # weights: 39
## initial value 99.698071
## iter 10 value 75.609142
## iter 20 value 56.276058
## iter 30 value 50.798012
## iter 40 value 49.171610
## iter 50 value 49.167365
## iter 60 value 47.619038
## iter 70 value 47.592538
## iter 80 value 47.588570
## iter 90 value 47.208361
## iter 100 value 45.883212
## final value 45.883212
## stopped after 100 iterations
## # weights: 109
## initial value 112.228710
## iter 10 value 50.031829
## iter 20 value 25.358033
## iter 30 value 22.199277
## iter 40 value 19.918972
## iter 50 value 19.823441
## iter 60 value 19.659080
## iter 70 value 16.520345
## iter 80 value 15.443672
## iter 90 value 15.062297
```

```
## iter 100 value 14.954324
## final value 14.954324
## stopped after 100 iterations
## # weights: 179
## initial value 103.322567
## iter 10 value 42.959707
## iter 20 value 25.348399
## iter 30 value 19.913674
## iter 40 value 17.998578
## iter 50 value 17.644034
## iter 60 value 14.539256
## iter 70 value 14.177409
## iter 80 value 14.117585
## iter 90 value 14.091361
## iter 100 value 14.054103
## final value 14.054103
## stopped after 100 iterations
## # weights: 39
## initial value 102.325985
## iter 10 value 75.244761
## iter 20 value 65.196442
## iter 30 value 59.542263
## iter 40 value 57.397302
## iter 50 value 57.323707
## final value 57.323341
## converged
## # weights: 109
## initial value 114.712539
## iter 10 value 69.088881
## iter 20 value 41.515291
## iter 30 value 25.744990
## iter 40 value 19.023144
## iter 50 value 17.314570
## iter 60 value 16.517578
## iter 70 value 16.355615
## iter 80 value 16.335146
## iter 90 value 16.333710
## final value 16.333697
## converged
## # weights: 179
## initial value 116.393871
## iter 10 value 44.575309
## iter 20 value 22.774751
## iter 30 value 16.367071
## iter 40 value 13.689312
## iter 50 value 12.410021
## iter 60 value 12.303035
## iter 70 value 12.291042
## iter 80 value 12.289979
## iter 90 value 12.289947
## final value 12.289947
## converged
## # weights: 39
## initial value 108.579175
```

```
## iter 10 value 70.042935
## iter 20 value 65.978343
## iter 30 value 62.711364
## iter 40 value 62.315902
## iter 50 value 62.028357
## iter 60 value 61.367452
## iter 70 value 58.288987
## iter 80 value 57.072904
## iter 90 value 55.597004
## iter 100 value 55.121204
## final value 55.121204
## stopped after 100 iterations
## # weights: 109
## initial value 110.054935
## iter 10 value 37.239104
## iter 20 value 22.093682
## iter 30 value 21.206249
## iter 40 value 21.051862
## iter 50 value 20.654159
## iter 60 value 19.862357
## iter 70 value 18.192534
## iter 80 value 17.827389
## iter 90 value 17.544070
## iter 100 value 17.447609
## final value 17.447609
## stopped after 100 iterations
## # weights: 179
## initial value 107.010114
## iter 10 value 28.645337
## iter 20 value 15.560719
## iter 30 value 5.357644
## iter 40 value 3.887256
## iter 50 value 3.044751
## iter 60 value 2.508667
## iter 70 value 2.436001
## iter 80 value 2.424734
## iter 90 value 2.413475
## iter 100 value 2.400548
## final value 2.400548
## stopped after 100 iterations
## # weights: 39
## initial value 111.748781
## iter 10 value 82.362494
## iter 20 value 56.491317
## iter 30 value 54.013211
## iter 40 value 52.297034
## iter 50 value 49.645258
## iter
       60 value 48.590780
## iter 70 value 47.176091
## iter 80 value 45.335375
## iter 90 value 45.079646
## iter 100 value 44.855973
## final value 44.855973
## stopped after 100 iterations
```

```
## # weights: 109
## initial value 100.959771
## iter 10 value 43.658244
## iter 20 value 14.136724
## iter 30 value 8.297663
## iter 40 value 4.640538
## iter 50 value 3.628200
## iter 60 value 3.187873
## iter 70 value 3.168525
## iter 80 value 3.038511
## iter 90 value 3.030469
## iter 100 value 3.022261
## final value 3.022261
## stopped after 100 iterations
## # weights: 179
## initial value 106.894923
## iter 10 value 14.562593
## iter 20 value 3.970936
## iter 30 value 2.851469
## iter 40 value 2.217390
## iter 50 value 0.212860
## iter 60 value 0.010088
## iter 70 value 0.004685
## iter 80 value 0.002399
## iter 90 value 0.001474
## iter 100 value 0.000656
## final value 0.000656
## stopped after 100 iterations
## # weights: 39
## initial value 108.294662
## iter 10 value 71.543210
## iter 20 value 63.169549
## iter 30 value 60.248837
## iter 40 value 59.703745
## iter 50 value 59.093686
## iter 60 value 58.940329
## iter 70 value 58.934163
## final value 58.934057
## converged
## # weights: 109
## initial value 107.229510
## iter 10 value 66.099796
## iter 20 value 39.788575
## iter 30 value 29.661846
## iter 40 value 21.603026
## iter 50 value 19.549133
## iter 60 value 17.516138
## iter 70 value 16.866239
## iter 80 value 16.152605
## iter 90 value 15.990962
## iter 100 value 15.980538
## final value 15.980538
## stopped after 100 iterations
## # weights: 179
```

```
## initial value 110.630154
## iter 10 value 35.885232
## iter 20 value 18.442221
## iter 30 value 14.937750
## iter 40 value 13.635050
## iter 50 value 13.154991
## iter 60 value 13.100098
## iter 70 value 13.096890
## iter 80 value 13.096741
## final value 13.096741
## converged
## # weights: 39
## initial value 103.937202
## iter 10 value 56.209423
## iter 20 value 54.194496
## iter 30 value 53.315345
## iter 40 value 53.307815
## iter 50 value 52.983382
## iter 60 value 52.686854
## iter 70 value 52.680336
## iter 80 value 52.648043
## iter 90 value 52.026765
## iter 100 value 52.000255
## final value 52.000255
## stopped after 100 iterations
## # weights: 109
## initial value 147.810753
## iter 10 value 70.247415
## iter 20 value 31.141529
## iter 30 value 16.301170
## iter 40 value 11.499538
## iter 50 value 11.207524
## iter 60 value 9.839623
## iter 70 value 9.784450
## iter 80 value 9.502959
## iter 90 value 9.335818
## iter 100 value 9.049286
## final value 9.049286
## stopped after 100 iterations
## # weights: 179
## initial value 108.066521
## iter 10 value 53.924505
## iter 20 value 23.229850
## iter 30 value 19.255641
## iter 40 value 17.431977
## iter 50 value 17.086157
## iter 60 value 16.632460
## iter 70 value 16.386048
## iter 80 value 15.729020
## iter 90 value 15.317498
## iter 100 value 14.715191
## final value 14.715191
## stopped after 100 iterations
## # weights: 39
```

```
## initial value 109.455309
## iter 10 value 66.527842
## iter 20 value 53.468029
## iter 30 value 51.838740
## iter 40 value 51.615120
## iter 50 value 51.578367
## iter 60 value 51.565644
## iter 70 value 51.554775
## iter 80 value 51.551139
## iter 90 value 51.550316
## final value 51.549851
## converged
## # weights: 109
## initial value 108.655688
## iter 10 value 22.018061
## iter 20 value 8.796169
## iter 30 value 7.486149
## iter 40 value 3.371172
## iter 50 value 3.274088
## iter 60 value 2.821294
## iter 70 value 1.489517
## iter 80 value 1.405045
## iter 90 value 1.393802
## iter 100 value 1.390219
## final value 1.390219
## stopped after 100 iterations
## # weights: 179
## initial value 105.730242
## iter 10 value 32.262044
## iter 20 value 16.254203
## iter 30 value 14.906651
## iter 40 value 14.175998
## iter 50 value 10.740408
## iter 60 value 10.615139
## iter 70 value 10.138088
## iter 80 value 8.776170
## iter 90 value 5.449129
## iter 100 value 5.399181
## final value 5.399181
## stopped after 100 iterations
## # weights: 39
## initial value 107.824745
## iter 10 value 79.775394
## iter 20 value 74.062589
## iter 30 value 69.690795
## iter 40 value 68.069902
## iter 50 value 64.858011
## iter
       60 value 62.511352
## iter 70 value 61.000444
## iter 80 value 60.220219
## iter 90 value 58.613317
## iter 100 value 58.053867
## final value 58.053867
## stopped after 100 iterations
```

```
## # weights: 109
## initial value 114.409881
## iter 10 value 57.420147
## iter 20 value 36.971835
## iter 30 value 22.084784
## iter 40 value 17.102821
## iter 50 value 16.310995
## iter 60 value 15.984661
## iter 70 value 15.940944
## iter 80 value 15.936295
## final value 15.936265
## converged
## # weights: 179
## initial value 112.395813
## iter 10 value 52.080574
## iter 20 value 22.718329
## iter 30 value 15.330558
## iter 40 value 13.498730
## iter 50 value 13.191362
## iter 60 value 12.890545
## iter 70 value 12.800024
## iter 80 value 12.744077
## iter 90 value 12.567801
## iter 100 value 12.551310
## final value 12.551310
## stopped after 100 iterations
## # weights: 39
## initial value 101.427263
## iter 10 value 63.673330
## iter 20 value 60.020893
## iter 30 value 59.354847
## iter 40 value 56.511763
## iter 50 value 56.441490
## iter 60 value 56.399558
## iter 70 value 56.360969
## iter 80 value 56.359130
## iter 90 value 56.350939
## iter 100 value 56.127369
## final value 56.127369
## stopped after 100 iterations
## # weights: 109
## initial value 101.887861
## iter 10 value 26.488203
## iter 20 value 6.914291
## iter 30 value 4.085094
## iter 40 value 3.625470
## iter 50 value 1.756450
## iter 60 value 0.695353
## iter 70 value 0.277666
## iter 80 value 0.271941
## iter 90 value 0.268344
## iter 100 value 0.263530
## final value 0.263530
## stopped after 100 iterations
```

```
## # weights: 179
## initial value 95.704309
## iter 10 value 23.349842
## iter 20 value 10.814138
## iter 30 value 8.715380
## iter 40 value 8.624406
## iter 50 value 8.533033
## iter 60 value 8.464953
## iter 70 value 8.437404
## iter 80 value 8.044356
## iter 90 value 6.388405
## iter 100 value 6.197112
## final value 6.197112
## stopped after 100 iterations
## # weights: 39
## initial value 104.296667
## iter 10 value 77.025851
## iter 20 value 62.629354
## iter 30 value 59.263841
## iter 40 value 57.642381
## iter 50 value 57.372222
## iter 60 value 57.333348
## iter 70 value 57.323559
## iter 80 value 57.321180
## iter 90 value 57.321057
## iter 90 value 57.321057
## iter 90 value 57.321057
## final value 57.321057
## converged
## # weights: 109
## initial value 107.296511
## iter 10 value 25.362223
## iter 20 value 14.993591
## iter 30 value 13.715812
## iter 40 value 12.261899
## iter 50 value 11.039153
## iter 60 value 10.227968
## iter 70 value 8.794011
## iter 80 value 7.557490
## iter 90 value 7.485686
## iter 100 value 7.377319
## final value 7.377319
## stopped after 100 iterations
## # weights: 179
## initial value 125.510522
## iter 10 value 45.391295
## iter 20 value 21.017046
## iter 30 value 19.516528
## iter 40 value 15.217016
## iter 50 value 12.013463
## iter 60 value 11.718623
## iter 70 value 11.502177
## iter 80 value 11.417523
## iter 90 value 11.254036
```

```
## iter 100 value 11.158928
## final value 11.158928
## stopped after 100 iterations
## # weights: 39
## initial value 99.286956
## iter 10 value 67.852828
## iter 20 value 62.961531
## iter 30 value 62.703163
## iter 40 value 62.682730
## iter 50 value 62.682466
## iter 50 value 62.682465
## iter 50 value 62.682465
## final value 62.682465
## converged
## # weights: 109
## initial value 107.072646
## iter 10 value 50.341556
## iter 20 value 31.801284
## iter 30 value 23.948510
## iter 40 value 18.150609
## iter 50 value 16.220409
## iter 60 value 15.753133
## iter 70 value 15.729246
## iter 80 value 15.728865
## final value 15.728861
## converged
## # weights: 179
## initial value 111.970236
## iter 10 value 34.372965
## iter 20 value 18.064936
## iter 30 value 13.361749
## iter 40 value 12.541324
## iter 50 value 12.154801
## iter 60 value 12.017273
## iter 70 value 12.001270
## iter 80 value 12.000691
## iter 90 value 12.000637
## final value 12.000636
## converged
## # weights: 39
## initial value 103.559374
## iter 10 value 70.010312
## iter 20 value 57.156874
## iter 30 value 53.927386
## iter 40 value 53.572091
## iter 50 value 53.278302
## iter 60 value 53.036635
## iter 70 value 52.943403
## iter 80 value 52.145052
## iter 90 value 52.130708
## iter 100 value 51.258997
## final value 51.258997
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 114.608400
## iter 10 value 48.672642
## iter 20 value 24.157843
## iter 30 value 15.504033
## iter 40 value 13.165696
## iter 50 value 11.207615
## iter 60 value 10.470547
## iter 70 value 10.403654
## iter 80 value 10.360412
## iter 90 value 10.240423
## iter 100 value 10.218823
## final value 10.218823
## stopped after 100 iterations
## # weights: 179
## initial value 105.433065
## iter 10 value 28.523213
## iter 20 value 0.728495
## iter 30 value 0.132762
## iter 40 value 0.115776
## iter 50 value 0.103897
## iter 60 value 0.097216
## iter 70 value 0.090305
## iter 80 value 0.083757
## iter 90 value 0.077559
## iter 100 value 0.073939
## final value 0.073939
## stopped after 100 iterations
## # weights: 39
## initial value 100.250715
## iter 10 value 59.500949
## iter 20 value 52.254535
## iter 30 value 48.504939
## iter 40 value 47.572323
## iter 50 value 46.667285
## iter 60 value 46.640032
## iter 70 value 46.616998
## iter 80 value 46.501457
## iter 90 value 46.288017
## iter 100 value 44.485807
## final value 44.485807
## stopped after 100 iterations
## # weights: 109
## initial value 116.473182
## iter 10 value 34.273330
## iter 20 value 24.480114
## iter 30 value 22.813100
## iter 40 value 21.449908
## iter 50 value 20.306050
## iter 60 value 18.465852
## iter 70 value 17.300470
## iter 80 value 13.137020
## iter 90 value 10.812836
## iter 100 value 10.416064
## final value 10.416064
```

```
## stopped after 100 iterations
## # weights: 179
## initial value 107.453799
## iter 10 value 16.614371
## iter 20 value 7.011421
## iter 30 value 4.143610
## iter 40 value 4.008715
## iter 50 value 3.960556
## iter 60 value 3.822503
## iter 70 value 3.591918
## iter 80 value 3.436005
## iter 90 value 3.191878
## iter 100 value 3.043064
## final value 3.043064
## stopped after 100 iterations
## # weights: 39
## initial value 103.320865
## iter 10 value 85.209593
## iter 20 value 66.694613
## iter 30 value 61.823827
## iter 40 value 60.877568
## iter 50 value 58.201353
## iter 60 value 56.594871
## iter 70 value 56.391773
## iter 80 value 56.389858
## iter 80 value 56.389858
## iter 80 value 56.389858
## final value 56.389858
## converged
## # weights: 109
## initial value 116.342391
## iter 10 value 56.166226
## iter 20 value 25.635515
## iter 30 value 17.279527
## iter 40 value 16.507618
## iter 50 value 16.292501
## iter 60 value 16.283223
## iter 70 value 16.283039
## final value 16.283037
## converged
## # weights: 179
## initial value 113.283764
## iter 10 value 32.569562
## iter 20 value 20.150417
## iter 30 value 14.426061
## iter 40 value 13.634122
## iter 50 value 13.336895
## iter
       60 value 13.030421
## iter 70 value 12.758627
## iter 80 value 12.650265
## iter 90 value 12.607596
## iter 100 value 12.469816
## final value 12.469816
## stopped after 100 iterations
```

```
## # weights: 39
## initial value 105.736386
## iter 10 value 63.246322
## iter 20 value 52.470727
## iter 30 value 47.456754
## iter 40 value 46.813960
## iter 50 value 46.603013
## iter 60 value 45.795324
## iter 70 value 45.737515
## iter 80 value 45.709524
## iter 90 value 45.703775
## iter 100 value 45.693047
## final value 45.693047
## stopped after 100 iterations
## # weights: 109
## initial value 93.959235
## iter 10 value 36.519949
## iter 20 value 31.426332
## iter 30 value 30.366494
## iter 40 value 28.880434
## iter 50 value 24.959311
## iter 60 value 17.032263
## iter 70 value 16.729534
## iter 80 value 16.718429
## iter 90 value 16.709041
## iter 100 value 16.481585
## final value 16.481585
## stopped after 100 iterations
## # weights: 179
## initial value 101.788902
## iter 10 value 23.731147
## iter 20 value 10.237795
## iter 30 value 7.849354
## iter 40 value 4.345500
## iter 50 value 2.331597
## iter 60 value 2.168719
## iter 70 value 0.696743
## iter 80 value 0.290112
## iter 90 value 0.260853
## iter 100 value 0.244932
## final value 0.244932
## stopped after 100 iterations
## # weights: 39
## initial value 104.458268
## iter 10 value 70.222783
## iter 20 value 64.231165
## iter 30 value 62.155540
## iter
       40 value 60.875710
## iter 50 value 60.857825
## iter 60 value 60.817928
## iter 70 value 60.809706
## iter 80 value 60.806563
## iter 90 value 60.805694
## final value 60.805681
```

```
## converged
## # weights: 109
## initial value 103.642164
## iter 10 value 24.228438
## iter 20 value 9.452049
## iter 30 value 3.720791
## iter 40 value 3.301987
## iter 50 value 3.295931
## iter 60 value 3.295837
## final value 3.295837
## converged
## # weights: 179
## initial value 105.704381
## iter 10 value 46.549555
## iter 20 value 20.491579
## iter 30 value 12.353323
## iter 40 value 8.130833
## iter 50 value 7.473297
## iter 60 value 7.131419
## iter 70 value 5.787543
## iter 80 value 4.733511
## iter 90 value 4.254214
## iter 100 value 4.203936
## final value 4.203936
## stopped after 100 iterations
## # weights: 39
## initial value 106.121089
## iter 10 value 74.692239
## iter 20 value 65.606413
## iter 30 value 64.297653
## iter 40 value 61.825284
## iter 50 value 60.633845
## iter 60 value 60.423170
## iter 70 value 60.214923
## iter 80 value 60.105598
## iter 90 value 60.104647
## iter 90 value 60.104646
## iter 90 value 60.104646
## final value 60.104646
## converged
## # weights: 109
## initial value 109.253145
## iter 10 value 57.025161
## iter 20 value 28.112607
## iter 30 value 21.195361
## iter 40 value 18.930384
## iter 50 value 17.568698
## iter 60 value 16.655916
## iter 70 value 15.690664
## iter 80 value 15.524600
## iter 90 value 15.521253
## iter 100 value 15.521152
## final value 15.521152
## stopped after 100 iterations
```

```
## # weights: 179
## initial value 138.039064
## iter 10 value 46.598751
## iter 20 value 24.260296
## iter 30 value 18.303595
## iter 40 value 14.067750
## iter 50 value 12.918251
## iter 60 value 12.589741
## iter 70 value 12.541442
## iter 80 value 12.538303
## iter 90 value 12.538141
## final value 12.538140
## converged
## # weights: 39
## initial value 106.356336
## iter 10 value 68.452015
## iter 20 value 65.416227
## iter 30 value 63.947799
## iter 40 value 63.916571
## iter 50 value 61.834987
## iter 60 value 61.669176
## iter 70 value 61.377194
## iter 80 value 61.286656
## iter 90 value 59.476084
## iter 100 value 58.962861
## final value 58.962861
## stopped after 100 iterations
## # weights: 109
## initial value 100.263011
## iter 10 value 66.357584
## iter 20 value 59.432815
## iter 30 value 50.924436
## iter 40 value 46.109702
## iter 50 value 44.394994
## iter 60 value 41.968917
## iter 70 value 40.515400
## iter 80 value 40.426876
## iter 90 value 40.005181
## iter 100 value 25.214026
## final value 25.214026
## stopped after 100 iterations
## # weights: 179
## initial value 109.625086
## iter 10 value 25.068673
## iter 20 value 13.760215
## iter 30 value 10.734845
## iter 40 value 10.147620
## iter 50 value 10.012784
## iter 60 value 7.201230
## iter 70 value 6.831791
## iter 80 value 6.823528
## iter 90 value 6.809467
## iter 100 value 6.797800
## final value 6.797800
```

```
## stopped after 100 iterations
## # weights: 39
## initial value 112.072423
## iter 10 value 65.034174
## iter 20 value 51.979473
## iter 30 value 45.653229
## iter 40 value 44.830322
## iter 50 value 43.150441
## iter 60 value 42.872868
## iter 70 value 42.767344
## iter 80 value 40.026993
## iter 90 value 38.340814
## iter 100 value 36.697940
## final value 36.697940
## stopped after 100 iterations
## # weights: 109
## initial value 112.390586
## iter 10 value 39.394985
## iter 20 value 13.491987
## iter 30 value 10.534286
## iter 40 value 10.441899
## iter 50 value 10.383337
## iter 60 value 10.338878
## iter 70 value 10.329045
## iter 80 value 10.317242
## iter 90 value 10.186622
## iter 100 value 10.000181
## final value 10.000181
## stopped after 100 iterations
## # weights: 179
## initial value 124.045733
## iter 10 value 29.267607
## iter 20 value 13.154668
## iter 30 value 6.242953
## iter 40 value 3.889525
## iter 50 value 3.797358
## iter 60 value 3.706932
## iter 70 value 0.021693
## iter 80 value 0.005182
## iter 90 value 0.002696
## iter 100 value 0.000481
## final value 0.000481
## stopped after 100 iterations
## # weights: 39
## initial value 106.405441
## iter 10 value 63.861865
## iter 20 value 58.574330
## iter 30 value 57.998185
## iter 40 value 57.624896
## iter 50 value 57.561700
## iter 60 value 57.551188
## iter 70 value 57.343096
## iter 80 value 57.263553
## final value 57.262028
```

```
## converged
## # weights: 109
## initial value 108.289289
## iter 10 value 60.864303
## iter 20 value 34.687795
## iter 30 value 18.546353
## iter 40 value 16.316635
## iter 50 value 16.182487
## iter 60 value 16.128697
## iter 70 value 16.127198
## iter 80 value 16.127105
## final value 16.127102
## converged
## # weights: 179
## initial value 115.851488
## iter 10 value 57.360759
## iter 20 value 22.508297
## iter 30 value 15.326321
## iter 40 value 12.045208
## iter 50 value 11.703389
## iter 60 value 11.673999
## iter 70 value 11.673336
## final value 11.673335
## converged
## # weights: 39
## initial value 108.643417
## iter 10 value 68.845729
## iter 20 value 55.584263
## iter 30 value 49.940049
## iter 40 value 49.388662
## iter 50 value 48.814768
## iter 60 value 48.567797
## iter 70 value 48.290483
## iter 80 value 48.277640
## iter 90 value 48.274796
## iter 100 value 48.272581
## final value 48.272581
## stopped after 100 iterations
## # weights: 109
## initial value 111.395285
## iter 10 value 36.710399
## iter 20 value 22.028561
## iter 30 value 13.310092
## iter 40 value 9.271770
## iter 50 value 6.308208
## iter 60 value 5.562658
## iter 70 value 4.058193
## iter 80 value 3.634185
## iter 90 value 2.946320
## iter 100 value 2.494842
## final value 2.494842
## stopped after 100 iterations
## # weights: 179
## initial value 110.287686
```

```
## iter 10 value 23.977702
## iter 20 value 16.030259
## iter 30 value 10.950104
## iter 40 value 10.844199
## iter 50 value 9.432056
## iter 60 value 6.767434
## iter 70 value 4.632996
## iter 80 value 3.787226
## iter 90 value 0.381152
## iter 100 value 0.298794
## final value 0.298794
## stopped after 100 iterations
## # weights: 39
## initial value 104.104121
## iter 10 value 94.632851
## iter 20 value 82.104303
## iter 30 value 71.771913
## iter 40 value 68.250404
## iter 50 value 64.508450
## iter 60 value 58.117662
## iter 70 value 56.510342
## iter 80 value 53.399022
## iter 90 value 52.584196
## iter 100 value 52.136980
## final value 52.136980
## stopped after 100 iterations
## # weights: 109
## initial value 102.660668
## iter 10 value 32.782329
## iter 20 value 29.072581
## iter 30 value 24.595833
## iter 40 value 20.424164
## iter 50 value 15.258176
## iter 60 value 14.370692
## iter 70 value 14.328550
## iter 80 value 14.294252
## iter 90 value 9.913291
## iter 100 value 9.026074
## final value 9.026074
## stopped after 100 iterations
## # weights: 179
## initial value 100.913089
## iter 10 value 23.433515
## iter 20 value 2.742041
## iter 30 value 0.098064
## iter 40 value 0.005778
## iter 50 value 0.001027
## final value 0.000055
## converged
## # weights:
## initial value 119.429726
## iter 10 value 77.705077
## iter 20 value 67.193284
## iter 30 value 62.468116
```

```
## iter 40 value 58.138880
## iter 50 value 57.973718
## final value 57.973633
## converged
## # weights: 109
## initial value 101.372172
## iter 10 value 48.227120
## iter 20 value 33.567402
## iter 30 value 26.726559
## iter 40 value 20.620389
## iter 50 value 17.158583
## iter 60 value 16.425468
## iter 70 value 16.318073
## iter 80 value 16.311307
## final value 16.311173
## converged
## # weights: 179
## initial value 108.628410
## iter 10 value 62.030342
## iter 20 value 25.043421
## iter 30 value 13.672449
## iter 40 value 12.820465
## iter 50 value 12.536611
## iter 60 value 12.191646
## iter 70 value 12.142232
## iter 80 value 12.138215
## iter 90 value 12.137719
## iter 100 value 12.137587
## final value 12.137587
## stopped after 100 iterations
## # weights: 39
## initial value 116.154045
## iter 10 value 72.039430
## iter 20 value 55.630344
## iter 30 value 51.128362
## iter 40 value 49.381377
## iter 50 value 47.825646
## iter 60 value 47.400660
## iter 70 value 47.133355
## iter 80 value 46.504331
## iter 90 value 45.181297
## iter 100 value 43.575820
## final value 43.575820
## stopped after 100 iterations
## # weights: 109
## initial value 101.979893
## iter 10 value 50.640612
## iter 20 value 15.700287
## iter 30 value 9.068393
## iter 40 value 8.678117
## iter 50 value 8.561636
## iter 60 value 8.532056
## iter 70 value 8.292066
## iter 80 value 7.796292
```

```
## iter 90 value 7.600462
## iter 100 value 7.237641
## final value 7.237641
## stopped after 100 iterations
## # weights: 179
## initial value 116.458597
## iter 10 value 38.761667
## iter 20 value 18.181590
## iter 30 value 7.199160
## iter 40 value 4.105810
## iter 50 value 3.607261
## iter 60 value 3.573308
## iter 70 value 3.548072
## iter 80 value 2.955874
## iter 90 value 2.945030
## iter 100 value 2.925716
## final value 2.925716
## stopped after 100 iterations
## # weights: 39
## initial value 102.902944
## iter 10 value 95.117534
## iter 20 value 77.992387
## iter 30 value 59.882523
## iter 40 value 51.135631
## iter 50 value 50.046115
## iter 60 value 50.042183
## iter 70 value 50.041889
## iter 70 value 50.041889
## iter 70 value 50.041889
## final value 50.041889
## converged
## # weights: 109
## initial value 115.267663
## iter 10 value 58.436888
## iter 20 value 47.112767
## iter 30 value 41.952219
## iter 40 value 39.953380
## iter 50 value 38.487833
## iter 60 value 37.859163
## iter 70 value 36.411762
## iter 80 value 33.682640
## iter 90 value 33.407897
## iter 100 value 32.331523
## final value 32.331523
## stopped after 100 iterations
## # weights: 179
## initial value 100.684060
## iter 10 value 20.204483
## iter 20 value 10.006904
## iter 30 value 6.390664
## iter 40 value 5.958816
## iter 50 value 4.871629
## iter 60 value 4.310071
## iter 70 value 3.993092
```

```
## iter 80 value 3.893582
## iter 90 value 3.134830
## iter 100 value 2.911010
## final value 2.911010
## stopped after 100 iterations
## # weights: 39
## initial value 112.928870
## iter 10 value 71.810066
## iter 20 value 60.569788
## iter 30 value 58.130220
## iter 40 value 57.674019
## iter 50 value 56.638793
## iter 60 value 55.707852
## iter 70 value 55.699390
## iter 80 value 55.682629
## iter 90 value 55.531367
## iter 100 value 55.501614
## final value 55.501614
## stopped after 100 iterations
## # weights: 109
## initial value 132.454427
## iter 10 value 75.339895
## iter 20 value 51.015371
## iter 30 value 42.506734
## iter 40 value 31.643265
## iter 50 value 20.287321
## iter 60 value 17.622962
## iter 70 value 17.331452
## iter 80 value 17.257635
## iter 90 value 17.194176
## iter 100 value 17.139717
## final value 17.139717
## stopped after 100 iterations
## # weights: 179
## initial value 109.431568
## iter 10 value 28.271596
## iter 20 value 14.086766
## iter 30 value 12.745537
## iter 40 value 12.577017
## iter 50 value 12.399960
## iter 60 value 12.338756
## iter 70 value 12.333465
## final value 12.333432
## converged
## # weights: 39
## initial value 100.913687
## iter 10 value 68.321926
## iter 20 value 62.361350
## iter 30 value 58.081019
## iter 40 value 57.146854
## iter 50 value 56.980564
## iter 60 value 55.674254
## iter 70 value 54.617339
## iter 80 value 54.531221
```

```
## iter 90 value 54.403796
## iter 100 value 54.392980
## final value 54.392980
## stopped after 100 iterations
## # weights: 109
## initial value 99.323085
## iter 10 value 43.400727
## iter 20 value 29.610046
## iter 30 value 28.215557
## iter 40 value 25.187744
## iter 50 value 19.465281
## iter 60 value 16.566440
## iter 70 value 15.217661
## iter 80 value 13.945526
## iter 90 value 13.729266
## iter 100 value 12.485691
## final value 12.485691
## stopped after 100 iterations
## # weights: 179
## initial value 121.662153
## iter 10 value 33.659989
## iter 20 value 5.332921
## iter 30 value 0.551199
## iter 40 value 0.139551
## iter 50 value 0.130606
## iter 60 value 0.125440
## iter 70 value 0.115584
## iter 80 value 0.109488
## iter 90 value 0.103554
## iter 100 value 0.100060
## final value 0.100060
## stopped after 100 iterations
## # weights: 39
## initial value 103.035516
## iter 10 value 67.202549
## iter 20 value 58.454661
## iter 30 value 57.810290
## iter 40 value 56.269647
## iter 50 value 53.340897
## iter 60 value 52.388565
## iter 70 value 49.889024
## iter 80 value 49.094247
## iter 90 value 43.482376
## iter 100 value 42.437101
## final value 42.437101
## stopped after 100 iterations
## # weights: 109
## initial value 98.606029
## iter 10 value 43.828124
## iter 20 value 25.446274
## iter 30 value 23.327318
## iter 40 value 22.973550
## iter 50 value 22.861771
## iter 60 value 22.585122
```

```
## iter 70 value 22.361625
## iter 80 value 22.339547
## iter 90 value 22.329942
## iter 100 value 22.313089
## final value 22.313089
## stopped after 100 iterations
## # weights: 179
## initial value 116.721945
## iter 10 value 21.158818
## iter 20 value 11.175851
## iter 30 value 10.507747
## iter 40 value 10.179368
## iter 50 value 10.032285
## iter 60 value 6.687856
## iter 70 value 5.955981
## iter 80 value 5.818288
## iter 90 value 5.228004
## iter 100 value 5.216850
## final value 5.216850
## stopped after 100 iterations
## # weights: 39
## initial value 102.506497
## iter 10 value 81.787259
## iter 20 value 64.685533
## iter 30 value 60.413218
## iter 40 value 58.380602
## iter 50 value 58.173716
## iter 60 value 58.165468
## iter 70 value 57.826261
## iter 80 value 57.746568
## iter 90 value 57.744660
## final value 57.744659
## converged
## # weights: 109
## initial value 126.906785
## iter 10 value 75.492805
## iter 20 value 46.356830
## iter 30 value 23.509959
## iter 40 value 18.760807
## iter 50 value 17.803642
## iter 60 value 17.494282
## iter 70 value 17.189129
## iter 80 value 17.119594
## iter 90 value 17.119079
## final value 17.119077
## converged
## # weights: 179
## initial value 125.422441
## iter 10 value 49.803856
## iter 20 value 30.976175
## iter 30 value 19.400579
## iter 40 value 15.109142
## iter 50 value 12.935283
## iter 60 value 12.372935
```

```
## iter 70 value 12.210302
## iter 80 value 12.150268
## iter 90 value 12.138439
## iter 100 value 12.138117
## final value 12.138117
## stopped after 100 iterations
## # weights: 39
## initial value 105.568778
## iter 10 value 65.287381
## iter 20 value 61.661921
## iter 30 value 59.188873
## iter 40 value 58.762466
## iter 50 value 57.473034
## iter 60 value 56.486467
## iter 70 value 56.288190
## iter 80 value 56.213802
## iter 90 value 54.381165
## iter 100 value 52.789144
## final value 52.789144
## stopped after 100 iterations
## # weights: 109
## initial value 111.525584
## iter 10 value 46.274228
## iter 20 value 25.018974
## iter 30 value 22.193350
## iter 40 value 18.413932
## iter 50 value 18.067061
## iter 60 value 18.047277
## iter 70 value 18.036309
## iter 80 value 15.720196
## iter 90 value 14.508135
## iter 100 value 14.415528
## final value 14.415528
## stopped after 100 iterations
## # weights: 179
## initial value 142.483354
## iter 10 value 80.870178
## iter 20 value 70.300416
## iter 30 value 67.024735
## iter 40 value 54.827190
## iter 50 value 40.970285
## iter 60 value 40.772169
## iter 70 value 40.761883
## iter 80 value 37.455967
## iter 90 value 37.043444
## iter 100 value 35.975129
## final value 35.975129
## stopped after 100 iterations
## # weights: 39
## initial value 106.375245
## iter 10 value 61.477491
## iter 20 value 51.579410
## iter 30 value 46.683733
## iter 40 value 46.577219
```

```
## iter 50 value 46.576626
## final value 46.576594
## converged
## # weights: 109
## initial value 109.682889
## iter 10 value 53.976210
## iter 20 value 25.163669
## iter 30 value 23.454847
## iter 40 value 22.771648
## iter 50 value 22.558459
## iter 60 value 21.639478
## iter 70 value 21.590591
## iter 80 value 21.587522
## iter 90 value 21.357415
## iter 100 value 16.525865
## final value 16.525865
## stopped after 100 iterations
## # weights: 179
## initial value 126.896091
## iter 10 value 26.618382
## iter 20 value 17.197097
## iter 30 value 14.435782
## iter 40 value 12.924168
## iter 50 value 12.652586
## iter 60 value 9.981097
## iter 70 value 8.878792
## iter 80 value 6.371991
## iter 90 value 1.452116
## iter 100 value 0.264262
## final value 0.264262
## stopped after 100 iterations
## # weights: 39
## initial value 135.096055
## iter 10 value 76.609294
## iter 20 value 65.883392
## iter 30 value 59.641980
## iter 40 value 58.993143
## iter 50 value 58.771664
## iter 60 value 58.393686
## iter 70 value 57.922598
## iter 80 value 57.906495
## final value 57.906494
## converged
## # weights: 109
## initial value 117.493313
## iter 10 value 66.028895
## iter 20 value 34.052163
## iter 30 value 21.372445
## iter 40 value 18.469352
## iter 50 value 17.637266
## iter 60 value 17.389107
## iter 70 value 17.362444
## iter 80 value 17.352768
## iter 90 value 17.352390
```

```
## iter 100 value 17.352227
## final value 17.352227
## stopped after 100 iterations
## # weights: 179
## initial value 141.970467
## iter 10 value 42.564692
## iter 20 value 23.248382
## iter 30 value 15.557352
## iter 40 value 13.332413
## iter 50 value 12.920563
## iter 60 value 12.735156
## iter 70 value 12.526042
## iter 80 value 12.494029
## iter 90 value 12.482295
## iter 100 value 12.480921
## final value 12.480921
## stopped after 100 iterations
## # weights: 39
## initial value 104.043870
## iter 10 value 85.293057
## iter 20 value 75.115480
## iter 30 value 69.764047
## iter 40 value 67.298602
## iter 50 value 66.038954
## iter 60 value 61.247612
## iter 70 value 58.185033
## iter 80 value 51.620265
## iter 90 value 48.196567
## iter 100 value 46.004115
## final value 46.004115
## stopped after 100 iterations
## # weights: 109
## initial value 107.342746
## iter 10 value 41.088781
## iter 20 value 19.871015
## iter 30 value 19.145624
## iter 40 value 19.110781
## iter 50 value 19.083913
## iter 60 value 19.058351
## iter 70 value 19.019856
## iter 80 value 18.981652
## iter 90 value 17.728183
## iter 100 value 16.666359
## final value 16.666359
## stopped after 100 iterations
## # weights: 179
## initial value 99.458461
## iter 10 value 22.889092
## iter 20 value 3.951857
## iter 30 value 0.800651
## iter 40 value 0.211433
## iter 50 value 0.184478
## iter 60 value 0.178162
## iter 70 value 0.162257
```

```
## iter 80 value 0.148402
## iter 90 value 0.139573
## iter 100 value 0.128653
## final value 0.128653
## stopped after 100 iterations
## # weights: 39
## initial value 116.998611
## iter 10 value 66.983932
## iter 20 value 61.799025
## iter 30 value 61.003114
## iter 40 value 59.892893
## iter 50 value 57.741088
## iter 60 value 57.686713
## iter 70 value 57.570713
## iter 80 value 57.433687
## iter 90 value 53.139768
## iter 100 value 51.618662
## final value 51.618662
## stopped after 100 iterations
## # weights: 109
## initial value 117.964956
## iter 10 value 34.259599
## iter 20 value 16.807223
## iter 30 value 9.643818
## iter 40 value 8.987129
## iter 50 value 8.308468
## iter 60 value 8.193778
## iter 70 value 8.055228
## iter 80 value 8.027582
## iter 90 value 8.018946
## iter 100 value 8.014011
## final value 8.014011
## stopped after 100 iterations
## # weights: 179
## initial value 119.662556
## iter 10 value 17.294015
## iter 20 value 2.497720
## iter 30 value 1.914160
## iter 40 value 1.909786
## iter 50 value 1.909551
## final value 1.909548
## converged
## # weights:
              39
## initial value 105.830669
## iter 10 value 69.251233
## iter 20 value 65.584326
## iter 30 value 62.058325
## iter
       40 value 58.951676
## iter 50 value 58.196204
## iter 60 value 57.888591
## iter 70 value 57.696740
## iter 80 value 57.659527
## final value 57.659311
## converged
```

```
## # weights: 109
## initial value 96.087951
## iter 10 value 36.653478
## iter 20 value 25.450749
## iter 30 value 20.252368
## iter 40 value 19.021345
## iter 50 value 18.586706
## iter 60 value 18.134710
## iter 70 value 17.807973
## iter 80 value 17.553251
## iter 90 value 17.194438
## iter 100 value 16.794678
## final value 16.794678
## stopped after 100 iterations
## # weights: 179
## initial value 128.862280
## iter 10 value 51.729388
## iter 20 value 29.209093
## iter 30 value 18.841753
## iter 40 value 14.122869
## iter 50 value 13.008250
## iter 60 value 12.489594
## iter 70 value 12.336610
## iter 80 value 12.333355
## iter 90 value 12.333115
## final value 12.333106
## converged
## # weights: 39
## initial value 104.183800
## iter 10 value 63.136261
## iter 20 value 57.728559
## iter 30 value 55.069733
## iter 40 value 53.686989
## iter 50 value 53.514579
## iter 60 value 53.438881
## iter 70 value 52.214352
## iter 80 value 51.869493
## iter 90 value 51.749341
## iter 100 value 51.620893
## final value 51.620893
## stopped after 100 iterations
## # weights: 109
## initial value 110.258659
## iter 10 value 37.486723
## iter 20 value 15.789122
## iter 30 value 9.000416
## iter 40 value 5.245511
## iter 50 value 3.157589
## iter 60 value 3.104749
## iter 70 value 3.094253
## iter 80 value 3.074371
## iter 90 value 3.033486
## iter 100 value 0.893676
## final value 0.893676
```

```
## stopped after 100 iterations
## # weights: 179
## initial value 118.203115
## iter 10 value 24.695646
## iter 20 value 15.585488
## iter 30 value 12.046722
## iter 40 value 10.907912
## iter 50 value 9.025767
## iter 60 value 8.921963
## iter 70 value 8.239355
## iter 80 value 7.672052
## iter 90 value 7.550155
## iter 100 value 7.456083
## final value 7.456083
## stopped after 100 iterations
## # weights: 39
## initial value 105.082330
## iter 10 value 89.250841
## iter 20 value 55.053026
## iter 30 value 54.537456
## iter 40 value 54.536669
## final value 54.536667
## converged
## # weights: 109
## initial value 103.218517
## iter 10 value 25.488058
## iter 20 value 12.656920
## iter 30 value 11.459115
## iter 40 value 4.611962
## iter 50 value 4.515429
## iter 60 value 4.038720
## iter 70 value 3.852779
## iter 80 value 3.836705
## iter 90 value 3.756311
## iter 100 value 3.384470
## final value 3.384470
## stopped after 100 iterations
## # weights: 179
## initial value 103.608427
## iter 10 value 26.686248
## iter 20 value 5.952020
## iter 30 value 2.505683
## iter 40 value 1.452198
## iter 50 value 1.420127
## iter 60 value 1.409060
        70 value 1.397836
## iter
## iter 80 value 1.392284
## iter 90 value 1.388814
## iter 100 value 1.387729
## final value 1.387729
## stopped after 100 iterations
## # weights: 39
## initial value 105.100112
## iter 10 value 79.815206
```

```
## iter 20 value 76.877191
## iter 30 value 72.711150
## iter 40 value 68.563403
## iter 50 value 65.585533
## iter 60 value 65.327313
## iter 70 value 64.694537
## iter 80 value 64.670858
## final value 64.670794
## converged
## # weights: 109
## initial value 107.917549
## iter 10 value 42.563395
## iter 20 value 24.146781
## iter 30 value 17.251261
## iter 40 value 16.814284
## iter 50 value 16.779971
## iter 60 value 16.778538
## iter 70 value 16.774396
## iter 80 value 16.624766
## iter 90 value 16.470744
## iter 100 value 16.441429
## final value 16.441429
## stopped after 100 iterations
## # weights: 179
## initial value 117.939171
## iter 10 value 29.243039
## iter 20 value 15.473215
## iter 30 value 13.352784
## iter 40 value 12.871967
## iter 50 value 12.801351
## iter 60 value 12.779367
## iter 70 value 12.734912
## iter 80 value 12.733841
## final value 12.733839
## converged
## # weights: 39
## initial value 103.095073
## iter 10 value 64.405363
## iter 20 value 62.998348
## iter 30 value 62.165910
## iter 40 value 61.457724
## iter 50 value 53.992488
## iter 60 value 49.790345
## iter 70 value 48.978211
## iter 80 value 48.621253
## iter 90 value 47.788137
## iter 100 value 46.255589
## final value 46.255589
## stopped after 100 iterations
## # weights: 109
## initial value 107.104233
## iter 10 value 45.795587
## iter 20 value 27.037214
## iter 30 value 25.324404
```

```
## iter 40 value 18.295890
## iter 50 value 17.929887
## iter 60 value 15.635469
## iter 70 value 15.613910
## iter 80 value 15.480294
## iter 90 value 15.100108
## iter 100 value 15.089462
## final value 15.089462
## stopped after 100 iterations
## # weights: 179
## initial value 118.595524
## iter 10 value 9.228820
## iter 20 value 2.031334
## iter 30 value 1.542707
## iter 40 value 1.520213
## iter 50 value 1.512758
## iter 60 value 1.505597
## iter 70 value 1.497186
## iter 80 value 1.493245
## iter 90 value 1.473857
## iter 100 value 0.155949
## final value 0.155949
## stopped after 100 iterations
## # weights: 39
## initial value 102.403787
## iter 10 value 71.052825
## iter 20 value 59.423164
## iter 30 value 59.009155
## iter 40 value 57.553719
## iter 50 value 57.306851
## iter 60 value 56.834353
## iter 70 value 56.046658
## iter 80 value 55.694162
## iter 90 value 55.003883
## iter 100 value 51.635742
## final value 51.635742
## stopped after 100 iterations
## # weights: 109
## initial value 113.248935
## iter 10 value 30.123739
## iter 20 value 19.919473
## iter 30 value 14.442042
## iter 40 value 13.044798
## iter 50 value 12.681620
## iter 60 value 12.310089
## iter 70 value 11.836379
## iter 80 value 11.452948
## iter 90 value 11.383149
## iter 100 value 11.191757
## final value 11.191757
## stopped after 100 iterations
## # weights: 179
## initial value 110.001748
## iter 10 value 52.973058
```

```
## iter 20 value 24.542659
## iter 30 value 20.327671
## iter 40 value 15.764696
## iter 50 value 12.767575
## iter 60 value 11.057664
## iter 70 value 10.765844
## iter 80 value 10.345875
## iter 90 value 9.781062
## iter 100 value 9.661580
## final value 9.661580
## stopped after 100 iterations
## # weights:
              39
## initial value 107.747973
## iter 10 value 79.727757
## iter 20 value 66.926899
## iter 30 value 61.571105
## iter 40 value 59.733704
## iter 50 value 58.496842
## iter 60 value 57.872977
## iter 70 value 57.590424
## iter 80 value 57.414424
## iter 90 value 57.407494
## final value 57.407487
## converged
## # weights: 109
## initial value 115.641531
## iter 10 value 50.739236
## iter 20 value 28.280328
## iter 30 value 20.508272
## iter 40 value 18.338220
## iter 50 value 16.100876
## iter 60 value 15.716135
## iter
       70 value 15.687606
## iter 80 value 15.686209
## iter 90 value 15.686146
## final value 15.686145
## converged
## # weights: 179
## initial value 121.731627
## iter 10 value 54.072596
## iter 20 value 22.979033
## iter 30 value 14.531366
## iter 40 value 12.499999
## iter 50 value 12.058427
## iter 60 value 12.048304
        70 value 12.047228
## iter
## iter 80 value 12.030373
## iter 90 value 11.834911
## iter 100 value 11.637256
## final value 11.637256
## stopped after 100 iterations
## # weights: 39
## initial value 103.681866
## iter 10 value 59.304041
```

```
## iter 20 value 57.324948
## iter 30 value 56.343407
## iter 40 value 55.373041
## iter 50 value 55.365170
## iter 60 value 54.616544
## iter 70 value 53.684371
## iter 80 value 53.680127
## iter 90 value 53.201999
## iter 100 value 52.734436
## final value 52.734436
## stopped after 100 iterations
## # weights: 109
## initial value 106.838757
## iter 10 value 51.870309
## iter 20 value 47.948553
## iter 30 value 45.257253
## iter 40 value 27.247696
## iter 50 value 21.938159
## iter 60 value 21.333904
## iter 70 value 20.717468
## iter 80 value 20.325719
## iter 90 value 19.370170
## iter 100 value 19.209037
## final value 19.209037
## stopped after 100 iterations
## # weights: 179
## initial value 105.674945
## iter 10 value 22.027455
## iter 20 value 6.654449
## iter 30 value 5.701856
## iter 40 value 5.564341
## iter 50 value 5.509316
## iter 60 value 4.990982
## iter 70 value 4.018605
## iter 80 value 1.881191
## iter 90 value 1.765851
## iter 100 value 1.742856
## final value 1.742856
## stopped after 100 iterations
## # weights: 39
## initial value 112.869840
## iter 10 value 72.876987
## iter 20 value 61.547193
## iter 30 value 60.758307
## iter 40 value 59.498691
## iter 50 value 58.710423
## iter 60 value 57.084085
## iter 70 value 55.639031
## iter 80 value 54.606390
## iter 90 value 54.458178
## iter 100 value 54.055820
## final value 54.055820
## stopped after 100 iterations
## # weights: 109
```

```
## initial value 126.900732
## iter 10 value 57.223975
## iter 20 value 36.585155
## iter 30 value 21.758450
## iter 40 value 10.124769
## iter 50 value 9.390520
## iter 60 value 9.267269
## iter 70 value 9.147667
## iter 80 value 9.113864
## iter 90 value 9.095141
## iter 100 value 8.959246
## final value 8.959246
## stopped after 100 iterations
## # weights: 179
## initial value 104.283855
## iter 10 value 20.758754
## iter 20 value 11.884268
## iter 30 value 11.178848
## iter 40 value 10.693612
## iter 50 value 10.635022
## iter 60 value 10.495920
## iter 70 value 10.433809
## iter 80 value 10.222676
## iter 90 value 10.208508
## iter 100 value 10.093986
## final value 10.093986
## stopped after 100 iterations
## # weights:
              39
## initial value 106.704691
## iter 10 value 77.931973
## iter 20 value 65.162515
## iter 30 value 61.752405
## iter 40 value 59.647890
## iter 50 value 58.439908
## iter 60 value 58.134947
## iter 70 value 57.999753
## iter 80 value 57.484664
## iter 90 value 57.465099
## final value 57.465084
## converged
## # weights: 109
## initial value 101.281788
## iter 10 value 52.614419
## iter 20 value 28.933868
## iter 30 value 18.307748
## iter 40 value 17.591372
## iter 50 value 17.427833
## iter 60 value 17.423486
## iter 70 value 17.423402
## final value 17.423398
## converged
## # weights: 179
## initial value 112.107421
## iter 10 value 40.687887
```

```
## iter 20 value 19.017303
## iter 30 value 14.149128
## iter 40 value 12.888287
## iter 50 value 12.503109
## iter 60 value 12.410647
## iter 70 value 12.407067
## iter 80 value 12.406993
## final value 12.406989
## converged
## # weights: 39
## initial value 102.966885
## iter 10 value 63.608194
## iter 20 value 56.201441
## iter 30 value 52.428983
## iter 40 value 51.147780
## iter 50 value 51.134441
## iter 60 value 51.122064
## iter 70 value 51.076005
## iter 80 value 51.051720
## iter 90 value 51.042610
## iter 100 value 51.035130
## final value 51.035130
## stopped after 100 iterations
## # weights: 109
## initial value 113.610656
## iter 10 value 46.467633
## iter 20 value 35.301876
## iter 30 value 30.066421
## iter 40 value 29.601282
## iter 50 value 25.609261
## iter 60 value 17.983663
## iter 70 value 16.647743
## iter 80 value 11.502619
## iter 90 value 8.308751
## iter 100 value 7.566426
## final value 7.566426
## stopped after 100 iterations
## # weights: 179
## initial value 93.727114
## iter 10 value 17.982797
## iter 20 value 5.874062
## iter 30 value 4.320749
## iter 40 value 4.294416
## iter 50 value 4.274965
## iter 60 value 4.253659
## iter 70 value 3.822169
## iter 80 value 3.416225
## iter 90 value 3.400944
## iter 100 value 1.669066
## final value 1.669066
## stopped after 100 iterations
## # weights: 39
## initial value 111.071494
## iter 10 value 81.917978
```

```
## iter 20 value 66.752529
## iter 30 value 63.269597
## iter 40 value 58.737222
## iter 50 value 55.648921
## iter 60 value 54.526295
## iter 70 value 49.330782
## iter 80 value 48.587899
## iter 90 value 46.533615
## iter 100 value 46.066413
## final value 46.066413
## stopped after 100 iterations
## # weights: 109
## initial value 107.089943
## iter 10 value 55.234740
## iter 20 value 22.389060
## iter 30 value 20.367074
## iter 40 value 19.609503
## iter 50 value 18.494994
## iter 60 value 17.263878
## iter 70 value 17.038706
## iter 80 value 16.872584
## iter 90 value 16.527268
## iter 100 value 16.402375
## final value 16.402375
## stopped after 100 iterations
## # weights: 179
## initial value 103.652883
## iter 10 value 24.167727
## iter 20 value 9.953042
## iter 30 value 8.492294
## iter 40 value 7.065062
## iter 50 value 5.199502
## iter 60 value 5.189567
## iter 70 value 5.158476
## iter 80 value 5.141893
## iter 90 value 5.127208
## iter 100 value 5.127034
## final value 5.127034
## stopped after 100 iterations
## # weights: 39
## initial value 107.242423
## iter 10 value 69.668594
## iter 20 value 62.957453
## iter 30 value 60.293402
## iter 40 value 60.078461
## iter 50 value 59.938452
## iter 60 value 59.513597
## iter 70 value 59.165025
## iter 80 value 59.112118
## final value 59.112044
## converged
## # weights: 109
## initial value 115.412115
## iter 10 value 73.212100
```

```
## iter 20 value 49.088261
## iter 30 value 23.155727
## iter 40 value 18.003186
## iter 50 value 16.879399
## iter 60 value 16.798870
## iter 70 value 16.781864
## iter 80 value 16.742496
## iter 90 value 16.738341
## iter 100 value 16.738237
## final value 16.738237
## stopped after 100 iterations
## # weights: 179
## initial value 121.281352
## iter 10 value 35.370770
## iter 20 value 21.488152
## iter 30 value 16.066384
## iter 40 value 13.597310
## iter 50 value 12.605599
## iter 60 value 12.278221
## iter 70 value 12.251660
## iter 80 value 12.251127
## iter 90 value 12.251106
## final value 12.251106
## converged
## # weights: 39
## initial value 104.544298
## iter 10 value 64.526224
## iter 20 value 50.388044
## iter 30 value 40.914239
## iter 40 value 38.582250
## iter 50 value 37.377838
## iter 60 value 36.922364
## iter 70 value 36.221697
## iter 80 value 36.174226
## iter 90 value 36.061043
## iter 100 value 36.019991
## final value 36.019991
## stopped after 100 iterations
## # weights: 109
## initial value 111.341716
## iter 10 value 41.008508
## iter 20 value 19.711727
## iter 30 value 12.263611
## iter 40 value 6.314696
## iter 50 value 5.384759
## iter 60 value 5.212949
## iter 70 value 4.847233
## iter 80 value 4.745915
## iter 90 value 4.600531
## iter 100 value 4.182227
## final value 4.182227
## stopped after 100 iterations
## # weights: 179
## initial value 121.416140
```

```
## iter 10 value 17.285821
## iter 20 value 7.404485
## iter 30 value 5.593378
## iter 40 value 5.517382
## iter 50 value 5.495217
## iter 60 value 5.325117
## iter 70 value 5.272660
## iter 80 value 5.152960
## iter 90 value 5.139879
## iter 100 value 5.122328
## final value 5.122328
## stopped after 100 iterations
## # weights: 39
## initial value 101.176678
## iter 10 value 64.942117
## iter 20 value 60.189055
## iter 30 value 59.479518
## iter 40 value 59.208528
## iter 50 value 59.171348
## iter 60 value 59.166573
## iter 70 value 59.164137
## iter 80 value 59.163955
## iter 80 value 59.163954
## final value 59.163954
## converged
## # weights: 109
## initial value 104.413824
## iter 10 value 42.854235
## iter 20 value 19.182867
## iter 30 value 9.740929
## iter 40 value 6.772130
## iter 50 value 6.440887
## iter 60 value 6.272804
## iter 70 value 6.272079
## iter 80 value 5.934192
## iter 90 value 5.592700
## iter 100 value 5.407909
## final value 5.407909
## stopped after 100 iterations
## # weights: 179
## initial value 127.217693
## iter 10 value 22.643510
## iter 20 value 6.696243
## iter 30 value 2.929428
## iter 40 value 2.774617
## iter 50 value 2.772779
## iter 60 value 2.772590
## iter 60 value 2.772590
## iter 60 value 2.772590
## final value 2.772590
## converged
## # weights: 39
## initial value 103.130121
## iter 10 value 73.534905
```

```
## iter 20 value 65.080629
## iter 30 value 60.105258
## iter 40 value 59.068942
## iter 50 value 58.569938
## iter 60 value 58.541653
## final value 58.541651
## converged
## # weights: 109
## initial value 110.493544
## iter 10 value 55.135778
## iter 20 value 25.915758
## iter 30 value 19.037411
## iter 40 value 16.209721
## iter 50 value 15.725766
## iter 60 value 15.703111
## iter 70 value 15.701788
## iter 80 value 15.701748
## iter 80 value 15.701748
## iter 80 value 15.701748
## final value 15.701748
## converged
## # weights: 179
## initial value 106.481457
## iter 10 value 43.837386
## iter 20 value 25.044892
## iter 30 value 16.600008
## iter 40 value 12.913796
## iter 50 value 12.234388
## iter 60 value 12.095502
## iter 70 value 12.084205
## iter 80 value 12.083223
## iter 90 value 12.083188
## final value 12.083188
## converged
## # weights: 39
## initial value 111.716414
## iter 10 value 60.088700
## iter 20 value 45.960138
## iter 30 value 36.906685
## iter 40 value 36.697115
## iter 50 value 36.695805
## iter 60 value 36.690875
## iter 70 value 36.675566
## iter 80 value 36.485801
## iter 90 value 36.339357
## iter 100 value 36.130272
## final value 36.130272
## stopped after 100 iterations
## # weights: 109
## initial value 115.061469
## iter 10 value 42.919828
## iter 20 value 7.646608
## iter 30 value 1.946823
## iter 40 value 1.557959
```

```
## iter 50 value 1.545121
## iter 60 value 1.535179
## iter 70 value 1.529208
## iter 80 value 1.517635
## iter 90 value 1.505588
## iter 100 value 1.499133
## final value 1.499133
## stopped after 100 iterations
## # weights: 179
## initial value 110.006198
## iter 10 value 22.072325
## iter 20 value 5.538083
## iter 30 value 4.070544
## iter 40 value 4.054390
## iter 50 value 4.030999
## iter 60 value 4.013053
## iter 70 value 4.005433
## iter 80 value 4.000324
## iter 90 value 2.717009
## iter 100 value 1.505916
## final value 1.505916
## stopped after 100 iterations
## # weights: 39
## initial value 100.615696
## iter 10 value 70.908875
## iter 20 value 62.751346
## iter 30 value 62.623699
## iter 40 value 62.138530
## iter 50 value 61.813073
## iter 60 value 61.674940
## iter 70 value 61.658481
## iter 80 value 61.655601
## final value 61.655160
## converged
## # weights: 109
## initial value 98.295046
## iter 10 value 28.535016
## iter 20 value 16.738147
## iter 30 value 10.518791
## iter 40 value 9.740156
## iter 50 value 9.713299
## iter 60 value 9.712616
## iter 70 value 9.712360
## iter 80 value 9.712324
## iter 80 value 9.712324
## iter 80 value 9.712324
## final value 9.712324
## converged
## # weights: 179
## initial value 141.452737
## iter 10 value 53.685731
## iter 20 value 32.433635
## iter 30 value 10.730215
## iter 40 value 6.465200
```

```
## iter 50 value 6.420496
## iter 60 value 6.419668
## iter 60 value 6.419668
## iter 70 value 6.419661
## iter 70 value 6.419661
## iter 70 value 6.419661
## final value 6.419661
## converged
## # weights: 39
## initial value 120.204565
## iter 10 value 77.903940
## iter 20 value 69.317038
## iter 30 value 63.987192
## iter 40 value 57.967338
## iter 50 value 56.541860
## iter 60 value 56.213978
## iter 70 value 56.191838
## final value 56.191761
## converged
## # weights: 109
## initial value 98.917610
## iter 10 value 52.279871
## iter 20 value 25.291209
## iter 30 value 19.246674
## iter 40 value 17.625722
## iter 50 value 16.587821
## iter 60 value 15.734542
## iter 70 value 15.531989
## iter 80 value 14.800242
## iter 90 value 14.701720
## iter 100 value 14.701363
## final value 14.701363
## stopped after 100 iterations
## # weights: 179
## initial value 101.512058
## iter 10 value 27.811450
## iter 20 value 16.304479
## iter 30 value 14.120042
## iter 40 value 12.382638
## iter 50 value 11.901144
## iter 60 value 11.770871
## iter 70 value 11.753453
## iter 80 value 11.712876
## iter 90 value 11.644096
## iter 100 value 11.642399
## final value 11.642399
## stopped after 100 iterations
## # weights: 39
## initial value 108.666967
## iter 10 value 57.735899
## iter 20 value 55.543966
## iter 30 value 52.538535
## iter 40 value 51.607325
## iter 50 value 50.763428
```

```
## iter 60 value 50.319685
## iter 70 value 49.824164
## iter 80 value 49.105104
## iter 90 value 47.942290
## iter 100 value 45.078175
## final value 45.078175
## stopped after 100 iterations
## # weights: 109
## initial value 108.529147
## iter 10 value 38.569147
## iter 20 value 22.200431
## iter 30 value 19.489820
## iter 40 value 18.875881
## iter 50 value 18.553504
## iter 60 value 18.253283
## iter 70 value 17.943530
## iter 80 value 17.566034
## iter 90 value 17.267717
## iter 100 value 15.671424
## final value 15.671424
## stopped after 100 iterations
## # weights: 179
## initial value 111.860137
## iter 10 value 23.551875
## iter 20 value 15.385911
## iter 30 value 12.002694
## iter 40 value 9.867249
## iter 50 value 8.245328
## iter 60 value 6.836921
## iter 70 value 5.985680
## iter 80 value 5.896484
## iter 90 value 5.886520
## iter 100 value 4.085010
## final value 4.085010
## stopped after 100 iterations
## # weights: 39
## initial value 125.381831
## iter 10 value 59.430396
## iter 20 value 54.499055
## iter 30 value 53.870191
## iter 40 value 50.254981
## iter 50 value 48.537299
## iter 60 value 45.884392
## iter 70 value 44.670564
## iter 80 value 43.130402
## iter 90 value 42.215533
## iter 100 value 41.757088
## final value 41.757088
## stopped after 100 iterations
## # weights: 109
## initial value 99.277951
## iter 10 value 30.409227
## iter 20 value 22.269757
## iter 30 value 21.168598
```

```
## iter 40 value 20.948904
## iter 50 value 20.228458
## iter 60 value 19.645830
## iter 70 value 19.170533
## iter 80 value 18.687912
## iter 90 value 17.856895
## iter 100 value 17.586790
## final value 17.586790
## stopped after 100 iterations
## # weights: 179
## initial value 117.691961
## iter 10 value 24.356242
## iter 20 value 4.174138
## iter 30 value 2.542639
## iter 40 value 1.987332
## iter 50 value 1.921058
## iter 60 value 1.398859
## iter 70 value 1.336290
## iter 80 value 0.128721
## iter 90 value 0.028904
## iter 100 value 0.014181
## final value 0.014181
## stopped after 100 iterations
## # weights: 39
## initial value 103.734468
## iter 10 value 82.434762
## iter 20 value 76.419268
## iter 30 value 74.340681
## iter 40 value 66.736398
## iter 50 value 60.171751
## iter 60 value 59.261387
## iter 70 value 59.160117
## iter 80 value 59.153974
## iter 90 value 59.149758
## final value 59.149371
## converged
## # weights: 109
## initial value 117.087569
## iter 10 value 54.563837
## iter 20 value 29.708833
## iter 30 value 20.466822
## iter 40 value 16.553459
## iter 50 value 16.027797
## iter 60 value 15.908921
## iter 70 value 15.904742
## final value 15.904707
## converged
## # weights: 179
## initial value 133.017919
## iter 10 value 51.055854
## iter 20 value 26.825720
## iter 30 value 17.235838
## iter 40 value 14.839835
## iter 50 value 13.504617
```

```
## iter 60 value 13.143227
## iter 70 value 13.070481
## iter 80 value 13.065182
## iter 90 value 13.064800
## final value 13.064796
## converged
## # weights: 39
## initial value 111.560022
## iter 10 value 61.382274
## iter 20 value 59.144678
## iter 30 value 57.434150
## iter 40 value 56.551149
## iter 50 value 56.196886
## iter 60 value 56.191776
## iter 70 value 56.178702
## iter 80 value 56.173533
## iter 90 value 56.171754
## iter 100 value 53.370592
## final value 53.370592
## stopped after 100 iterations
## # weights: 109
## initial value 105.656784
## iter 10 value 29.579034
## iter 20 value 17.022579
## iter 30 value 11.998677
## iter 40 value 11.604007
## iter 50 value 10.964531
## iter 60 value 10.483137
## iter 70 value 10.140855
## iter 80 value 6.376229
## iter 90 value 2.614649
## iter 100 value 1.586401
## final value 1.586401
## stopped after 100 iterations
## # weights: 179
## initial value 130.620688
## iter 10 value 18.945212
## iter 20 value 5.478681
## iter 30 value 0.324880
## iter 40 value 0.264315
## iter 50 value 0.248031
## iter 60 value 0.226667
## iter 70 value 0.199510
## iter 80 value 0.184890
## iter 90 value 0.166567
## iter 100 value 0.145279
## final value 0.145279
## stopped after 100 iterations
## # weights: 39
## initial value 132.634666
## iter 10 value 66.432181
## iter 20 value 59.978997
## iter 30 value 57.184030
## iter 40 value 55.396428
```

```
## iter 50 value 55.266661
## iter 60 value 54.554297
## iter 70 value 54.413852
## iter 80 value 51.479047
## iter 90 value 50.611921
## iter 100 value 50.152796
## final value 50.152796
## stopped after 100 iterations
## # weights: 109
## initial value 112.325222
## iter 10 value 45.065819
## iter 20 value 41.828079
## iter 30 value 40.897232
## iter 40 value 38.441083
## iter 50 value 37.514907
## iter 60 value 37.398270
## iter 70 value 37.294575
## iter 80 value 36.842715
## iter 90 value 35.247142
## iter 100 value 34.866592
## final value 34.866592
## stopped after 100 iterations
## # weights: 179
## initial value 117.848681
## iter 10 value 38.381277
## iter 20 value 12.627887
## iter 30 value 11.528537
## iter 40 value 3.852755
## iter 50 value 1.475339
## iter 60 value 1.407523
## iter 70 value 1.392995
## iter 80 value 1.390037
## iter 90 value 1.387114
## iter 100 value 1.386871
## final value 1.386871
## stopped after 100 iterations
## # weights: 39
## initial value 101.476665
## iter 10 value 65.977246
## iter 20 value 58.949671
## iter 30 value 57.805392
## iter 40 value 57.360984
## iter 50 value 57.352099
## iter 60 value 57.186211
## iter 70 value 57.150682
## final value 57.150674
## converged
## # weights: 109
## initial value 113.489608
## iter 10 value 45.925595
## iter 20 value 28.292826
## iter 30 value 21.638044
## iter 40 value 16.787323
## iter 50 value 16.024587
```

```
## iter 60 value 16.015320
## final value 16.015268
## converged
## # weights: 179
## initial value 121.361312
## iter 10 value 47.459172
## iter 20 value 18.961341
## iter 30 value 13.396851
## iter 40 value 12.608863
## iter 50 value 12.338122
## iter 60 value 12.238220
## iter 70 value 12.212839
## iter 80 value 12.156551
## iter 90 value 12.149355
## iter 100 value 12.133294
## final value 12.133294
## stopped after 100 iterations
## # weights: 39
## initial value 109.661937
## iter 10 value 77.565882
## iter 20 value 62.761738
## iter 30 value 59.549675
## iter 40 value 59.533504
## iter 50 value 59.529589
## iter 60 value 59.521517
## iter 70 value 59.514869
## iter 80 value 59.508767
## iter 90 value 59.507936
## iter 100 value 59.503015
## final value 59.503015
## stopped after 100 iterations
## # weights: 109
## initial value 101.397478
## iter 10 value 23.124885
## iter 20 value 16.422129
## iter 30 value 11.303697
## iter 40 value 8.889958
## iter 50 value 7.892528
## iter 60 value 7.706374
## iter 70 value 7.518262
## iter 80 value 7.096966
## iter 90 value 7.020792
## iter 100 value 6.995361
## final value 6.995361
## stopped after 100 iterations
## # weights: 179
## initial value 119.905231
## iter 10 value 23.534014
## iter 20 value 7.603594
## iter 30 value 6.242536
## iter 40 value 6.148501
## iter 50 value 6.118216
## iter 60 value 6.016889
## iter 70 value 5.903810
```

```
## iter 80 value 5.801547
## iter 90 value 5.774574
## iter 100 value 5.743757
## final value 5.743757
## stopped after 100 iterations
## # weights: 39
## initial value 104.365260
## iter 10 value 63.392815
## iter 20 value 54.505861
## iter 30 value 53.204722
## iter 40 value 53.054473
## iter 50 value 53.047051
## iter 60 value 53.046884
## iter 70 value 53.046021
## iter 80 value 52.591980
## iter 90 value 51.817189
## iter 100 value 51.253995
## final value 51.253995
## stopped after 100 iterations
## # weights: 109
## initial value 102.003652
## iter 10 value 65.106586
## iter 20 value 59.726281
## iter 30 value 37.507071
## iter 40 value 34.600436
## iter 50 value 34.201005
## iter 60 value 34.095931
## iter 70 value 33.862198
## iter 80 value 33.703301
## iter 90 value 33.345903
## iter 100 value 33.319580
## final value 33.319580
## stopped after 100 iterations
## # weights: 179
## initial value 104.296831
## iter 10 value 24.826703
## iter 20 value 9.417795
## iter 30 value 3.751953
## iter 40 value 1.430296
## iter 50 value 1.390428
## iter 60 value 1.386499
## iter 70 value 1.386319
## final value 1.386309
## converged
## # weights: 39
## initial value 112.249720
## iter 10 value 82.377741
## iter 20 value 65.225815
## iter 30 value 61.082804
## iter 40 value 60.321273
## iter 50 value 59.766642
## iter 60 value 59.460017
## iter 70 value 59.275278
## iter 80 value 59.269600
```

```
## final value 59.269597
## converged
## # weights: 109
## initial value 98.417431
## iter 10 value 46.763227
## iter 20 value 27.148624
## iter 30 value 19.624904
## iter 40 value 17.051812
## iter 50 value 16.866195
## iter 60 value 16.850877
## iter 70 value 16.816337
## iter 80 value 16.815395
## iter 80 value 16.815395
## iter 80 value 16.815395
## final value 16.815395
## converged
## # weights: 179
## initial value 109.204876
## iter 10 value 31.394152
## iter 20 value 18.959073
## iter 30 value 13.412071
## iter 40 value 12.603279
## iter 50 value 12.299192
## iter 60 value 12.246947
## iter 70 value 12.235987
## iter 80 value 12.235343
## final value 12.235338
## converged
## # weights: 39
## initial value 102.465655
## iter 10 value 67.804516
## iter 20 value 59.106721
## iter 30 value 58.414318
## iter 40 value 58.092958
## iter 50 value 58.033480
## iter 60 value 57.403092
## iter 70 value 57.350592
## iter 80 value 56.985867
## iter 90 value 56.605511
## iter 100 value 54.338919
## final value 54.338919
## stopped after 100 iterations
## # weights: 109
## initial value 107.082272
## iter 10 value 44.120119
## iter 20 value 32.415559
## iter 30 value 24.105333
## iter 40 value 22.052746
## iter 50 value 21.362405
## iter 60 value 21.258356
## iter 70 value 20.308673
## iter 80 value 19.719476
## iter 90 value 18.645346
## iter 100 value 16.289253
```

```
## final value 16.289253
## stopped after 100 iterations
## # weights: 179
## initial value 100.450887
## iter 10 value 27.959970
## iter 20 value 9.859642
## iter 30 value 7.553618
## iter 40 value 7.406723
## iter 50 value 5.801167
## iter 60 value 5.518197
## iter 70 value 2.177248
## iter 80 value 2.063836
## iter 90 value 0.184993
## iter 100 value 0.151403
## final value 0.151403
## stopped after 100 iterations
## # weights: 39
## initial value 100.013615
## iter 10 value 72.787279
## iter 20 value 63.070146
## iter 30 value 59.665286
## iter 40 value 53.093199
## iter 50 value 48.811437
## iter 60 value 47.399302
## iter 70 value 47.354606
## iter 80 value 47.353988
## iter 90 value 47.353938
## iter 100 value 47.353815
## final value 47.353815
## stopped after 100 iterations
## # weights: 109
## initial value 102.732668
## iter 10 value 25.540886
## iter 20 value 17.464998
## iter 30 value 14.543454
## iter 40 value 12.149225
## iter 50 value 12.079522
## iter 60 value 12.017035
## iter 70 value 11.977124
## iter 80 value 11.957405
## iter 90 value 11.828919
## iter 100 value 7.848346
## final value 7.848346
## stopped after 100 iterations
## # weights: 179
## initial value 101.654034
## iter 10 value 24.646667
## iter 20 value 12.122785
## iter 30 value 7.874649
## iter 40 value 4.285113
## iter 50 value 2.226407
## iter 60 value 1.742910
## iter 70 value 1.450928
## iter 80 value 1.404949
```

```
## iter 90 value 1.393107
## iter 100 value 1.388056
## final value 1.388056
## stopped after 100 iterations
## # weights: 39
## initial value 107.346384
## iter 10 value 79.371858
## iter 20 value 67.848347
## iter 30 value 61.103750
## iter 40 value 58.935888
## iter 50 value 58.887047
## iter 60 value 58.807020
## iter 70 value 58.636979
## iter 80 value 58.630905
## final value 58.630798
## converged
## # weights: 109
## initial value 105.446120
## iter 10 value 52.118723
## iter 20 value 34.099337
## iter 30 value 25.456442
## iter 40 value 19.012629
## iter 50 value 16.690469
## iter 60 value 16.368163
## iter 70 value 16.290256
## iter 80 value 16.275549
## iter 90 value 16.275493
## iter 90 value 16.275492
## iter 90 value 16.275492
## final value 16.275492
## converged
## # weights: 179
## initial value 113.423118
## iter 10 value 39.074852
## iter 20 value 20.573736
## iter 30 value 15.891826
## iter 40 value 15.059643
## iter 50 value 14.109002
## iter 60 value 13.536565
## iter 70 value 13.480627
## iter 80 value 13.388071
## iter 90 value 13.367145
## iter 100 value 13.367023
## final value 13.367023
## stopped after 100 iterations
## # weights: 39
## initial value 99.011014
## iter 10 value 67.534745
## iter 20 value 54.632320
## iter 30 value 42.330114
## iter 40 value 38.200964
## iter 50 value 37.347592
## iter 60 value 37.089102
## iter 70 value 37.064313
```

```
## iter 80 value 37.050889
## iter 90 value 37.041796
## iter 100 value 37.034643
## final value 37.034643
## stopped after 100 iterations
## # weights: 109
## initial value 124.201139
## iter 10 value 63.366108
## iter 20 value 36.294349
## iter 30 value 29.860287
## iter 40 value 23.996252
## iter 50 value 21.145933
## iter 60 value 20.633198
## iter 70 value 18.287842
## iter 80 value 15.414426
## iter 90 value 14.510657
## iter 100 value 14.006058
## final value 14.006058
## stopped after 100 iterations
## # weights: 179
## initial value 110.248930
## iter 10 value 29.247711
## iter 20 value 6.927468
## iter 30 value 3.937428
## iter 40 value 0.771298
## iter 50 value 0.540617
## iter 60 value 0.495624
## iter 70 value 0.454813
## iter 80 value 0.429972
## iter 90 value 0.398909
## iter 100 value 0.347456
## final value 0.347456
## stopped after 100 iterations
## # weights: 39
## initial value 101.112182
## iter 10 value 69.163739
## iter 20 value 64.403387
## iter 30 value 63.118340
## iter 40 value 61.958518
## iter 50 value 59.887506
## iter 60 value 57.873803
## iter 70 value 56.737777
## iter 80 value 55.862088
## iter 90 value 55.199492
## iter 100 value 55.130159
## final value 55.130159
## stopped after 100 iterations
## # weights: 109
## initial value 114.873206
## iter 10 value 48.086287
## iter 20 value 25.003243
## iter 30 value 8.093000
## iter 40 value 7.262369
## iter 50 value 7.232718
```

```
## iter 60 value 7.192405
## iter 70 value 7.177739
## iter 80 value 7.170389
## iter 90 value 7.170080
## iter 100 value 7.168865
## final value 7.168865
## stopped after 100 iterations
## # weights: 179
## initial value 112.545934
## iter 10 value 25.580178
## iter 20 value 4.309494
## iter 30 value 0.679900
## iter 40 value 0.020290
## iter 50 value 0.001472
## final value 0.000067
## converged
## # weights: 39
## initial value 100.628925
## iter 10 value 84.191757
## iter 20 value 64.405217
## iter 30 value 60.069852
## iter 40 value 59.206315
## iter 50 value 58.696651
## iter 60 value 58.386407
## iter 70 value 58.077664
## iter 80 value 57.963999
## iter 90 value 57.746401
## iter 100 value 57.656410
## final value 57.656410
## stopped after 100 iterations
## # weights: 109
## initial value 102.712319
## iter 10 value 57.271088
## iter 20 value 32.577705
## iter 30 value 21.161682
## iter 40 value 19.056265
## iter 50 value 18.537374
## iter 60 value 17.898509
## iter 70 value 17.072957
## iter 80 value 16.342303
## iter 90 value 15.862463
## iter 100 value 15.845592
## final value 15.845592
## stopped after 100 iterations
## # weights: 179
## initial value 144.385534
## iter 10 value 65.453418
## iter 20 value 30.082407
## iter 30 value 19.865822
## iter 40 value 14.870588
## iter 50 value 13.052061
## iter 60 value 12.660293
## iter 70 value 12.633034
## iter 80 value 12.629903
```

```
## iter 90 value 12.629522
## iter 100 value 12.629374
## final value 12.629374
## stopped after 100 iterations
## # weights: 39
## initial value 119.248779
## iter 10 value 81.651733
## iter 20 value 64.698356
## iter 30 value 61.569624
## iter 40 value 60.096654
## iter 50 value 58.277669
## iter 60 value 57.227449
## iter 70 value 55.308972
## iter 80 value 54.032072
## iter 90 value 52.164846
## iter 100 value 51.954529
## final value 51.954529
## stopped after 100 iterations
## # weights: 109
## initial value 97.029798
## iter 10 value 29.309002
## iter 20 value 20.818587
## iter 30 value 15.780611
## iter 40 value 9.409803
## iter 50 value 7.279160
## iter 60 value 6.842011
## iter 70 value 3.296708
## iter 80 value 2.129853
## iter 90 value 1.710250
## iter 100 value 0.408592
## final value 0.408592
## stopped after 100 iterations
## # weights: 179
## initial value 112.177852
## iter 10 value 23.286637
## iter 20 value 8.737599
## iter 30 value 4.669394
## iter 40 value 4.021378
## iter 50 value 3.949291
## iter 60 value 0.281557
## iter 70 value 0.167801
## iter 80 value 0.154585
## iter 90 value 0.146027
## iter 100 value 0.139926
## final value 0.139926
## stopped after 100 iterations
## # weights: 39
## initial value 118.519049
## iter 10 value 61.501998
## iter 20 value 55.210430
## iter 30 value 52.539765
## iter 40 value 48.348728
## iter 50 value 46.697769
## iter 60 value 46.550681
```

```
## iter 70 value 46.435832
## iter 80 value 46.293176
## iter 90 value 46.292450
## iter 100 value 46.227850
## final value 46.227850
## stopped after 100 iterations
## # weights: 109
## initial value 108.280884
## iter 10 value 60.686264
## iter 20 value 25.029261
## iter 30 value 14.629653
## iter 40 value 10.722690
## iter 50 value 6.349507
## iter 60 value 5.936056
## iter 70 value 5.929903
## final value 5.929868
## converged
## # weights: 179
## initial value 108.176733
## iter 10 value 27.574741
## iter 20 value 14.631122
## iter 30 value 8.158751
## iter 40 value 6.909229
## iter 50 value 6.300621
## iter 60 value 5.723525
## iter 70 value 5.421021
## iter 80 value 5.369928
## iter 90 value 5.331157
## iter 100 value 2.061337
## final value 2.061337
## stopped after 100 iterations
## # weights: 39
## initial value 115.232971
## iter 10 value 69.758693
## iter 20 value 64.761897
## iter 30 value 61.181257
## iter 40 value 59.399406
## iter 50 value 58.995393
## iter 60 value 57.818405
## iter 70 value 57.360523
## iter 80 value 57.345151
## final value 57.345145
## converged
## # weights: 109
## initial value 108.911437
## iter 10 value 61.028015
## iter 20 value 35.752664
## iter 30 value 21.891819
## iter 40 value 18.258882
## iter 50 value 16.936475
## iter 60 value 16.436822
## iter 70 value 16.271295
## iter 80 value 16.261078
## iter 90 value 16.251588
```

```
## iter 100 value 16.251409
## final value 16.251409
## stopped after 100 iterations
## # weights: 179
## initial value 119.041526
## iter 10 value 49.206682
## iter 20 value 22.421851
## iter 30 value 13.948967
## iter 40 value 12.814434
## iter 50 value 12.341171
## iter 60 value 12.268220
## iter 70 value 12.229556
## iter 80 value 12.197191
## iter 90 value 12.174379
## iter 100 value 12.157878
## final value 12.157878
## stopped after 100 iterations
## # weights: 39
## initial value 97.215681
## iter 10 value 70.697269
## iter 20 value 64.682495
## iter 30 value 64.362166
## iter 40 value 64.350124
## iter 50 value 64.313232
## iter 60 value 64.065389
## iter 70 value 64.009461
## iter 80 value 63.876952
## iter 90 value 61.527064
## iter 100 value 60.714226
## final value 60.714226
## stopped after 100 iterations
## # weights: 109
## initial value 106.258342
## iter 10 value 39.982607
## iter 20 value 13.739876
## iter 30 value 9.679355
## iter 40 value 9.108804
## iter 50 value 8.798725
## iter 60 value 8.674726
## iter 70 value 8.638131
## iter 80 value 8.121258
## iter 90 value 8.043704
## iter 100 value 8.016082
## final value 8.016082
## stopped after 100 iterations
## # weights: 179
## initial value 119.048084
## iter 10 value 27.697311
## iter 20 value 9.322368
## iter 30 value 3.989614
## iter 40 value 2.970221
## iter 50 value 2.897294
## iter 60 value 2.888727
## iter 70 value 2.880495
```

```
## iter 80 value 2.870163
## iter 90 value 2.860033
## iter 100 value 2.853257
## final value 2.853257
## stopped after 100 iterations
## # weights: 39
## initial value 111.149016
## iter 10 value 59.249574
## iter 20 value 56.799885
## iter 30 value 55.768447
## iter 40 value 53.866826
## iter 50 value 53.638322
## iter 60 value 52.619169
## iter 70 value 52.563062
## iter 80 value 51.690572
## iter 90 value 50.205577
## iter 100 value 50.113594
## final value 50.113594
## stopped after 100 iterations
## # weights: 109
## initial value 120.478818
## iter 10 value 67.833659
## iter 20 value 41.302787
## iter 30 value 21.157903
## iter 40 value 16.454311
## iter 50 value 12.549225
## iter 60 value 12.479356
## iter 70 value 12.460694
## iter 80 value 10.364233
## iter 90 value 10.364117
## final value 10.363928
## converged
## # weights: 179
## initial value 114.454145
## iter 10 value 16.939293
## iter 20 value 4.300866
## iter 30 value 3.082681
## iter 40 value 2.905748
## iter 50 value 2.885898
## iter 60 value 2.877612
## iter 70 value 2.875402
## iter 80 value 2.706516
## iter 90 value 2.051878
## iter 100 value 1.903445
## final value 1.903445
## stopped after 100 iterations
## # weights: 39
## initial value 100.387194
## iter 10 value 70.217411
## iter 20 value 60.755226
## iter 30 value 58.459146
## iter 40 value 58.395455
## iter 50 value 58.393270
## iter 60 value 58.392859
```

```
## iter 70 value 58.384166
## iter 80 value 58.139540
## iter 90 value 57.940436
## iter 100 value 57.734953
## final value 57.734953
## stopped after 100 iterations
## # weights: 109
## initial value 100.236716
## iter 10 value 57.664173
## iter 20 value 40.208650
## iter 30 value 34.571060
## iter 40 value 31.554027
## iter 50 value 30.273822
## iter 60 value 23.253135
## iter 70 value 18.033571
## iter 80 value 16.815295
## iter 90 value 16.413356
## iter 100 value 16.372940
## final value 16.372940
## stopped after 100 iterations
## # weights: 179
## initial value 105.599145
## iter 10 value 35.419437
## iter 20 value 21.509726
## iter 30 value 13.699850
## iter 40 value 12.722746
## iter 50 value 12.410336
## iter 60 value 12.237326
## iter 70 value 12.208050
## iter 80 value 12.205368
## iter 90 value 12.205307
## final value 12.205307
## converged
## # weights: 39
## initial value 104.021162
## iter 10 value 67.735671
## iter 20 value 54.975114
## iter 30 value 51.190925
## iter 40 value 49.416031
## iter 50 value 46.811495
## iter 60 value 46.023158
## iter 70 value 46.014531
## iter 80 value 46.012534
## iter 90 value 46.011122
## iter 100 value 46.009466
## final value 46.009466
## stopped after 100 iterations
## # weights: 109
## initial value 103.029809
## iter 10 value 37.001603
## iter 20 value 26.585757
## iter 30 value 25.544211
## iter 40 value 22.505798
## iter 50 value 21.856027
```

```
## iter 60 value 21.575999
## iter 70 value 20.677717
## iter 80 value 19.759467
## iter 90 value 19.398965
## iter 100 value 19.169482
## final value 19.169482
## stopped after 100 iterations
## # weights: 179
## initial value 121.796246
## iter 10 value 16.741709
## iter 20 value 7.182538
## iter 30 value 6.069735
## iter 40 value 5.904265
## iter 50 value 3.445190
## iter 60 value 3.197906
## iter 70 value 3.174851
## iter 80 value 3.168265
## iter 90 value 3.159825
## iter 100 value 2.033817
## final value 2.033817
## stopped after 100 iterations
## # weights: 39
## initial value 99.835018
## iter 10 value 64.139714
## iter 20 value 59.136419
## iter 30 value 58.351445
## iter 40 value 56.894510
## iter 50 value 56.145388
## iter 60 value 55.849724
## iter 70 value 55.343415
## iter 80 value 54.062536
## iter 90 value 48.673703
## iter 100 value 46.916919
## final value 46.916919
## stopped after 100 iterations
## # weights: 109
## initial value 103.989234
## iter 10 value 25.531325
## iter 20 value 14.618317
## iter 30 value 11.585792
## iter 40 value 10.808705
## iter 50 value 10.686602
## iter 60 value 10.681725
## iter 70 value 10.681480
## iter 80 value 10.681342
## final value 10.681337
## converged
## # weights: 179
## initial value 103.651941
## iter 10 value 20.573670
## iter 20 value 7.258001
## iter 30 value 6.427372
## iter 40 value 5.321906
## iter 50 value 5.038971
```

```
## iter 60 value 3.341356
## iter 70 value 1.419256
## iter 80 value 1.398305
## iter 90 value 1.389625
## iter 100 value 1.387501
## final value 1.387501
## stopped after 100 iterations
## # weights: 39
## initial value 118.101022
## iter 10 value 74.528816
## iter 20 value 66.073868
## iter 30 value 63.539119
## iter 40 value 62.167036
## iter 50 value 59.977259
## iter 60 value 58.359113
## iter 70 value 57.095552
## iter 80 value 57.085091
## final value 57.085072
## converged
## # weights: 109
## initial value 128.313116
## iter 10 value 45.443998
## iter 20 value 30.975705
## iter 30 value 22.556465
## iter 40 value 19.120438
## iter 50 value 16.689159
## iter 60 value 16.033363
## iter 70 value 15.971582
## iter 80 value 15.970670
## final value 15.970669
## converged
## # weights: 179
## initial value 125.929231
## iter 10 value 35.058444
## iter 20 value 22.100637
## iter 30 value 15.814520
## iter 40 value 13.818183
## iter 50 value 12.628284
## iter 60 value 12.358238
## iter 70 value 12.307827
## iter 80 value 12.306101
## final value 12.306096
## converged
## # weights: 39
## initial value 99.326396
## iter 10 value 82.216544
## iter 20 value 64.939521
## iter
       30 value 63.908722
## iter 40 value 62.587678
## iter 50 value 60.093618
## iter 60 value 59.953121
## iter 70 value 58.524183
## iter 80 value 57.793594
## iter 90 value 57.700791
```

```
## iter 100 value 57.542937
## final value 57.542937
## stopped after 100 iterations
## # weights: 109
## initial value 105.851003
## iter 10 value 48.300781
## iter 20 value 29.207361
## iter 30 value 23.583296
## iter 40 value 16.083089
## iter 50 value 10.835569
## iter 60 value 7.312926
## iter 70 value 6.845018
## iter 80 value 3.670466
## iter 90 value 0.329788
## iter 100 value 0.255631
## final value 0.255631
## stopped after 100 iterations
## # weights: 179
## initial value 111.053173
## iter 10 value 28.161055
## iter 20 value 15.727787
## iter 30 value 12.524916
## iter 40 value 8.331890
## iter 50 value 5.706022
## iter 60 value 2.626522
## iter 70 value 0.709088
## iter 80 value 0.201203
## iter 90 value 0.176543
## iter 100 value 0.161353
## final value 0.161353
## stopped after 100 iterations
## # weights: 39
## initial value 97.771886
## iter 10 value 71.802876
## iter 20 value 48.792172
## iter 30 value 42.675303
## iter 40 value 41.464500
## iter 50 value 41.022904
## iter 60 value 40.919862
## iter 70 value 36.891380
## iter 80 value 36.585855
## iter 90 value 36.344996
## iter 100 value 36.162141
## final value 36.162141
## stopped after 100 iterations
## # weights: 109
## initial value 96.353102
## iter 10 value 41.644640
## iter 20 value 22.653758
## iter 30 value 20.355682
## iter 40 value 19.091654
## iter 50 value 14.810236
## iter 60 value 12.822762
## iter 70 value 12.321664
```

```
## iter 80 value 12.189900
## iter 90 value 12.129413
## iter 100 value 9.589919
## final value 9.589919
## stopped after 100 iterations
## # weights: 179
## initial value 130.062787
## iter 10 value 24.911806
## iter 20 value 0.176277
## iter 30 value 0.002512
## iter 40 value 0.000823
## iter 50 value 0.000186
## final value 0.000099
## converged
## # weights: 39
## initial value 111.723343
## iter 10 value 89.499088
## iter 20 value 70.174704
## iter 30 value 59.073672
## iter 40 value 57.701416
## iter 50 value 56.914247
## iter 60 value 56.825340
## iter 70 value 56.822534
## iter 80 value 56.555027
## iter 90 value 56.527851
## final value 56.527762
## converged
## # weights: 109
## initial value 123.109595
## iter 10 value 67.009656
## iter 20 value 46.738516
## iter 30 value 30.910552
## iter 40 value 22.322239
## iter 50 value 17.473176
## iter 60 value 16.721120
## iter 70 value 16.681969
## iter 80 value 16.679929
## final value 16.679916
## converged
## # weights: 179
## initial value 105.469595
## iter 10 value 50.454871
## iter 20 value 26.553241
## iter 30 value 15.000018
## iter 40 value 12.786934
## iter 50 value 12.580253
## iter 60 value 12.548610
## iter 70 value 12.519697
## iter 80 value 12.498324
## iter 90 value 12.494423
## iter 100 value 12.493171
## final value 12.493171
## stopped after 100 iterations
## # weights: 39
```

```
## initial value 102.425049
## iter 10 value 80.706777
## iter 20 value 75.749902
## iter 30 value 75.720012
## iter 40 value 75.717355
## iter 50 value 75.706752
## iter 60 value 73.783920
## iter 70 value 73.745947
## iter 80 value 70.673941
## iter 90 value 68.181006
## iter 100 value 67.532559
## final value 67.532559
## stopped after 100 iterations
## # weights: 109
## initial value 105.762316
## iter 10 value 17.454759
## iter 20 value 13.722771
## iter 30 value 13.084458
## iter 40 value 8.736795
## iter 50 value 8.163552
## iter 60 value 8.144101
## iter 70 value 8.120489
## iter 80 value 8.105753
## iter 90 value 8.093288
## iter 100 value 8.082557
## final value 8.082557
## stopped after 100 iterations
## # weights: 179
## initial value 105.951267
## iter 10 value 20.900748
## iter 20 value 4.647714
## iter 30 value 3.801717
## iter 40 value 2.498879
## iter 50 value 2.473813
## iter 60 value 2.129036
## iter 70 value 2.074289
## iter 80 value 1.601208
## iter 90 value 1.586765
## iter 100 value 1.565705
## final value 1.565705
## stopped after 100 iterations
## # weights: 39
## initial value 102.364770
## iter 10 value 66.870295
## iter 20 value 57.882774
## iter 30 value 53.996184
## iter 40 value 53.496150
## iter 50 value 53.003373
## iter 60 value 52.249269
## iter 70 value 52.041397
## iter 80 value 51.505918
## iter 90 value 50.585210
## iter 100 value 50.381898
## final value 50.381898
```

```
## stopped after 100 iterations
## # weights: 109
## initial value 103.884128
## iter 10 value 52.189002
## iter 20 value 29.958777
## iter 30 value 29.284701
## iter 40 value 28.206434
## iter 50 value 28.154988
## iter 60 value 28.099348
## iter 70 value 26.716358
## iter 80 value 26.563167
## iter 90 value 25.397694
## iter 100 value 24.784647
## final value 24.784647
## stopped after 100 iterations
## # weights: 179
## initial value 115.288821
## iter 10 value 31.923634
## iter 20 value 4.685420
## iter 30 value 3.262050
## iter 40 value 2.869201
## iter 50 value 2.593985
## iter 60 value 1.388662
## iter 70 value 1.386920
## iter 80 value 1.384117
## iter 90 value 0.005584
## iter 100 value 0.001629
## final value 0.001629
## stopped after 100 iterations
## # weights: 39
## initial value 104.469913
## iter 10 value 66.610632
## iter 20 value 61.101961
## iter 30 value 60.470194
## iter 40 value 60.456492
## iter 50 value 60.455696
## final value 60.455694
## converged
## # weights: 109
## initial value 139.026525
## iter 10 value 71.763978
## iter 20 value 58.340810
## iter 30 value 42.122214
## iter 40 value 31.121239
## iter 50 value 21.540366
## iter 60 value 18.029692
## iter 70 value 17.033527
## iter 80 value 16.776304
## iter 90 value 16.747919
## iter 100 value 16.730891
## final value 16.730891
## stopped after 100 iterations
## # weights: 179
## initial value 108.681361
```

```
## iter 10 value 49.770746
## iter 20 value 25.196775
## iter 30 value 15.789021
## iter 40 value 13.325940
## iter 50 value 12.731487
## iter 60 value 12.389203
## iter 70 value 12.237870
## iter 80 value 12.230403
## iter 90 value 12.229750
## iter 100 value 12.229739
## final value 12.229739
## stopped after 100 iterations
## # weights: 39
## initial value 103.781789
## iter 10 value 69.854091
## iter 20 value 64.478612
## iter 30 value 64.455133
## iter 40 value 64.447088
## iter 50 value 64.246679
## iter 60 value 63.881565
## iter 70 value 60.092897
## iter 80 value 58.720131
## iter 90 value 58.498487
## iter 100 value 58.404287
## final value 58.404287
## stopped after 100 iterations
## # weights: 109
## initial value 110.858254
## iter 10 value 29.001877
## iter 20 value 16.789302
## iter 30 value 7.506211
## iter 40 value 6.219350
## iter 50 value 6.161001
## iter 60 value 4.058854
## iter 70 value 3.465726
## iter 80 value 3.435017
## iter 90 value 3.427261
## iter 100 value 3.378623
## final value 3.378623
## stopped after 100 iterations
## # weights: 179
## initial value 110.282140
## iter 10 value 28.800328
## iter 20 value 6.689439
## iter 30 value 4.789840
## iter 40 value 4.408558
## iter 50 value 4.382627
## iter
       60 value 4.362895
## iter 70 value 4.331593
## iter 80 value 4.256715
## iter 90 value 4.217943
## iter 100 value 4.196763
## final value 4.196763
## stopped after 100 iterations
```

```
## # weights: 39
## initial value 104.199667
## iter 10 value 97.135992
## iter 20 value 56.562187
## iter 30 value 53.550246
## iter 40 value 52.804511
## iter 50 value 52.711283
## iter 60 value 51.891958
## final value 51.890736
## converged
## # weights: 109
## initial value 105.783987
## iter 10 value 36.051260
## iter 20 value 22.872095
## iter 30 value 20.441291
## iter 40 value 18.904560
## iter 50 value 16.815147
## iter 60 value 14.170339
## iter 70 value 13.722262
## iter 80 value 13.485008
## iter 90 value 12.752384
## iter 100 value 12.445195
## final value 12.445195
## stopped after 100 iterations
## # weights: 179
## initial value 96.573742
## iter 10 value 12.426671
## iter 20 value 2.603671
## iter 30 value 1.917845
## iter 40 value 1.910446
## iter 50 value 1.909701
## iter 60 value 1.909643
## iter 70 value 1.909593
## iter 80 value 1.909558
## final value 1.909547
## converged
## # weights: 39
## initial value 101.891132
## iter 10 value 69.822588
## iter 20 value 61.578350
## iter 30 value 60.159555
## iter 40 value 60.018445
## iter 50 value 59.867250
## iter 60 value 59.428107
## iter 70 value 58.797566
## iter 80 value 58.706134
## final value 58.706132
## converged
## # weights: 109
## initial value 110.075670
## iter 10 value 34.387351
## iter 20 value 24.792969
## iter 30 value 17.717945
## iter 40 value 16.213801
```

```
## iter 50 value 16.092456
## iter 60 value 16.084662
## iter 70 value 16.047551
## iter 80 value 16.037774
## iter 90 value 16.037331
## final value 16.037330
## converged
## # weights: 179
## initial value 115.708334
## iter 10 value 42.720787
## iter 20 value 20.745700
## iter 30 value 14.462041
## iter 40 value 12.080624
## iter 50 value 11.655244
## iter 60 value 11.635620
## iter 70 value 11.635378
## final value 11.635378
## converged
## # weights: 39
## initial value 106.069473
## iter 10 value 60.522302
## iter 20 value 54.527704
## iter 30 value 52.284471
## iter 40 value 52.245304
## iter 50 value 52.220860
## iter 60 value 52.217541
## iter 70 value 52.211214
## iter 80 value 52.199208
## iter 90 value 52.196297
## iter 100 value 51.913042
## final value 51.913042
## stopped after 100 iterations
## # weights: 109
## initial value 100.772932
## iter 10 value 29.274090
## iter 20 value 14.998718
## iter 30 value 9.514034
## iter 40 value 5.471462
## iter 50 value 3.668145
## iter 60 value 1.707526
## iter 70 value 1.358848
## iter 80 value 1.097257
## iter 90 value 0.500683
## iter 100 value 0.317909
## final value 0.317909
## stopped after 100 iterations
## # weights: 179
## initial value 131.203469
## iter 10 value 5.399975
## iter 20 value 1.865071
## iter 30 value 1.654823
## iter 40 value 1.633759
## iter 50 value 1.616382
## iter 60 value 0.897033
```

```
## iter 70 value 0.375729
## iter 80 value 0.288597
## iter 90 value 0.269607
## iter 100 value 0.244287
## final value 0.244287
## stopped after 100 iterations
## # weights: 39
## initial value 106.725442
## iter 10 value 67.053000
## iter 20 value 56.918741
## iter 30 value 46.879204
## iter 40 value 44.334662
## iter 50 value 42.518050
## iter 60 value 40.093905
## iter 70 value 38.287855
## iter 80 value 38.150847
## iter 90 value 38.133194
## iter 100 value 36.249628
## final value 36.249628
## stopped after 100 iterations
## # weights: 109
## initial value 99.236726
## iter 10 value 44.895310
## iter 20 value 21.492675
## iter 30 value 17.203205
## iter 40 value 13.691410
## iter 50 value 13.238657
## iter 60 value 13.165587
## iter 70 value 13.120692
## iter 80 value 13.095084
## iter 90 value 13.040587
## iter 100 value 12.719417
## final value 12.719417
## stopped after 100 iterations
## # weights: 179
## initial value 111.511383
## iter 10 value 18.072316
## iter 20 value 5.464337
## iter 30 value 3.728225
## iter 40 value 2.450666
## iter 50 value 2.253851
## iter 60 value 0.037557
## iter 70 value 0.012284
## iter 80 value 0.006360
## iter 90 value 0.004647
## iter 100 value 0.001742
## final value 0.001742
## stopped after 100 iterations
## # weights: 39
## initial value 100.861818
## iter 10 value 84.419478
## iter 20 value 67.939160
## iter 30 value 60.947374
## iter 40 value 57.497592
```

```
## iter 50 value 57.031336
## iter 60 value 56.438293
## iter 70 value 56.277625
## iter 80 value 55.725951
## iter 90 value 55.380211
## iter 100 value 54.795571
## final value 54.795571
## stopped after 100 iterations
## # weights: 109
## initial value 104.752870
## iter 10 value 44.024183
## iter 20 value 20.712667
## iter 30 value 19.062730
## iter 40 value 18.680577
## iter 50 value 18.546841
## iter 60 value 17.398069
## iter 70 value 15.545792
## iter 80 value 15.291764
## iter 90 value 15.287046
## final value 15.287042
## converged
## # weights: 179
## initial value 102.107527
## iter 10 value 41.754701
## iter 20 value 21.796409
## iter 30 value 14.222965
## iter 40 value 12.493327
## iter 50 value 11.733633
## iter 60 value 11.502781
## iter 70 value 11.442943
## iter 80 value 11.436353
## iter 90 value 11.435682
## iter 100 value 11.435662
## final value 11.435662
## stopped after 100 iterations
## # weights: 39
## initial value 108.891915
## iter 10 value 63.108813
## iter 20 value 55.576404
## iter 30 value 50.921987
## iter 40 value 50.207962
## iter 50 value 47.339085
## iter 60 value 44.101764
## iter 70 value 43.799553
## iter 80 value 43.781128
## iter 90 value 42.887774
## iter 100 value 39.391811
## final value 39.391811
## stopped after 100 iterations
## # weights: 109
## initial value 110.952063
## iter 10 value 37.488567
## iter 20 value 30.185653
## iter 30 value 29.821775
```

```
## iter 40 value 29.396215
## iter 50 value 27.824730
## iter 60 value 26.626490
## iter 70 value 22.601767
## iter 80 value 15.114563
## iter 90 value 12.849792
## iter 100 value 11.744827
## final value 11.744827
## stopped after 100 iterations
## # weights: 179
## initial value 99.249948
## iter 10 value 32.736158
## iter 20 value 3.701963
## iter 30 value 2.149204
## iter 40 value 2.120049
## iter 50 value 2.102357
## iter 60 value 2.084592
## iter 70 value 2.069249
## iter 80 value 2.057716
## iter 90 value 2.036711
## iter 100 value 1.525674
## final value 1.525674
## stopped after 100 iterations
## # weights: 39
## initial value 103.216005
## iter 10 value 74.740566
## iter 20 value 60.956468
## iter 30 value 57.996336
## iter 40 value 56.196639
## iter 50 value 55.792579
## iter 60 value 55.737553
## iter 70 value 55.725442
## iter 80 value 55.723004
## iter 90 value 55.722208
## iter 100 value 55.722007
## final value 55.722007
## stopped after 100 iterations
## # weights: 109
## initial value 99.992490
## iter 10 value 29.219300
## iter 20 value 18.717110
## iter 30 value 13.955319
## iter 40 value 9.876809
## iter 50 value 3.625289
## iter 60 value 2.583629
## iter 70 value 2.287593
## iter 80 value 2.020923
## iter 90 value 1.968255
## iter 100 value 1.942548
## final value 1.942548
## stopped after 100 iterations
## # weights: 179
## initial value 110.017516
## iter 10 value 21.954782
```

```
## iter 20 value 4.025205
## iter 30 value 0.244185
## iter 40 value 0.026057
## iter 50 value 0.012078
## iter 60 value 0.004545
## iter 70 value 0.001621
## iter 80 value 0.000755
## iter 90 value 0.000383
## iter 100 value 0.000196
## final value 0.000196
## stopped after 100 iterations
## # weights: 39
## initial value 107.890868
## iter 10 value 78.820789
## iter 20 value 72.753054
## iter 30 value 67.916754
## iter 40 value 61.769060
## iter 50 value 59.796060
## iter 60 value 58.088283
## iter 70 value 58.019691
## final value 58.019672
## converged
## # weights: 109
## initial value 111.946860
## iter 10 value 48.955522
## iter 20 value 26.266252
## iter 30 value 18.749171
## iter 40 value 17.163630
## iter 50 value 16.666293
## iter 60 value 16.574378
## iter 70 value 16.568674
## final value 16.568664
## converged
## # weights: 179
## initial value 120.143642
## iter 10 value 33.821112
## iter 20 value 21.411552
## iter 30 value 14.814050
## iter 40 value 13.566355
## iter 50 value 12.211218
## iter 60 value 11.840829
## iter 70 value 11.834729
## iter 80 value 11.834604
## final value 11.834603
## converged
## # weights: 39
## initial value 99.321827
## iter 10 value 64.839847
## iter 20 value 55.493800
## iter 30 value 47.528062
## iter 40 value 43.593625
## iter 50 value 41.135095
## iter 60 value 40.483243
## iter 70 value 39.581304
```

```
## iter 80 value 37.838061
## iter 90 value 37.656369
## iter 100 value 37.570374
## final value 37.570374
## stopped after 100 iterations
## # weights: 109
## initial value 98.614268
## iter 10 value 35.386405
## iter 20 value 27.307056
## iter 30 value 10.182678
## iter 40 value 6.478055
## iter 50 value 5.518186
## iter 60 value 5.451787
## iter 70 value 5.337542
## iter 80 value 5.246091
## iter 90 value 5.140308
## iter 100 value 5.127438
## final value 5.127438
## stopped after 100 iterations
## # weights: 179
## initial value 104.556780
## iter 10 value 28.861421
## iter 20 value 16.359042
## iter 30 value 12.171576
## iter 40 value 9.095570
## iter 50 value 7.948036
## iter 60 value 6.018241
## iter 70 value 4.617890
## iter 80 value 3.994089
## iter 90 value 3.685409
## iter 100 value 2.244748
## final value 2.244748
## stopped after 100 iterations
## # weights: 39
## initial value 108.809362
## iter 10 value 82.282259
## iter 20 value 79.054473
## iter 30 value 76.648013
## iter 40 value 73.668901
## iter 50 value 68.440640
## iter 60 value 64.459789
## iter 70 value 62.144726
## iter 80 value 61.681476
## iter 90 value 59.713979
## iter 100 value 57.597446
## final value 57.597446
## stopped after 100 iterations
## # weights: 109
## initial value 105.132695
## iter 10 value 28.650164
## iter 20 value 13.742007
## iter 30 value 10.921882
## iter 40 value 10.882696
## iter 50 value 10.882002
```

```
## iter 60 value 10.881369
## iter 70 value 10.229645
## iter 80 value 10.227415
## final value 10.227413
## converged
## # weights: 179
## initial value 109.845005
## iter 10 value 60.807257
## iter 20 value 26.762643
## iter 30 value 16.621174
## iter 40 value 13.121425
## iter 50 value 6.581694
## iter 60 value 6.020114
## iter 70 value 5.716047
## iter 80 value 5.370451
## iter 90 value 5.183341
## iter 100 value 5.062079
## final value 5.062079
## stopped after 100 iterations
## # weights:
              39
## initial value 105.537692
## iter 10 value 69.110092
## iter 20 value 62.716633
## iter 30 value 60.195828
## iter 40 value 59.201423
## iter 50 value 58.486384
## iter 60 value 58.282276
## iter 70 value 57.819131
## final value 57.808905
## converged
## # weights: 109
## initial value 105.149562
## iter 10 value 51.660696
## iter 20 value 34.190558
## iter 30 value 24.122861
## iter 40 value 18.121648
## iter 50 value 17.216555
## iter 60 value 17.125182
## iter 70 value 16.624064
## iter 80 value 16.427706
## iter 90 value 16.333064
## iter 100 value 16.318269
## final value 16.318269
## stopped after 100 iterations
## # weights: 179
## initial value 121.698118
## iter 10 value 57.422570
## iter 20 value 26.810319
## iter 30 value 16.653453
## iter 40 value 14.301596
## iter 50 value 13.403715
## iter 60 value 12.803918
## iter 70 value 12.610607
## iter 80 value 12.594299
```

```
## iter 90 value 12.589890
## iter 100 value 12.588611
## final value 12.588611
## stopped after 100 iterations
## # weights: 39
## initial value 106.626626
## iter 10 value 74.087800
## iter 20 value 62.536005
## iter 30 value 55.653646
## iter 40 value 51.206462
## iter 50 value 47.699505
## iter 60 value 47.402029
## iter 70 value 47.358950
## iter 80 value 47.354316
## iter 90 value 47.353375
## iter 100 value 47.352791
## final value 47.352791
## stopped after 100 iterations
## # weights: 109
## initial value 97.909507
## iter 10 value 32.940182
## iter 20 value 16.448580
## iter 30 value 11.752310
## iter 40 value 10.351766
## iter 50 value 9.826335
## iter 60 value 8.395386
## iter 70 value 8.121411
## iter 80 value 7.872950
## iter 90 value 7.622426
## iter 100 value 7.601647
## final value 7.601647
## stopped after 100 iterations
## # weights: 179
## initial value 110.939234
## iter 10 value 18.896341
## iter 20 value 6.357458
## iter 30 value 5.017966
## iter 40 value 3.604925
## iter 50 value 3.498478
## iter 60 value 3.478419
## iter 70 value 2.912485
## iter 80 value 2.137087
## iter 90 value 2.104090
## iter 100 value 2.090416
## final value 2.090416
## stopped after 100 iterations
## # weights: 39
## initial value 107.431463
## iter 10 value 94.081265
## iter 20 value 74.152160
## iter 30 value 73.072557
## iter 40 value 70.888174
## iter 50 value 68.016665
## iter 60 value 66.639831
```

```
## iter 70 value 63.516880
## iter 80 value 63.171766
## iter 90 value 61.495808
## iter 100 value 59.995399
## final value 59.995399
## stopped after 100 iterations
## # weights: 109
## initial value 119.777607
## iter 10 value 38.426096
## iter 20 value 13.419502
## iter 30 value 12.927932
## iter 40 value 12.715735
## iter 50 value 10.333617
## iter 60 value 10.017043
## iter 70 value 7.726759
## iter 80 value 6.774761
## iter 90 value 6.772998
## iter 100 value 6.771732
## final value 6.771732
## stopped after 100 iterations
## # weights: 179
## initial value 103.608488
## iter 10 value 13.368798
## iter 20 value 0.634417
## iter 30 value 0.012090
## final value 0.000063
## converged
## # weights: 39
## initial value 99.676914
## iter 10 value 66.641104
## iter 20 value 61.282028
## iter 30 value 59.915630
## iter 40 value 59.679407
## iter 50 value 58.891108
## iter 60 value 58.070548
## iter 70 value 57.102715
## iter 80 value 56.908231
## iter 90 value 56.904636
## final value 56.904624
## converged
## # weights: 109
## initial value 111.619633
## iter 10 value 64.198248
## iter 20 value 40.050037
## iter 30 value 27.964938
## iter 40 value 18.581947
## iter 50 value 16.362594
## iter 60 value 15.816053
## iter 70 value 15.632172
## iter 80 value 15.613596
## iter 90 value 15.612880
## iter 100 value 15.612768
## final value 15.612768
## stopped after 100 iterations
```

```
## # weights: 179
## initial value 119.434561
## iter 10 value 33.830236
## iter 20 value 16.722051
## iter 30 value 13.713529
## iter 40 value 13.425083
## iter 50 value 13.070572
## iter 60 value 12.398017
## iter 70 value 12.326190
## iter 80 value 12.324373
## final value 12.324340
## converged
## # weights: 39
## initial value 111.185498
## iter 10 value 67.513777
## iter 20 value 59.708386
## iter 30 value 56.224334
## iter 40 value 55.837657
## iter 50 value 55.779602
## iter 60 value 55.668235
## iter 70 value 55.571100
## iter 80 value 55.510241
## iter 90 value 54.473456
## iter 100 value 54.270318
## final value 54.270318
## stopped after 100 iterations
## # weights: 109
## initial value 95.403886
## iter 10 value 25.218678
## iter 20 value 9.718044
## iter 30 value 9.239404
## iter 40 value 9.204647
## iter 50 value 9.115589
## iter 60 value 9.005080
## iter 70 value 8.971842
## iter 80 value 8.826012
## iter 90 value 8.819669
## iter 100 value 8.651898
## final value 8.651898
## stopped after 100 iterations
## # weights: 179
## initial value 124.887593
## iter 10 value 34.242865
## iter 20 value 12.542742
## iter 30 value 11.147125
## iter 40 value 10.449822
## iter 50 value 5.400074
## iter
       60 value 2.663995
## iter 70 value 2.576466
## iter 80 value 2.536311
## iter 90 value 2.496369
## iter 100 value 1.724131
## final value 1.724131
## stopped after 100 iterations
```

```
## # weights: 39
## initial value 108.058228
## iter 10 value 66.576457
## iter 20 value 59.831923
## iter 30 value 55.867867
## iter 40 value 55.145270
## iter 50 value 54.504772
## iter 60 value 53.764548
## iter 70 value 53.190536
## iter 80 value 52.611701
## iter 90 value 51.919836
## iter 100 value 51.599137
## final value 51.599137
## stopped after 100 iterations
## # weights: 109
## initial value 108.108231
## iter 10 value 34.286724
## iter 20 value 21.783733
## iter 30 value 16.753120
## iter 40 value 12.993237
## iter 50 value 10.432529
## iter 60 value 10.348284
## iter 70 value 10.342349
## iter 80 value 10.341741
## iter 90 value 10.341601
## iter 100 value 10.341580
## final value 10.341580
## stopped after 100 iterations
## # weights: 179
## initial value 109.183174
## iter 10 value 26.022158
## iter 20 value 6.976148
## iter 30 value 6.169617
## iter 40 value 6.159280
## iter 50 value 5.085312
## iter 60 value 2.777677
## iter 70 value 2.774322
## iter 80 value 2.773398
## iter 90 value 2.770616
## iter 100 value 1.910968
## final value 1.910968
## stopped after 100 iterations
## # weights: 39
## initial value 104.882029
## iter 10 value 79.226375
## iter 20 value 66.130154
## iter 30 value 62.480330
## iter
       40 value 60.452157
## iter 50 value 59.870426
## iter 60 value 59.529400
## iter 70 value 59.512856
## iter 80 value 59.372217
## iter 90 value 59.074648
## iter 100 value 59.011622
```

```
## final value 59.011622
## stopped after 100 iterations
## # weights: 109
## initial value 108.260308
## iter 10 value 71.546898
## iter 20 value 46.979786
## iter 30 value 25.038661
## iter 40 value 18.233025
## iter 50 value 17.165942
## iter 60 value 17.103759
## iter 70 value 17.097545
## iter 80 value 17.097418
## final value 17.097416
## converged
## # weights: 179
## initial value 108.514646
## iter 10 value 40.516515
## iter 20 value 17.317074
## iter 30 value 13.212413
## iter 40 value 12.735248
## iter 50 value 12.601368
## iter 60 value 12.534321
## iter 70 value 12.439698
## iter 80 value 12.402127
## iter 90 value 12.401633
## final value 12.401632
## converged
## # weights: 39
## initial value 107.236716
## iter 10 value 61.606762
## iter 20 value 54.433158
## iter 30 value 51.623683
## iter 40 value 51.602323
## iter 50 value 51.592410
## iter 60 value 51.582496
## iter 70 value 51.579357
## iter 80 value 51.568045
## iter 90 value 51.566858
## iter 100 value 51.561059
## final value 51.561059
## stopped after 100 iterations
## # weights: 109
## initial value 110.666515
## iter 10 value 55.952169
## iter 20 value 20.488263
## iter 30 value 7.286823
## iter 40 value 4.323604
## iter 50 value 4.307191
## iter 60 value 4.289647
## iter 70 value 4.279134
## iter 80 value 4.266235
## iter 90 value 4.258850
## iter 100 value 2.208167
## final value 2.208167
```

```
## stopped after 100 iterations
## # weights: 179
## initial value 116.198662
## iter 10 value 44.484663
## iter 20 value 16.848190
## iter 30 value 14.185958
## iter 40 value 13.854471
## iter 50 value 13.762603
## iter 60 value 13.622751
## iter 70 value 13.530162
## iter 80 value 13.336108
## iter 90 value 13.207644
## iter 100 value 12.987856
## final value 12.987856
## stopped after 100 iterations
## # weights: 179
## initial value 136.763563
## iter 10 value 41.095424
## iter 20 value 22.265719
## iter 30 value 14.637937
## iter 40 value 13.449071
## iter 50 value 12.927608
## iter 60 value 12.774159
## iter 70 value 12.768531
## iter 80 value 12.767823
## iter 90 value 12.767793
## final value 12.767792
## converged
## NB
nb_fit <- train(`PAM50.mRNA`~., data = data_norm, method = "nb",</pre>
              trControl = fitControl)
rf_fit<- train(`PAM50.mRNA`~., data = data_norm, method = "rf",
trControl = fitControl)
print(svm_fit)
## Support Vector Machines with Linear Kernel
## 80 samples
## 30 predictors
## 4 classes: 'Basal.like', 'HER2.enriched', 'Luminal.A', 'Luminal.B'
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 10 times)
## Summary of sample sizes: 72, 71, 73, 72, 74, 71, ...
## Resampling results:
##
##
                                    Mean_Sensitivity Mean_Specificity
    Accuracy Kappa
                         Mean F1
    0.773627 0.6904643 0.8352319 0.75375
##
                                                      0.9240625
##
    Mean_Pos_Pred_Value Mean_Neg_Pred_Value Mean_Precision Mean_Recall
##
    0.8312189
                         0.9320476
                                              0.8312189
                                                              0.75375
```

```
##
     Mean_Detection_Rate Mean_Balanced_Accuracy
##
     0.1934067
                          0.8389062
##
## Tuning parameter 'C' was held constant at a value of 1
print(nn_fit)
## Neural Network
##
## 80 samples
## 30 predictors
  4 classes: 'Basal.like', 'HER2.enriched', 'Luminal.A', 'Luminal.B'
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 10 times)
## Summary of sample sizes: 73, 72, 73, 72, 71, 71, ...
## Resampling results across tuning parameters:
##
##
     size decay Accuracy
                                                   Mean Sensitivity
                             Kappa
                                        Mean F1
##
           0e+00
                  0.4443770 0.2324942
                                              NaN
                                                   0.4108333
##
           1e-04 0.4661508 0.2619683
                                       0.6958333
                                                   0.4291667
##
           1e-01 0.4891270 0.2907450
                                              NaN 0.4345833
     1
##
     3
           0e+00
                  0.6697738 0.5484242
                                        0.8102799
                                                   0.6450000
##
     3
           1e-04 0.6868968 0.5741659 0.8183970
                                                   0.6725000
##
     3
           1e-01 0.7526468 0.6616359
                                       0.8332393 0.7362500
##
     5
           0e+00 0.7228413 0.6208831 0.8096024
                                                   0.7033333
##
     5
           1e-04 0.7325635
                             0.6342318
                                        0.8218778
                                                   0.7183333
##
           1e-01 0.7535198 0.6617535 0.8208433 0.7362500
##
     Mean_Specificity Mean_Pos_Pred_Value Mean_Neg_Pred_Value Mean_Precision
##
     0.8087054
                       0.2343750
                                            0.8414683
                                                                  0.2343750
##
     0.8159762
                       0.5156250
                                            0.8470428
                                                                  0.5156250
##
     0.8236101
                                            0.8504375
                             NaN
                                                                        NaN
##
     0.8890833
                       0.6986111
                                                                  0.6986111
                                            0.8997192
##
                                                                  0.7744253
     0.8957649
                       0.7744253
                                            0.9047500
##
     0.9169792
                       0.8092857
                                            0.9243720
                                                                  0.8092857
##
     0.9068125
                       0.7873134
                                            0.9150595
                                                                  0.7873134
##
     0.9099464
                       0.8121094
                                            0.9185050
                                                                  0.8121094
##
                                                                  0.8301329
     0.9163363
                       0.8301329
                                            0.9250337
##
    Mean_Recall Mean_Detection_Rate Mean_Balanced_Accuracy
                  0.1110942
##
     0.4108333
                                       0.6097693
##
     0.4291667
                  0.1165377
                                       0.6225714
##
     0.4345833
                  0.1222817
                                       0.6290967
##
     0.6450000
                  0.1674435
                                       0.7670417
##
     0.6725000
                  0.1717242
                                       0.7841324
##
                  0.1881617
     0.7362500
                                       0.8266146
##
     0.7033333
                  0.1807103
                                       0.8050729
##
     0.7183333
                  0.1831409
                                       0.8141399
##
     0.7362500
                  0.1883800
                                       0.8262932
##
## Accuracy was used to select the optimal model using the largest value.
```

The final values used for the model were size = 5 and decay = 0.1.

#print(nb_fit)2

print(rf_fit)

```
## Random Forest
##
## 80 samples
## 30 predictors
    4 classes: 'Basal.like', 'HER2.enriched', 'Luminal.A', 'Luminal.B'
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 10 times)
## Summary of sample sizes: 72, 72, 73, 71, 72, 72, ...
## Resampling results across tuning parameters:
##
##
     mtry
           Accuracy
                      Kappa
                                  Mean_F1
                                              Mean_Sensitivity
                                                                Mean_Specificity
##
      2
           0.7795238
                      0.6920946
                                  0.8228571
                                             0.7362500
                                                                0.9227470
##
     16
           0.7430873
                      0.6407666
                                  0.8059382
                                             0.6979167
                                                                0.9101488
##
           0.7337460 \quad 0.6283982 \quad 0.7890110 \quad 0.6891667
                                                                0.9067500
##
     Mean_Pos_Pred_Value Mean_Neg_Pred_Value Mean_Precision Mean_Recall
     0.8804167
                           0.9350367
                                                                 0.7362500
##
                                                 0.8804167
##
     0.8441123
                           0.9238899
                                                 0.8441123
                                                                 0.6979167
##
     0.8331349
                           0.9207579
                                                 0.8331349
                                                                 0.6891667
##
     Mean_Detection_Rate Mean_Balanced_Accuracy
##
     0.1948810
                           0.8294985
##
     0.1857718
                           0.8040327
##
     0.1834365
                           0.7979583
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was mtry = 2.
```

Before, K-fold cross validation, The SVM model achieved a accuracy of 76% The neural networks model achieved accuracy of 71% The naive bayes model achieved a accuracy of 76%

k-Fold Cross Validation is done for the whole dataset. I have used k = 10 which means 10 folds take place along with 10 repetitions. For testing the data, I have used 3 models to test the k-fold CV

(10 fold, repeated 10 times)

1. SVM:

```
Accuracy - 0.773627 Kappa - 0.6904643 Mean_F1 - 0.8352319 Mean_Sensitivity - 0.75375 Mean_Specificity - 0.9240625 Mean_Precision - 0.8312189 Mean_Recall - 0.75375
```

There's no much change after applying k-fold cross validation

2. Neural Networks:

```
Accuracy - 0.7786581 Kappa - 0.6980310 Mean_F1 - 0.7569600 Mean_Sensitivity - 0.7568452 Mean_Specificity - 0.9250575 Mean_Precision - 0.7997022 Mean_Recall - 0.7568452
```

- 3. Naive baye's: Accuracy 0.7535198 Kappa 0.6617535 Mean_F1 - 0.8208433 Mean_Sensitivity - 0.7362500 Mean_Specificity - 0.8410706 Mean_Precision - 0.9163363 Mean_Recall - 0.7362500
- 4. Random Forests:

```
Accuracy - 0.7795238 Kappa - 0.6920946 Mean_F1 - 0.8228571 Mean_Sensitivity - 0.7362500 Mean_Specificity - 0.9227470 Mean_Precision - 0.8331349 Mean Recall - 0.6891667
```

HYPERPARAMETER TUNING

1. SVM

Specificity

The tuning and training of an SVM with a linear kernel is demonstrated in the code below, which also controls crossvalidation for tuning the hyperparameter C.

```
set.seed(10)
#Configuring train control for cross validation and hyperparameter calibration
train_control <- trainControl(method="repeatedcv", number=10, repeats=10, savePredictions = TRUE, summa
#Tunegrid for various C values
grid \leftarrow expand.grid(C = seq(0.000001, 0.15, 0.002))
set.seed(10)
svm.lin.mod <- train(PAM50.mRNA ~ ., data=data_norm[samp,], trControl=train_control, method="svmLinear"</pre>
svm.predicts <- predict(svm.lin.mod, newdata = data_norm[-samp,])</pre>
confusionMatrix(svm.predicts, factor(data_norm$PAM50.mRNA[-samp]), mode = "everything")
## Confusion Matrix and Statistics
##
                  Reference
##
## Prediction
                   Basal.like HER2.enriched Luminal.A Luminal.B
    Basal.like
                                           2
                                                      0
##
                             5
                                                      0
##
     HER2.enriched
                             0
                                                                 0
                                           1
##
    Luminal.A
                             0
                                           0
                                                      5
                                                                 1
    Luminal.B
                             0
                                           Λ
                                                      1
                                                                 6
##
## Overall Statistics
##
##
                  Accuracy : 0.8095
                    95% CI: (0.5809, 0.9455)
##
       No Information Rate: 0.3333
##
       P-Value [Acc > NIR] : 1.026e-05
##
##
##
                     Kappa: 0.7358
##
   Mcnemar's Test P-Value : NA
##
##
## Statistics by Class:
##
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                    1.0000
                                                         0.33333
```

1.00000

0.9333

0.8750

```
## Pos Pred Value
                                   0.7143
                                                         1.00000
                                                                           0.8333
## Neg Pred Value
                                   1.0000
                                                        0.90000
                                                                           0.9333
## Precision
                                   0.7143
                                                        1.00000
                                                                           0.8333
## Recall
                                    1.0000
                                                        0.33333
                                                                           0.8333
## F1
                                    0.8333
                                                        0.50000
                                                                           0.8333
## Prevalence
                                    0.2381
                                                        0.14286
                                                                           0.2857
## Detection Rate
                                    0.2381
                                                                           0.2381
                                                        0.04762
## Detection Prevalence
                                   0.3333
                                                        0.04762
                                                                           0.2857
## Balanced Accuracy
                                    0.9375
                                                        0.66667
                                                                           0.8833
                        Class: Luminal.B
##
## Sensitivity
                                   0.8571
                                   0.9286
## Specificity
## Pos Pred Value
                                   0.8571
## Neg Pred Value
                                   0.9286
## Precision
                                   0.8571
## Recall
                                   0.8571
## F1
                                   0.8571
## Prevalence
                                   0.3333
## Detection Rate
                                   0.2857
## Detection Prevalence
                                   0.3333
## Balanced Accuracy
                                   0.8929
```

The SVM model achieved accuracy of 80% after hyper-parameter tuning. Therefore, no much change after tuning

NN

```
set.seed(5000)
cctrlR <- trainControl(method = "repeatedcv", number=10, repeats=10, returnResamp = "all", search = "ra</pre>
nn <- train(PAM50.mRNA ~., data= data_norm[samp,],</pre>
               method = "nnet",
               trControl = cctrlR,
               preProc = c("center", "scale"),
               trace = FALSE)
nnpred <- predict(nn, newdata= data_norm[-samp,])</pre>
#viewing confusion matrix
confusionMatrix(nnpred, factor(data_norm$PAM50.mRNA[-samp]), mode = "everything")
## Confusion Matrix and Statistics
##
##
                  Reference
                   Basal.like HER2.enriched Luminal.A Luminal.B
## Prediction
    Basal.like
                                                      0
##
                                            1
##
    HER2.enriched
                             0
                                            1
                                                      0
                                                                 0
    Luminal.A
                             0
                                            0
                                                      6
                                                                 2
##
##
    Luminal.B
                             0
                                            1
                                                      0
                                                                 5
##
## Overall Statistics
##
##
                  Accuracy: 0.8095
```

```
##
                    95% CI: (0.5809, 0.9455)
##
       No Information Rate: 0.3333
       P-Value [Acc > NIR] : 1.026e-05
##
##
##
                     Kappa: 0.7358
##
   Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##
                        Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                    1.0000
                                                        0.33333
                                                                           1.0000
## Specificity
                                    0.9375
                                                         1.00000
                                                                           0.8667
## Pos Pred Value
                                    0.8333
                                                         1.00000
                                                                           0.7500
## Neg Pred Value
                                    1.0000
                                                        0.90000
                                                                           1.0000
## Precision
                                    0.8333
                                                         1.00000
                                                                           0.7500
## Recall
                                   1.0000
                                                        0.33333
                                                                           1.0000
## F1
                                    0.9091
                                                        0.50000
                                                                           0.8571
## Prevalence
                                   0.2381
                                                        0.14286
                                                                           0.2857
## Detection Rate
                                    0.2381
                                                        0.04762
                                                                           0.2857
## Detection Prevalence
                                   0.2857
                                                        0.04762
                                                                           0.3810
## Balanced Accuracy
                                    0.9688
                                                        0.66667
                                                                           0.9333
##
                        Class: Luminal.B
## Sensitivity
                                  0.7143
## Specificity
                                  0.9286
## Pos Pred Value
                                  0.8333
## Neg Pred Value
                                   0.8667
## Precision
                                   0.8333
## Recall
                                   0.7143
## F1
                                   0.7692
## Prevalence
                                   0.3333
## Detection Rate
                                   0.2381
## Detection Prevalence
                                   0.2857
## Balanced Accuracy
                                   0.8214
```

The model achieved accuracy of 80.1% after hyper parameter tuning

```
## Confusion Matrix and Statistics
##
##
                  Reference
## Prediction
                   Basal.like HER2.enriched Luminal.A Luminal.B
##
    Basal.like
                            5
                                          3
                                                     0
                                                               1
##
     HER2.enriched
                            0
                                           0
                                                     0
                                                               0
    Luminal.A
                                                     5
```

```
##
## Overall Statistics
##
##
                  Accuracy: 0.7143
##
                    95% CI: (0.4782, 0.8872)
##
       No Information Rate: 0.3333
       P-Value [Acc > NIR] : 0.0004045
##
##
##
                     Kappa: 0.6038
##
##
    Mcnemar's Test P-Value : NA
## Statistics by Class:
##
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
                                    1,0000
                                                          0.0000
## Sensitivity
                                                                            0.8333
## Specificity
                                    0.7500
                                                          1.0000
                                                                            0.9333
## Pos Pred Value
                                    0.5556
                                                                            0.8333
                                                             NaN
## Neg Pred Value
                                    1.0000
                                                          0.8571
                                                                            0.9333
## Precision
                                    0.5556
                                                              NΑ
                                                                            0.8333
## Recall
                                    1.0000
                                                          0.0000
                                                                            0.8333
## F1
                                    0.7143
                                                                            0.8333
                                                              NA
## Prevalence
                                    0.2381
                                                          0.1429
                                                                            0.2857
## Detection Rate
                                    0.2381
                                                          0.0000
                                                                            0.2381
## Detection Prevalence
                                    0.4286
                                                          0.0000
                                                                            0.2857
## Balanced Accuracy
                                    0.8750
                                                          0.5000
                                                                            0.8833
                         Class: Luminal.B
## Sensitivity
                                   0.7143
## Specificity
                                   0.9286
## Pos Pred Value
                                   0.8333
## Neg Pred Value
                                   0.8667
## Precision
                                   0.8333
## Recall
                                   0.7143
## F1
                                   0.7692
## Prevalence
                                   0.3333
## Detection Rate
                                   0.2381
## Detection Prevalence
                                   0.2857
## Balanced Accuracy
                                   0.8214
The model achieved 71.4% accuracy after hyper parameter tuning
rb <- train(`PAM50.mRNA`~., data = data_norm[samp,], method = "rf",
               trControl = trainControl(method = "repeatedcv", number=10, repeats=10))
rbps <- predict(rb, newdata=data_norm[-samp,])</pre>
#viewing confusion matrix
confusionMatrix(rbps, factor(data_norm$PAM50.mRNA[-samp]), mode = "everything")
## Confusion Matrix and Statistics
##
##
                  Reference
                   Basal.like HER2.enriched Luminal.A Luminal.B
## Prediction
     Basal.like
                                           1
                                                      0
```

0

1

5

##

Luminal.B

```
##
     HER2.enriched
                                                       0
                                                                  0
##
                                            0
                                                       5
                                                                  2
     Luminal.A
                             1
     Luminal.B
##
                                            1
                                                       1
                                                                  4
##
##
  Overall Statistics
##
                   Accuracy : 0.6667
##
                     95% CI: (0.4303, 0.8541)
##
##
       No Information Rate: 0.3333
##
       P-Value [Acc > NIR] : 0.001827
##
                      Kappa: 0.5377
##
##
##
    Mcnemar's Test P-Value : NA
##
  Statistics by Class:
##
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
                                     0.8000
                                                          0.33333
## Sensitivity
                                                                             0.8333
## Specificity
                                     0.8750
                                                          1.00000
                                                                             0.8000
## Pos Pred Value
                                     0.6667
                                                          1.00000
                                                                             0.6250
## Neg Pred Value
                                                          0.90000
                                     0.9333
                                                                             0.9231
## Precision
                                     0.6667
                                                                             0.6250
                                                          1.00000
## Recall
                                     0.8000
                                                          0.33333
                                                                             0.8333
## F1
                                     0.7273
                                                          0.50000
                                                                             0.7143
## Prevalence
                                     0.2381
                                                          0.14286
                                                                             0.2857
## Detection Rate
                                     0.1905
                                                          0.04762
                                                                             0.2381
## Detection Prevalence
                                     0.2857
                                                          0.04762
                                                                             0.3810
                                                          0.66667
## Balanced Accuracy
                                     0.8375
                                                                             0.8167
##
                         Class: Luminal.B
## Sensitivity
                                    0.5714
## Specificity
                                    0.8571
## Pos Pred Value
                                    0.6667
## Neg Pred Value
                                    0.8000
## Precision
                                    0.6667
## Recall
                                    0.5714
## F1
                                    0.6154
## Prevalence
                                    0.3333
## Detection Rate
                                    0.1905
## Detection Prevalence
                                    0.2857
## Balanced Accuracy
                                    0.7143
```

The model achieved 66.6% accuracy after hyper parameter tuning

The SVM model achieved a accuracy of 80% and the accuracy remained same after hyper-parameter tuning. The neural networks model achieved accuracy of 80% and the accuracy remained the same after hyper-parameter tuning. The naive bayes model achieved a accuracy of 76% and the accuracy decreased to 71.4% after hyper-parameter tuning. The Random Forest model achieved a accuracy of 61% and the accuracy increased to 66.6% after hyper-parameter tuning.

Comparison of models after tuning

• It is observed by comparing Accuracy that SVM and NNmodel performs the best amongst the others both before tuning and after tuning.

Compairing Model Accuracies:

SVM on original data set values: Accuracy: 0.809, Kappa: 0.7399 Neural Networks on original data set values: Accuracy: 0.809, Kappa: 0.7358 Naive bayes on original data set values: Accuracy: 0.7619, Kappa: 0.6729 Random Forest on original data set values: Accuracy: 0.619, Kappa: 0.4734

SVM with k-fold cross validation: Accuracy: 0.7723365, Kappa: 0.6914544 Neural Networks with k-fold cross validation: Accuracy: 0.7786581, Kappa: 0.6980310 Naive bayes with k-fold cross validation: Accuracy: 0.7535198, Kappa: 0.6617535 Random Forest with k-fold cross validation: Accuracy: 0.7795238, Kappa: 0.6920946

SVM on original data set values after hyper parameter tuning: Accuracy: 0.8095, Kappa: 0.7358 Neural Networks on original data set values after hyper parameter tuning: 0.8095, Kappa: 0.7358 Naive bayes on original data set values after hyper parameter tuning: Accuracy: 0.7143, Kappa: 0.6038 Random Forest on original data set values after hyper parameter tuning: Accuracy: 0.6667, Kappa: 0.5377

-From the above, we can say that both SVM and Neural Network model has done a good job predicting the cancer subtype

Deployment

- (i) bagging: use of bagging with homogeneous learners
- (ii) Stacked ensemble using SVM, Neural Networks, Naive Bayes Majority voting
- (iii) boosting: Extreme Gradient Boosting

Bagging

use of bagging with homogeneous learners

```
set.seed(1000)
#fit the bagged model
bag <- bagging(formula = PAM50.mRNA ~ .,data = train,nbagg = 150,</pre>
                                                                      coob = TRUE,control = rpart.control
#display fitted bagged model
bag
##
## Bagging classification trees with 150 bootstrap replications
##
## Call: bagging.data.frame(formula = PAM50.mRNA ~ ., data = train, nbagg = 150,
       coob = TRUE, control = rpart.control(minsplit = 2, cp = 0))
##
##
## Out-of-bag estimate of misclassification error: 0.2542
bag pred <- predict(bag, test)</pre>
confusionMatrix(test$`PAM50.mRNA`,bag_pred)
## Confusion Matrix and Statistics
##
##
                  Reference
## Prediction
                   Basal.like HER2.enriched Luminal.A Luminal.B
    Basal.like
                                                      0
    HER2.enriched
                                            1
                                                      0
                             1
                                                                1
##
```

```
##
     Luminal.A
     Luminal.B
                                                      2
##
##
## Overall Statistics
##
##
                  Accuracy: 0.619
                     95% CI: (0.3844, 0.8189)
##
       No Information Rate: 0.381
##
##
       P-Value [Acc > NIR] : 0.02313
##
##
                      Kappa: 0.475
##
##
    Mcnemar's Test P-Value : NA
##
##
  Statistics by Class:
##
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                     0.5000
                                                          1.00000
                                                                             0.6667
                                     0.9231
                                                          0.90000
                                                                             0.8667
## Specificity
## Pos Pred Value
                                     0.8000
                                                          0.33333
                                                                             0.6667
## Neg Pred Value
                                     0.7500
                                                          1.00000
                                                                             0.8667
## Prevalence
                                     0.3810
                                                          0.04762
                                                                             0.2857
## Detection Rate
                                     0.1905
                                                          0.04762
                                                                             0.1905
## Detection Prevalence
                                     0.2381
                                                          0.14286
                                                                             0.2857
## Balanced Accuracy
                                     0.7115
                                                          0.95000
                                                                             0.7667
                         Class: Luminal.B
## Sensitivity
                                   0.6667
                                   0.8000
## Specificity
## Pos Pred Value
                                   0.5714
## Neg Pred Value
                                   0.8571
## Prevalence
                                   0.2857
## Detection Rate
                                   0.1905
## Detection Prevalence
                                   0.3333
## Balanced Accuracy
                                   0.7333
```

Accuracy achieved using bagging is 0.619

Stacking

The breast cancer sub-type classifier will be deployed as a model ensemble. A model ensemble is a prediction model that is an aggregate of a set of models. Specifically, a model ensemble aggregates predictions across all the individual models in the ensemble using a voting mechanism. In general, we expect that a collection of independent models would perform better than any individual model. The voting mechanism that will be used for the breast cancer ensemble is the mode prediction for an patient across the three models.

Define a function to calculate the mode across values.

```
Mode <- function(x) {
  ux <- unique(x)
  ux[which.max(tabulate(match(x, ux)))]
}</pre>
```

Next, define a function to loop through the observations in the test data and generate the modal prediction for each observation across the three classifiers.

```
vote <- function (p1, p2, p3) {</pre>
   m <- length(p1) # number of predictions in the test data
   ds <- numeric(m) # creates numeric vector to hold final prediction
   # loops through predictions in the test data
   for (i in 1:m) {
       # calculate mode prediction for an obs across classifiers
       p <- c(p1[i],p2[i],p3[i])
       # store modal prediction in return vector
       ds[i] <- Mode(p)</pre>
   }
    # return vector
   return(ds)
}
Use functions to generate the model ensemble.
           <- vote(p1 = svm_pred, p2 = nnpred_model2, p3 = nbpred_model3)</pre>
ens_pred
ens_pred
  [1] 1 1 1 1 1 2 2 4 3 3 3 3 3 3 4 1 4 4 4 3 4
# Factor
ens_pred[which(ens_pred == 1)] = "Basal.like"
ens_pred[which(ens_pred == 2)] = "HER2.enriched"
ens_pred[which(ens_pred == 3)] = "Luminal.A"
ens_pred[which(ens_pred == 4)] = "Luminal.B"
label <- factor(test$PAM50.mRNA)</pre>
label
## [1] Basal.like
                                                               Basal.like
                     Basal.like
                                   Basal.like
                                                 Basal.like
## [6] HER2.enriched HER2.enriched Luminal.A
                                                               Luminal.A
## [11] Luminal.A Luminal.A
                                   Luminal.A
                                                 Luminal.A
                                                               Luminal.B
## [16] Luminal.B
                                                               Luminal.B
                     Luminal.B
                                   Luminal.B
                                                 Luminal.B
## [21] Luminal.B
## Levels: Basal.like HER2.enriched Luminal.A Luminal.B
confusionMatrix(factor(ens_pred), label , mode = "everything")
## Confusion Matrix and Statistics
##
                 Reference
##
## Prediction
                  Basal.like HER2.enriched Luminal.A Luminal.B
    Basal.like
                                                  0
##
                           5
                                         Ω
##
    HER2.enriched
                           0
                                         2
                                                  0
                                                             0
                           0
                                        0
                                                  6
##
    Luminal.A
                                                             1
##
    Luminal.B
                           0
                                         1
                                                 0
##
```

```
## Overall Statistics
##
                  Accuracy : 0.8571
##
##
                    95% CI: (0.6366, 0.9695)
##
       No Information Rate: 0.3333
       P-Value [Acc > NIR] : 1.102e-06
##
##
##
                     Kappa: 0.8037
##
   Mcnemar's Test P-Value : NA
##
##
## Statistics by Class:
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
##
## Sensitivity
                                    1.0000
                                                         0.66667
                                                                            1.0000
## Specificity
                                    0.9375
                                                         1.00000
                                                                            0.9333
## Pos Pred Value
                                    0.8333
                                                         1.00000
                                                                            0.8571
## Neg Pred Value
                                    1.0000
                                                         0.94737
                                                                            1.0000
## Precision
                                    0.8333
                                                         1.00000
                                                                            0.8571
## Recall
                                    1.0000
                                                         0.66667
                                                                            1.0000
## F1
                                    0.9091
                                                         0.80000
                                                                            0.9231
## Prevalence
                                    0.2381
                                                         0.14286
                                                                            0.2857
## Detection Rate
                                    0.2381
                                                                            0.2857
                                                         0.09524
## Detection Prevalence
                                    0.2857
                                                         0.09524
                                                                            0.3333
## Balanced Accuracy
                                    0.9688
                                                         0.83333
                                                                            0.9667
                         Class: Luminal.B
## Sensitivity
                                   0.7143
## Specificity
                                   0.9286
## Pos Pred Value
                                   0.8333
## Neg Pred Value
                                   0.8667
## Precision
                                   0.8333
## Recall
                                   0.7143
## F1
                                   0.7692
## Prevalence
                                   0.3333
## Detection Rate
                                   0.2381
## Detection Prevalence
                                   0.2857
## Balanced Accuracy
                                   0.8214
mv_auc <- multiclass.roc(label, as.ordered(ens_pred))</pre>
auc(mv_auc)
```

```
## Multi-class area under the curve: 0.8452
```

Macro-averaged Metrics : The per-class metrics can be averaged over all the classes resulting in macro-averaged precision, recall and F-1.

```
# macro-averaged precision
precision_stack<- c(0.8333,1.00000,0.8571,0.8333)
macro_precision_stack <- mean(precision_stack)
# macro-averaged recall
recall_stack <- c(1.0000,0.66667,1.0000,0.7143)
macro_recall_stack<- mean(recall_stack)
# macro-averaged F-1</pre>
```

```
F1_stack<- c(0.9091,0.80000,0.9231,0.7692)
macroF1_stack <- mean(F1_stack)</pre>
macro_average_stack <-data.frame(macro_precision_stack, macro_recall_stack, macroF1_stack)
macro_average_stack
##
     macro_precision_stack macro_recall_stack macroF1_stack
## 1
                  0.880925
                                     0.8452425
Name_metrics <- c("Accuracy", "Precision", "Recall", "F-1", "AUC", "Kappa")
values_stack <- c(0.8571, 0.880925, 0.8452425, 0.85035, 0.8452, 0.8037)</pre>
metrics stack <- data.frame(Name metrics, values stack)</pre>
print (metrics_stack)
    Name_metrics values_stack
##
## 1
        Accuracy
                     0.8571000
        Precision
## 2
                     0.8809250
## 3
           Recall
                     0.8452425
## 4
              F-1
                     0.8503500
## 5
              AUC
                     0.8452000
## 6
                     0.8037000
            Kappa
```

comparison of ensemble to individual models

Of the four models (SVM, NN, NB, RF)- The SVM and NN works better and both the models have similar Accuracy, Precision, Recall, F-1, AUC and Kappa Values

Now, comparing Accuracy, Precision, Recall, F-1, AUC and Kappa values with that of stacked ensemble model (SVM, NN, NB)

I have saved the Accuracy, Precision, Recall, F-1, AUC and Kappa values under metric_model

```
metrics_svm
```

```
## Name_metrics values_svm
## 1 Accuracy 0.8095000
## 2 Precision 0.8428500
## 3 Recall 0.8095175
## 4 F-1 0.8057750
## 5 AUC 0.7897000
## 6 Kappa 0.7399000
```

```
metrics_nn
```

```
## Name_metrics values_nn
## 1 Accuracy 0.8095000
## 2 Precision 0.8541500
## 3 Recall 0.7619075
## 4 F-1 0.7588500
## 5 AUC 0.8571000
## 6 Kappa 0.7358000
```

```
##
     Name_metrics values_stack
## 1
         Accuracy
                      0.8571000
## 2
        Precision
                      0.8809250
## 3
           Recall
                      0.8452425
              F-1
## 4
                      0.8503500
## 5
               AUC
                      0.8452000
## 6
            Kappa
                      0.8037000
```

By comparing above values, we can say that stacked ensemble model worked better than other individual models with the highest accuracy of 85.7% and Precision-0.8809250, Recall-0.8452425, F-1-0.8503500, AUC - 0.8452000, Kappa - 0.8037000

Therefore, These values are higher than individual models.

Conclusion: A model ensemble is a prediction model that is an aggregate of a set of models. Specifically, a model ensemble aggregates predictions across all the individual models (SVM, NN, NB) in the ensemble using a voting mechanism. And this model has high accuracy of predicting the subtype of cancer than individual models.

Boosting

Extreme gradient boosting

The term "gradient boosting" comes from the idea of "boosting" or improving a single weak model by combining it with a number of other weak models in order to generate a collectively strong model. Gradient boosting is an extension of boosting where the process of additively generating weak models is formalized as a gradient descent algorithm over an objective function. Gradient boosting sets targeted outcomes for the next model in an effort to minimize errors. Targeted outcomes for each case are based on the gradient of the error (hence the name gradient boosting) with respect to the prediction.

XGBoost (eXtreme Gradient Boosting) is a machine learning classifier/predictor, which produces a model in a form of an ensemble of weak prediction models. XGBoost helps to reduce overfitting.

```
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c api/c api.cc:935: 'ntree limit' is deprecated, use 'iteration range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c api/c api.cc:935: 'ntree limit' is deprecated, use 'iteration range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
```

```
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
## [20:50:44] WARNING: src/c_api/c_api.cc:935: 'ntree_limit' is deprecated, use 'iteration_range' inste
bpred_xgb <- predict(xgb, newdata=testDF)</pre>
confusionMatrix(testDF$\text{PAM50.mRNA}\text{,bpred_xgb})
## Confusion Matrix and Statistics
##
##
                  Reference
## Prediction
                    Basal.like HER2.enriched Luminal.A Luminal.B
##
                                            0
                                                      2
                                                                 0
     Basal.like
                             3
##
     HER2.enriched
                             1
                                            2
                                                      0
                                                                 0
                                            0
##
     Luminal.A
                             0
                                                      5
                                                                 1
##
     Luminal.B
                                                      1
                                                                 6
##
## Overall Statistics
##
##
                  Accuracy : 0.7619
##
                    95% CI: (0.5283, 0.9178)
##
       No Information Rate: 0.381
       P-Value [Acc > NIR] : 0.0004398
##
##
##
                      Kappa: 0.6698
##
##
    Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##
                         Class: Basal.like Class: HER2.enriched Class: Luminal.A
## Sensitivity
                                    0.7500
                                                         1.00000
                                                                            0.6250
## Specificity
                                    0.8824
                                                         0.94737
                                                                            0.9231
## Pos Pred Value
                                    0.6000
                                                         0.66667
                                                                            0.8333
## Neg Pred Value
                                    0.9375
                                                         1.00000
                                                                            0.8000
## Prevalence
                                    0.1905
                                                         0.09524
                                                                            0.3810
## Detection Rate
                                    0.1429
                                                         0.09524
                                                                            0.2381
## Detection Prevalence
                                    0.2381
                                                         0.14286
                                                                            0.2857
## Balanced Accuracy
                                    0.8162
                                                         0.97368
                                                                            0.7740
##
                         Class: Luminal.B
## Sensitivity
                                   0.8571
## Specificity
                                   0.9286
## Pos Pred Value
                                   0.8571
## Neg Pred Value
                                   0.9286
## Prevalence
                                   0.3333
## Detection Rate
                                   0.2857
## Detection Prevalence
                                   0.3333
## Balanced Accuracy
                                   0.8929
```

I have also built extreme gradient boosting model. This model has an accuracy of 61.9%

Therefore, Stacked ensemble stacked ensemble (SVM, NN, NB) model works better than extereme gradient boosting model

Conclusion

It's really intriguing to me that ML techniques may be used to identify a group of predictor proteins that outperform proteins known to be linked to the genetic test that determines the classification in terms of identifying cancer subtypes.

The lasso-selected variables consistently outperformed the PAM50 ones, but no other methods produced classification results that were more accurate than the SVM NN and stacked ensemble (SVM, NN, NB).