ECOG314 -- Suggested project guidelines -- A worked example using Wine dataset from UCI data repository

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Demo on how to analyze a multivariate dataset from UCI ML data repository

Reference materials

- 1. This R Data Import Tutorial Is Everything You Need -- R bloggers
- 2. Importing Data Into R Part Two -- R bloggers

Data repositories

- Financial Accounts of the United States
- DATA.GOV
- UCI Data Repository -- Machine learning datasets

Financial Data repositories

- Federal Reserve -- few clicks required financial_accounts_of_the_us.png
- Data Market one of the user friendly Websites -- few clicks required
- INFORUM -- requires Inforum's database and regression package, G, to access that data

Big Data repositories

- Data Science Central
- Big data made simple
- Amazon Web Service (AWS)
- caesar0301/awesome-public-datasets

Steps:

- 1. Visit the UCI wine data repository for a detailed description, the data dictionary and maintainer of the data
 - results of a chemical analysis of the qualities of wines grown in the same region in Italy but derived from 3 different cultivars (a plant variety that has been produced in cultivation by selective breeding)
 - 14 columns (dimensions, attributes)
 - with 1st column containing the cultivar of a wine sample (labelled 1, 2 or 3) next 13 columns contain the concentrations of the 13 different chemicals in that sample

- 2. Setup your project workspace and cd into that direction (Session -> Set Working Directory -> Choose Directory (CNTRL+Shift+h)
- 3. Read dataset into R read.table(...)
 I always get a segment of the file first (for very large files) read.table("my_some_datatable_name.txt" header = TRUE, nrows = 25) # easier to work with smaller dataset

Step 1 of N: Point browser to the UCI wine dataset repository

• UCI ML wine dataset repository

Step 2 of N: Clear workspace, and set and cd into working directory

R-CODE

Step 3 of N: Read the multivariate dataset into R

Data structure

- Data consists of 1 row per of record (observation) wine sample.
- First column, the cultivar of a wine sample is labelled 1, 2 or 3
- Next 12 columns contain the concentrations of the 13 different chemicals in that sample.
- Columns are separated by commas.

Segments from the wine dataset:

Reading the data

- We use read.table() function to read in the data into R
- We use sep="," argument in read.table() to specify that columns are separated by commas.

R-CODE

R-CODE

```
#--
# check dimension of the data you have read -- data sanity check 1

dim(wine)
Output: [1] 25 14
```

R-CODE

```
class(wine)
Output: [1] "data.frame"
```

```
head(wine, n=5)

Output:

X1 X14.23 X1.71 X2.43 X15.6 X127 X2.8 X3.06 X.28 X2.29 X5.64 X1.04 X3.92 X1065
1 1 13.20 1.78 2.14 11.2 100 2.65 2.76 0.26 1.28 4.38 1.05 3.40 1050
2 1 13.16 2.36 2.67 18.6 101 2.80 3.24 0.30 2.81 5.68 1.03 3.17 1185
3 1 14.37 1.95 2.50 16.8 113 3.85 3.49 0.24 2.18 7.80 0.86 3.45 1480
4 1 13.24 2.59 2.87 21.0 118 2.80 2.69 0.39 1.82 4.32 1.04 2.93 735
5 1 14.20 1.76 2.45 15.2 112 3.27 3.39 0.34 1.97 6.75 1.05 2.85 1450
```

BIG DATA files, import and extract a random sample into R, then offload master file onto hard drive to free up memory

- Use read.table(...) function to read in the data into R
- Use sample(...) function to get a random sample from the raw data
- Use saveRDS(...) and rm(...) functions to saVE file to hard drive and remove BIG DATA file to free up memory
- For Ubuntu and Linux users, this can be done at the Operating System (OS) level shuf -n 10 > small_wine_data.txt
 Use nl wine_data.txt | shuf -n 10 to verify that the lines in file has been shuffled (randomized)

R-CODE

Output:

```
## Rows selected at random: 175, 24, 97, 51, 5, 121, 104, 15, 149, 18, 78, 108, 146, 161, 111, 54, 26, 64, 166, 147, 58, 154, 49, 98, 60
```

```
# Read the data from source
                                  # we know from above that the class of ob
big wine.df <- read.table(</pre>
ject returned by read.table is a data.frame (.df extension)
        file = "http://archive.ics.uci.edu/ml/machine-learning-databases/win
e/wine.data", # filename at the UCI repository
                                  # file has a header
      header = TRUE,
          sep = ","
                                  # rows separated by commas. NOTE: We to
ok out the comma after at the end of the line
                                 # we are ready every line in the file
      # nrow = mysample
# column names
colnames(big_wine.df) <- paste("v", 1:dim(big_wine.df)[2], sep="")</pre>
                                                                    # v1.
v2, ....v13, v14
head(big wine.df)
```

Output:

```
      v1
      v2
      v3
      v4
      v5
      v6
      v7
      v8
      v9
      v10
      v11
      v12
      v13
      v14

      1
      1
      13.20
      1.78
      2.14
      11.2
      100
      2.65
      2.76
      0.26
      1.28
      4.38
      1.05
      3.40
      1050

      2
      1
      13.16
      2.36
      2.67
      18.6
      101
      2.80
      3.24
      0.30
      2.81
      5.68
      1.03
      3.17
      1185

      3
      1
      14.37
      1.95
      2.50
      16.8
      113
      3.85
      3.49
      0.24
      2.18
      7.80
      0.86
      3.45
      1480

      4
      1
      13.24
      2.59
      2.87
      21.0
      118
      2.80
      2.69
      0.39
      1.82
      4.32
      1.04
      2.93
      735

      5
      1
      14.20
      1.76
      2.45
      15.2
      112
      3.27
      3.39
      0.34
      1.97
      6.75
      1.05
      2.85
      1450

      6
      1
      14.39
      1.87
      2.45
      14.6
      96
      2.50
      2.52
      0.
```

R-CODE

check dimension of the data you have read -- data sanity check 1
dim(big_wine.df)

Output: [1] 177 14

R-CODE

small_wine.df <- big_wine.df[n,] #every row specified by n. That is 86, 1 53, 78, 43, 13, 18 this changes if you do not specify a seed

dim(small_wine.df)

Output: [1] 25 14

R-CODE

small_wine.df # see that indeed we are not using the entire dataset

Output:

```
v5 v6
                              v7 v8
                                          v9 v10
                                                    v11 v12 v13
              v3
                   v4
175 3 13.27 4.28 2.26 20.0 120 1.59 0.69 0.43 1.35 10.20 0.59 1.56
                                                                  835
    1 13.50 1.81 2.61 20.0 96 2.53 2.61 0.28 1.66 3.52 1.12 3.82
                                                                  845
    2 12.29 1.41 1.98 16.0 85 2.55 2.50 0.29 1.77 2.90 1.23 2.74 428
97
    1 13.83 1.65 2.60 17.2 94 2.45 2.99 0.22 2.29 5.60 1.24 3.37 1265
    1 14.20 1.76 2.45 15.2 112 3.27 3.39 0.34 1.97
                                                   6.75 1.05 2.85 1450
121 2 11.56 2.05 3.23 28.5 119 3.18 5.08 0.47 1.87 6.00 0.93 3.69
104 2 12.51 1.73 1.98 20.5 85 2.20 1.92 0.32 1.48 2.94 1.04 3.57 672
    1 13.63 1.81 2.70 17.2 112 2.85 2.91 0.30 1.46 7.30 1.28 2.88 1310
149 3 13.08 3.90 2.36 21.5 113 1.41 1.39 0.34 1.14 9.40 0.57 1.33 550
    1 14.19 1.59 2.48 16.5 108 3.30 3.93 0.32 1.86 8.70 1.23 2.82 1680
18
    2 12.33 0.99 1.95 14.8 136 1.90 1.85 0.35 2.76 3.40 1.06 2.31 750
108 2 12.22 1.29 1.94 19.0 92 2.36 2.04 0.39 2.08 2.70 0.86 3.02
                                                                 312
146 3 13.88 5.04 2.23 20.0 80 0.98 0.34 0.40 0.68 4.90 0.58 1.33
                                                                  415
    3 13.69 3.26 2.54 20.0 107 1.83 0.56 0.50 0.80 5.88 0.96 1.82
                                                                  680
111 2 12.52 2.43 2.17 21.0 88 2.55 2.27 0.26 1.22 2.00 0.90 2.78 325
```

```
54
    1 13.74 1.67 2.25 16.4 118 2.60 2.90 0.21 1.62
                                                     5.85 0.92 3.20 1060
26
    1 13.39 1.77 2.62 16.1 93 2.85 2.94 0.34 1.45
                                                     4.80 0.92 3.22 1195
     2 12.17 1.45 2.53 19.0 104 1.89 1.75 0.45 1.03
64
                                                     2.95 1.45 2.23
                                                                     355
166 3 13.45 3.70 2.60 23.0 111 1.70 0.92 0.43 1.46 10.68 0.85 1.56
                                                                     695
147 3 12.87 4.61 2.48 21.5 86 1.70 0.65 0.47 0.86
                                                     7.65 0.54 1.86
                                                                    625
     1 13.72 1.43 2.50 16.7 108 3.40 3.67 0.19 2.04
                                                     6.80 0.89 2.87 1285
58
154 3 12.58 1.29 2.10 20.0 103 1.48 0.58 0.53 1.40
                                                   7.60 0.58 1.55
49
     1 13.94 1.73 2.27 17.4 108 2.88 3.54 0.32 2.08
                                                    8.90 1.12 3.10 1260
     2 12.37 1.07 2.10 18.5 88 3.52 3.75 0.24 1.95 4.50 1.04 2.77
98
                                                                     660
     2 12.33 1.10 2.28 16.0 101 2.05 1.09 0.63 0.41
60
                                                     3.27 1.25 1.67
                                                                     680
```

```
#change the column names
colnames(small_wine.df) <- paste("v", 1:dim(small_wine.df)[2], sep="")</pre>
v1, v2, \ldots v13, v14
head(small_wine.df, n=5)
Output:
           v2
                     v4
                          v5 v6
                                              v9 v10
                                                        v11
                                                             v12 v13
     ν1
                v3
                                    v7
                                         ν8
                                                                        v14
     3 13.27 4.28 2.26 20.0 120 1.59 0.69 0.43 1.35 10.20 0.59 1.56
                                                                        835
 175
      1 13.50 1.81 2.61 20.0
                              96 2.53 2.61 0.28 1.66
                                                       3.52 1.12 3.82
                                                                        845
 97
      2 12.29 1.41 1.98 16.0
                              85 2.55 2.50 0.29 1.77
                                                       2.90 1.23 2.74
                                                                        428
 51
      1 13.83 1.65 2.60 17.2 94 2.45 2.99 0.22 2.29
                                                       5.60 1.24 3.37 1265
      1 14.20 1.76 2.45 15.2 112 3.27 3.39 0.34 1.97 6.75 1.05 2.85 1450
```

Step 4 of N: Handle Missing Values in data

After reading the dataset into R, do a summary on your data and deal with missing values in the data, See Remove/Replace/Deal with NA entries

```
#maybe
t(summary(small_wine.df)) # t(..) to transpose

Output:
```

```
:3.00
          :1.00
v1 Min.
                    1st Qu.:1.00
                                      Median :2.00
                                                        Mean
                                                               :1.92
                                                                         3rd Qu.:3.00
                                                                                           Max.
                                      Median :13.27
v2 Min.
          :11.56
                    1st Qu.:12.37
                                                       Mean
                                                               :13.09
                                                                         3rd Qu.:13.72
                                                                                           Max.
                                                                                                   :14.20
                                      Median :1.730
v3 Min.
          :0.990
                                                               :2.193
                                                                                                   :5.040
                    1st Qu.:1.430
                                                       Mean
                                                                         3rd Qu.:2.430
                                                                                           Max.
v4 Min.
          :1.940
                                      Median :2.360
                                                                                                   :3.230
                    1st Qu.:2.170
                                                        Mean
                                                               :2.368
                                                                         3rd Qu.:2.540
                                                                                           Max.
                                      Median :19.00
v5 Min.
          :14.80
                    1st Qu.:16.50
                                                        Mean
                                                               :18.88
                                                                         3rd Qu.:20.00
                                                                                           Max.
                                                                                                   :28.50
v6 Min.
          : 80.0
                                      Median :104.0
                                                                                                   :136.0
                    1st Qu.: 92.0
                                                        Mean
                                                               :102.7
                                                                         3rd Qu.:112.0
                                                                                           Max.
          :0.980
                                                               :2.361
v7 Min.
                                      Median :2.450
                                                                         3rd Qu.:2.850
                                                                                           Max.
                                                                                                   :3.520
                    1st Qu.:1.830
                                                        Mean
                                      Median :2.27
v8 Min.
          :0.34
                                                               :2.25
                                                                         3rd Qu.:2.99
                                                                                           Max.
                                                                                                  :5.08
                    1st Qu.:1.09
                                                        Mean
v9 Min.
          :0.1900
                    1st Qu.:0.2900
                                      Median :0.3400
                                                               :0.3608
                                                                         3rd Qu.:0.4300
                                                                                           Max.
                                                                                                  :0.6300
                                                       Mean
```

```
:0.410
v10 Min.
                   1st Qu.:1.220
                                   Median :1.480
                                                   Mean
                                                          :1.548
                                                                    3rd Qu.:1.950
                                                                                   Max.
                                                                                          :2.760
                                   Median : 5.850
v11 Min.
          : 2.000 1st Qu.: 3.400
                                                          : 5.808
                                                                    3rd Qu.: 7.600
                                                                                   Max.
                                                                                          :10.680
                                                   Mean
v12 Min.
          :0.540
                   1st Qu.:0.860
                                   Median :0.960
                                                          :0.968
                                                                    3rd Qu.:1.120
                                                                                   Max.
                                                                                          :1.450
                                                   Mean
v13 Min.
          :1.330
                   1st Qu.:1.820
                                   Median :2.780
                                                   Mean
                                                          :2.557
                                                                    3rd Qu.:3.100
                                                                                   Max.
                                                                                          :3.820
v14 Min. : 312.0 1st Qu.: 550.0 Median : 680.0
                                                   Mean
                                                          : 817.5
                                                                   3rd Qu.:1195.0
                                                                                   Max.
                                                                                          :1680.0
```

R-CODE

```
# what about big wine
t(summary(big_wine.df))
```

```
Output:
    v1 Min.
              :1.000
                       1st Qu.:1.000
                                        Median :2.000
                                                                :1.944
                                                                         3rd Qu.:3.000
                                                                                                 :3.000
                                                         Mean
                                                                                          Max.
                                                                                          Max.
    v2 Min.
              :11.03
                                        Median :13.05
                                                                :12.99
                                                                                                 :14.83
                       1st Qu.:12.36
                                                         Mean
                                                                         3rd Qu.:13.67
    v3 Min.
              :0.74
                       1st Qu.:1.60
                                        Median :1.87
                                                         Mean
                                                                :2.34
                                                                         3rd Qu.:3.10
                                                                                          Max.
                                                                                                 :5.80
    v4 Min.
              :1.360
                       1st Qu.:2.210
                                        Median :2.360
                                                         Mean
                                                                :2.366
                                                                         3rd Qu.:2.560
                                                                                          Max.
                                                                                                 :3.230
                                                                         3rd Qu.:21.50
    v5 Min.
              :10.60
                       1st Qu.:17.20
                                        Median :19.50
                                                                :19.52
                                                                                                 :30.00
                                                         Mean
                                                                                          Max.
    v6 Min.
              : 70.00 1st Qu.: 88.00
                                        Median : 98.00
                                                         Mean
                                                                : 99.59
                                                                         3rd Qu.:107.00
                                                                                          Max.
                                                                                                 :162.00
    v7 Min.
              :0.980
                                        Median :2.350
                                                                :2.292
                       1st Qu.:1.740
                                                         Mean
                                                                         3rd Qu.:2.800
                                                                                          Max.
                                                                                                 :3.880
    v8 Min.
              :0.340
                       1st Qu.:1.200
                                        Median :2.130
                                                                :2.023
                                                                         3rd Qu.:2.860
                                                                                                 :5.080
                                                         Mean
                                                                                          Max.
              :0.1300 1st Qu.:0.2700
    v9 Min.
                                        Median :0.3400
                                                        Mean
                                                               :0.3623
                                                                         3rd Qu.:0.4400
                                                                                          Max.
                                                                                                 :0.6600
   v10 Min.
              :0.410
                       1st Qu.:1.250
                                                               :1.587
                                                                         3rd Qu.:1.950
                                        Median :1.550
                                                         Mean
                                                                                          Max.
                                                                                                 :3.580
   v11 Min.
                                        Median : 4.680
                                                                         3rd Qu.: 6.200
              : 1.280 1st Qu.: 3.210
                                                         Mean
                                                                : 5.055
                                                                                          Max.
                                                                                                 :13.000
   v12 Min.
              :0.480
                       1st Qu.:0.780
                                        Median :0.960
                                                         Mean
                                                                :0.957
                                                                         3rd Qu.:1.120
                                                                                          Max.
                                                                                                 :1.710
   v13 Min.
              :1.270
                       1st Qu.:1.930
                                        Median :2.780
                                                         Mean
                                                                :2.604
                                                                         3rd Qu.:3.170
                                                                                          Max.
                                                                                                 :4.000
   v14 Min. : 278.0 1st Qu.: 500.0 Median : 672.0
                                                        Mean : 745.1 3rd Qu.: 985.0
                                                                                                :1680.0
                                                                                         Max.
```

R-CODE

#Elegant way to report missing values in a data.frame: http://stackoverflow.
com/questions/8317231/elegant-way-to-report-missing-values-in-a-data-frame
sapply(big_wine.df, function(x) sum(is.na(x)))

Output:

```
v1 v2 v3 v4 v5 v6 v7 v8 v9 v10 v11 v12 v13 v14 0 0 0 0 0 0 0 0 0 0 0 0 0
```

Step 4 of N: Plotting Multivariate Data

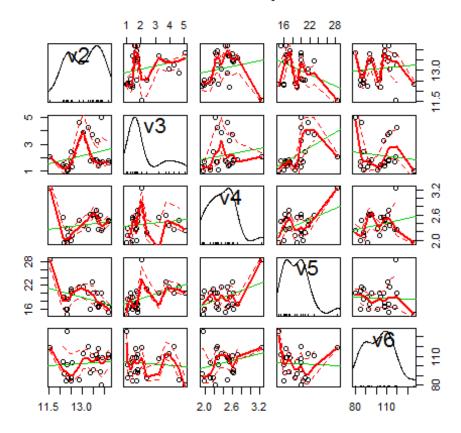
After reading the dataset into R, the next step is usually to plot of the resulting data See Scatterplot matrices in R

A Matrix Scatterplot

For multivariate data, one usually makes a matrix scatterplot, showing each pair of variables plotted against each other. The scatterplotMatrix(...) function is available in the car package. That is: install.packages("car"); library(car)

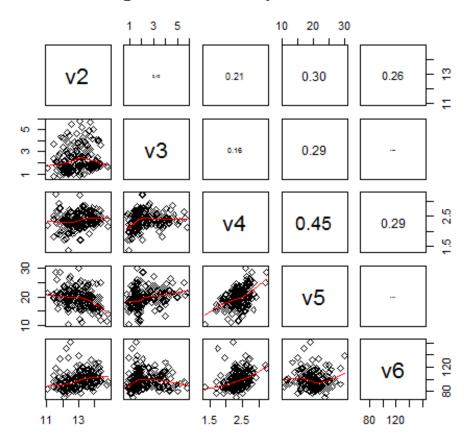
R-CODE

Small Wine Scatterplot Matrix



```
# panel.smooth function is built in.
# panel.cor puts correlation in upper panels, size proportional to correlatio
panel.cor <- function(x, y, digits=2, prefix="", cex.cor, ...)</pre>
    usr <- par("usr"); on.exit(par(usr))</pre>
    par(usr = c(0, 1, 0, 1))
    r \leftarrow abs(cor(x, y))
    txt <- format(c(r, 0.123456789), digits=digits)[1]</pre>
    txt <- paste(prefix, txt, sep="")</pre>
    if(missing(cex.cor)) cex.cor <- 0.8/strwidth(txt)</pre>
    text(0.5, 0.5, txt, cex = cex.cor * r)
}
# Plot #2: same as above, but add loess smoother in lower and correlation in
upper
windows()
pairs(big_wine.df[, 2:6],
      lower.panel=panel.smooth, # lower portion of plot is the smooth plot
      upper.panel=panel.cor, # uppper portion is a correlation plot
                             # plot symbols, see http://www.statmethods.net/ad
      pch=23,
vgraphs/parameters.html
      main="Big Wine Scatterplot Matrix",
      na.action = na.omit
                                 # causes cases with missing values in any of
the variables to be omitted entirely.
```

Big Wine Scatterplot Matrix



Scatterplot with data Points labelled by group

Observation If you observe any interesting scatterplot for any two variables in the matrix scatterplot, plot that scatterplot in more detail, with the data points labelled by their group (their cultivar, i.e., v1 ... v14)

For example, the pair plot above, shows the 3rd column of the 4th row down is a scatterplot of V4 (x-axis) against V5 (y-axis). The figure shows a correlation of 0.45, that is a positive relationship between V4 and V5.

Zoom in on any interesting relationship with a plot

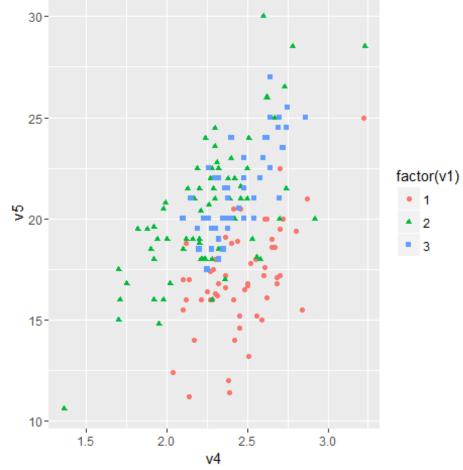
#Use plot(x=, y=) or qplot(x=, y=). I will use ggplot(data=...)

```
# simplr plot
## define base for the graphs and store in object 'p'

v4_v5.plot <- ggplot(data = big_wine.df, aes(x = v4, y = v5, group=v1, color=
factor(v1), shape=factor(v1)))

## just plotting points (a.k.a., scatterplot)
windows()
v4_v5.plot + geom_point() # + #geom_line() # simple spaghetti plot</pre>
```

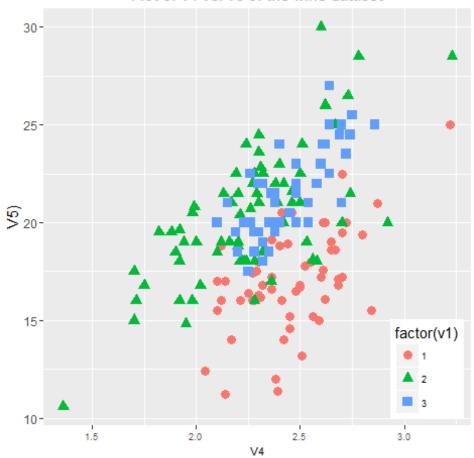




```
# A more elaborate plot
# setup plot
v4_v5.plot <- ggplot(
                                          # cool plotting package
                                          # dataset in this case a data.frame
  data=big wine.df,
  aes(
         x = v4
                                          # x-axis
         y = v5
                                          # y-axis
    colour= factor(v1),
                                          # color to use for the points
    shape = factor(v1)
    ))+
  geom_point( size = 3) +
                                         # plot points only, with increased
size
  xlab("V4")
                                          # x and y lables
  ylab("V5)")
  ggtitle("Plot of V4 vs. V5 of the wine dataset")
v4_v5.plot <- v4_v5.plot +
  theme(
                               # the beauty of ggplot, you can add layers lat
er
           plot.title = element_text(lineheight=1.2, face="bold", size=10.5 )
  #beautify title
           legend.position = c(1.01, 0.28), legend.justification=c(1,1),
#position the legend
      legend.position = "bottom", legend.justification=c(1,1),
#position the legend
          legend.text = element_text(size=6.5),
#format legend text
          axis.text.x = element_text(size=6.5),
#format x-axis text
         axis.title.x = element text(size = rel(0.75))
#format x-axis title
  )
#--
```

```
#open a display window, else R will use RStudio's window
windows()
#-
# Add aesthetic mappings
#v4_v5.plot <- v4_v5.plot + geom_text(aes(colour = factor(v1), size=2))
# now display the plot
v4_v5.plot</pre>
Output
```

Plot of V4 vs. V5 of the wine dataset



Observation fom detailed plot

We can see from the plot of x=V4 versus y=V5 that

wines from cultivar 1 seem to have lower values of V5 compared to the wines of cultivar 2 and 3 $\,$

wines from cultivar 2 seem to have lower values of V4 compared to the wines of cultivar 1

A Profile Plot

Good reference for profile plot

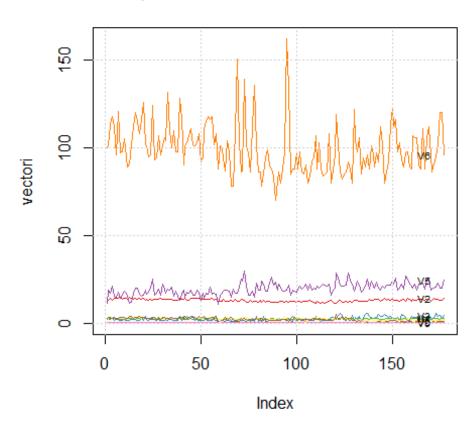
Useful functions for Multivariate data analysis

Another useful plot is a profile plot which plots the value of each of the variables for each of the samples to show the variation in each of the variables.

 Use the function makeProfilePlot(...) to make a profile plot. ++ Note: The makeProfilePlot(...) function requires the RColorBrewer library

A profile plot of the concentrations of the first 9 chemicals in the wine samples in columns V2, V3, V4, V5, V6, v7, v8, v9 of the big wine dataset

Profile plot of: V2, V3, V4, V5, V6, V7, V8, V9



Observation fom detailed plot

We can see from the profile plot that * the mean and standard deviation for V6 is quite a lot higher than that for the other variables.

Calculating Summary Statistics for Multivariate Data

- For statistics on a dataset refer to the Mosaic package.
- Here we would calculate 2 summary statistics for each of the variables in the wine data set
- Use sapply mean standard deviation

```
      sapply(big_wine.df[, 2:9], mean)
      #mean and std of 2nd to 9th variables

      Output
      v2
      v3
      v4
      v5
      v6
      v7
      v8
      v9

      12.9936723
      2.3398870
      2.3661582
      19.5169492
      99.5875706
      2.2922599
      2.0234463
      0.3623164
```

R-CODE

R-CODE

```
# per sample type
group1_index = which(big_wine.df$v1 == 1); head(group1_index) # get the ind
ex
Output
[1] 1 2 3 4 5 6
```

R-CODE

```
group2_index = which(big_wine.df$v1 == 2); head(group2_index)
Output
[1] 59 60 61 62 63 64
```

```
group3_index = which(big_wine.df$v1 == 3); head(group3_index)
Output
[1] 130 131 132 133 134 135
```

```
sapply(big_wine.df[ group1_index, 2:9], mean) #mean and std of 2nd to
9th variables

Output

v2  v3  v4  v5  v6  v7  v8  v9
13.7363793  2.0158621  2.4560345  17.0620690 105.9827586  2.8408621  2.9810345  0.2901724
```

R-CODE

```
sapply(big_wine.df[ group2_index, 2:9], mean)

Output

v2     v3     v4     v5     v6     v7     v8     v9
12.278732  1.932676  2.244789  20.238028  94.549296  2.258873  2.080845  0.363662
```

R-CODE

```
sapply(big_wine.df[ group3_index, 2:9], mean)

Output

v2     v3     v4     v5     v6     v7     v8     v9
13.1537500    3.3337500    2.4370833    21.4166667    99.3125000    1.6787500    0.7814583    0.4475000
```

Observation

- We are only able to compute the mean and standard deviation of the 2-9 chemicals' concentrations for just cultivar 1 samples, or for just cultivar 3 samples, in a similar way. statistics by variable for the entire group.
- What about by group with with just 1 line of command? Use google, e.g.: "r calculating standard deviation by group in a data"

Means and Variances Per Group

R-CODE

```
big_wine_groups.df <- t( big_wine.df %>% group_by(v1) %>% summarise_each(funs
(mean, sd)) ) # transpose
colnames(big wine groups.df) <- c("Group 1", "Group 2", "Group 3")</pre>
#--
# exclude row 1, V1
big_wine_groups.df <- big_wine_groups.df[-(1), ]</pre>
head(big wine groups.df)
Output
                      Group 2
            Group 1
                                Group 3
 v2 mean 13.736379 12.278732 13.153750
 v3 mean 2.015862 1.932676 3.333750
 v4 mean 2.456034 2.244789 2.437083
 v5 mean 17.062069 20.238028 21.416667
 v6 mean 105.982759 94.549296 99.312500
 v7_mean
           2.840862 2.258873 1.678750
```

R-CODE

```
#show me rowsnames containing the word "mean"
# Use google: search for: "R show rownames that contain a word""
#http://stackoverflow.com/questions/13043928/selecting-rows-where-a-column-ha
s-a-string-like-hsa-partial-string-match
index_of_row_name_contain_mean <- grep("_mean", rownames(big_wine_groups.df))
Output
2  3  4  5  6  7  8  9 10 11 12 13 14</pre>
```

```
# Now show the rows containing the word "mean"
group.means <- big_wine_groups.df[index_of_row_name_contain_mean, ]</pre>
group.means
Output
              Group_1
                          Group_2
                                      Group_3
           13.7363793 12.278732 13.1537500
v2 mean
v3 mean
             2.0158621
                        1.932676 3.3337500
                         2.244789
v4_mean
             2.4560345
                                   2.4370833
v5 mean
           17.0620690 20.238028 21.4166667
v6_mean
          105.9827586 94.549296 99.3125000
```

```
v7 mean
                      2.258873
                                1.6787500
           2.8408621
v8 mean
           2.9810345
                      2.080845
                                0.7814583
v9_mean
           0.2901724
                      0.363662
                                0.4475000
v10 mean
           1.8925862 1.630282
                                1.1535417
v11_mean 5.5263793
                      3.086620
                                7.3962500
v12 mean
           1.0624138
                      1.056282
                                0.6827083
v13 mean
           3.1446552
                      2.785352
                                1.6835417
v14 mean 1116.5862069 519.507042 629.8958333
```

```
#---
# do the same for variance
group.variance <- big_wine_groups.df[-c(index_of_row_name_contain_mean), ]</pre>
group.variance
Output
                         Group_2
             Group_1
                                     Group_3
v2 sd
          0.46163211
                       0.5379642
                                   0.5302413
v3 sd
         0.69340005
                       1.0155687
                                  1.0879057
v4_sd
         0.22912449
                       0.3154673
                                   0.1846902
v5 sd
         2.56137488
                       3.3497704
                                  2.2581609
v6_sd 10.22465438 16.7534975 10.8904726
v7_sd
         0.34187966
                       0.5453611
                                  0.3569709
v8 sd
         0.40083111
                       0.7057008
                                  0.2935041
v9 sd
         0.07064841
                       0.1239613
                                   0.1241396
v10_sd
         0.41241931
                       0.6020678
                                   0.4088359
v11 sd
         1.24930114
                       0.9249293
                                   2.3109421
v12_sd
         0.11746310
                       0.2029368
                                   0.1144411
v13 sd
         0.34550353
                       0.4965735
                                   0.2721114
v14 sd 223.35276437 157.2112204 115.0970432
```

*** NEXT WE ARE GOING TO PERFORM ANALYSIS OF VARIANCE TEST ON A SMALLER DATASET ***

Between-groups Variance and Within-groups Variations

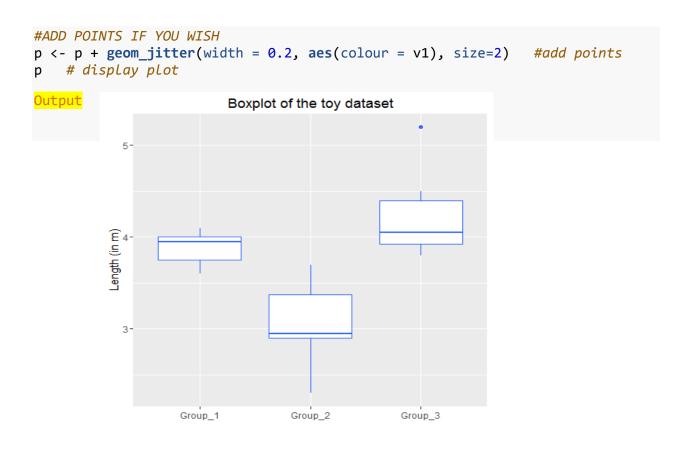
This following shows how we can calculate the within-groups and between-groups variation (e.g., within-group variation for a particular chemical's concentration)

* To tackle this problem, lets look at a smaller toy dataset

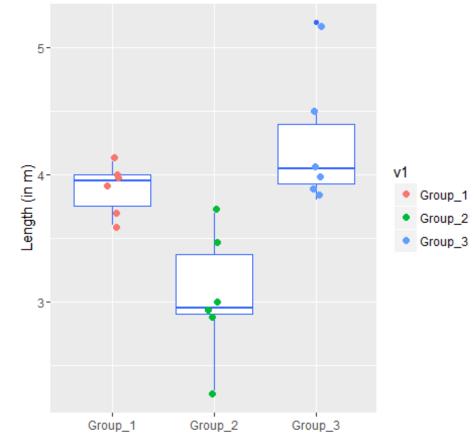
R-CODE

```
toy.data <- read.table(header = TRUE, text = "</pre>
Group_1 Group_2 Group_3
4.0 2.9 4.5
3.6 2.3 3.8
3.7 2.9 4.0
4.1 3.5 5.2
3.9 3.7 3.9
4.0 3.0 4.1
")
                                              #show the data
toy.data
Output
   Group_1 Group_2 Group_3
 1
      4.0
              2.9
                      4.5
                      3.8
      3.6
              2.3
 2
 3
      3.7
             2.9
                      4.0
      4.1
 4
                      5.2
              3.5
 5
      3.9
                      3.9
              3.7
     4.0
              3.0 4.1
 6
```

```
4 Group 1 4.1
 5 Group 1 3.9
 6 Group_1 4.0
 7 Group_2 2.9
 8 Group_2 2.3
 9 Group_2 2.9
 10 Group 2 3.5
 11 Group_2 3.7
 12 Group_2 3.0
 13 Group_3 4.5
 14 Group_3 3.8
 15 Group 3 4.0
 16 Group_3 5.2
 17 Group_3 3.9
 18 Group_3 4.1
##### Boxplot
p <- ggplot(toy.df, aes(v1, v2))</pre>
p <- p + geom_boxplot(fill = "white", colour = "#3366FF") +</pre>
     ylab("Length (in m)") + xlab("") +
     ggtitle("Boxplot of the toy dataset")
p # display plot
```



Boxplot of the toy dataset



```
Descriptive Statistics
```

```
#--
# Get N
N.dim = dim( toy.df )
N <- ( N.dim[1] * N.dim[2] ) - N.dim[1] # could have simply said N.dim
[1], but setting program up for big_wine.df
N
Output [1] 18</pre>
```

```
toy.table <- toy.df %>% group_by(v1) %>% summarise_each(funs(length, mean, sd
, min, max))
toy.table$stderr <- toy.table$sd / sqrt(toy.table$length)</pre>
toy.table$E
                     <- qnorm(.975)*toy.table$stderr</pre>
                                                         # margin of error
95% CI
toy.table$lowerbound <- toy.table$mean - toy.table$E</pre>
toy.table$upperbound <- toy.table$mean + toy.table$E</pre>
descriptive.stats <- data.frame(</pre>
           Groups = c(1:3),
                N = c(toy.table$length),
             Mean = c(toy.table$mean),
  "Std Deviation" = c(toy.table$sd),
      "Std Error" = c(toy.table$stderr),
 "CI_Lower Bound" = c(toy.table$lowerbound),
 "CI_Upper Bound" = c(toy.table$upperbound),
        "Minimum" = c(toy.table$min),
       "Maximum" = c(toy.table$max)
 )
```

Analysis of Variance (ANOVA) Table

R-CODE

```
#Anova
#--
#SSW
SSW.v1 <- sum( ( toy.df$v2[ 1: 6] - toy.table$mean[1] )^2 )
SSW.v2 \leftarrow sum( (toy.df$v2[7:12] - toy.table$mean[2])^2)
SSW.v3 \leftarrow sum( ( toy.df$v2[13:18] - toy.table$mean[3] )^2 )
SSW \leftarrow SSW.v1 + SSW.v2 + SSW.v3
#SST
SST.v1 \leftarrow sum((toy.df$v2[1:6] - x_2bar)^2)
SST.v2 \leftarrow sum((toy.df$v2[7:12] - x_2bar)^2)
SST.v3 \leftarrow sum((toy.df$v2[13:18] - x_2bar)^2)
SST <-SST.v1 + SST.v2 + SST.v3
\#SSB + SSW = SST
SSB.v1 \leftarrow sum( 6 * ( toy.table$mean[1] - x_2bar )^2 )
SSB.v2 <- sum(6 * (toy.table\$mean[2] - x_2bar)^2)
SSB.v3 <- sum(6 * (toy.table\$mean[3] - x_2bar)^2)
SSB <-SSB.v1 + SSB.v2 + SSB.v3
F = (SSB / (K - 1)) / (SSW / (N - K))
anova.table <- data.frame("sum of squares"=c(SSB, SSW, SSW),
                          df = c(K-1, N-K, N-1),
                          "mean square" = c(SSB/(K-1), SSW/(N-K), SST/(N-1)),
                           F = c(F, 0, 0)
```

Show descriptive statistics and ANOVA table

Output from SPSS (Analyze -> Compare means->ANOVA

R-CODE

```
descriptive.stats
```

Output

```
# Show ANOVA Table
```

anova.table

Output

```
sum.of.squares df mean.square
1
   4.537778 2 2.2688889 12.162
2
      2.798333 15
                   0.1865556
                             0.000
      2.798333 17
                   0.4315359 0.000
```

Next we compare with critical value from table

Compare with critical value from table



With N-1 (18-1=17) and N-K (18-3=15) degrees of freedom and a two-tailed α of 0.5 the critical value of F = 2.37



Because a F_{obs} of 12.162 is outside (greater than) a F_{crit} of 2.37, we must reject the null hypothesis that the population means from the groups are equal.

** We will extend this to out wine dataset in out next lecture ***