

# Trabajo fin de master

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## Base de datos

Base de datos con los años 2013, 2014 y 2015 con las capturas de *Thanasimus formicarius* y las capturas de los escolítidos de interés.

Tabla descriptiva de los datos:

```
## DatosModelo
##
## 26 Variables      525 Observations
## -----
## Plot
##      n missing distinct
##    525      0      15
##
## Value      1      2      3      4      5      6      7      8      9     10     11
## Frequency   35     35     35     35     35     35     35     35     35     35     35
## Proportion 0.067 0.067 0.067 0.067 0.067 0.067 0.067 0.067 0.067 0.067 0.067
##
## Value      12     13     14     16
## Frequency   35     35     35     35
## Proportion 0.067 0.067 0.067 0.067
## -----
## Periodo
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    525      0      35     0.999      18     11.68      2      4
##      .25      .50      .75      .90      .95
##      9      18      27      32      34
##
## lowest : 1 2 3 4 5, highest: 31 32 33 34 35
## -----
## Month
##      n missing distinct
##    525      0      12
##
## Value      1     10     11     12      2      3      4      5      6      7      8
## Frequency   45     45     45     30     45     45     45     45     45     45     45
## Proportion 0.086 0.086 0.086 0.057 0.086 0.086 0.086 0.086 0.086 0.086 0.086
##
## Value      9
## Frequency   45
## Proportion 0.086
## -----
## Year
```

```

##          n missing distinct
##        525         0         3
##
## Value          2013  2014  2015
## Frequency      180   180   165
## Proportion 0.343 0.343 0.314
## -----
## Codigo
##          n missing distinct
##        525         0         15
##
## 10Penyagolps (35, 0.067), 11Penyagolpn (35, 0.067), 12Penyagolpp (35, 0.067),
## 13Aitana (35, 0.067), 14Jérica (35, 0.067), 16Morella (35, 0.067), 1Guardamar
## (35, 0.067), 2Crevillent (35, 0.067), 3Biar (35, 0.067), 4Agres (35, 0.067),
## 5Saler (35, 0.067), 6PSMps (35, 0.067), 7PSMpn (35, 0.067), 8Sinarcas (35,
## 0.067), 9Poblatornesa (35, 0.067)
## -----
## Captures
##          n missing distinct      Info      Mean      Gmd      .05      .10
##        525         0         81    0.881    13.69    22.29     0.0     0.0
##        .25      .50      .75      .90      .95
##        0.0      1.0     11.0     45.0     73.4
##
## lowest :    0    1    2    3    4, highest: 154 162 179 186 191
## -----
## Provincia
##          n missing distinct
##        525         0         3
##
## Value          Alicante Castellón  Valencia
## Frequency          175      210      140
## Proportion      0.333      0.400      0.267
## -----
## Comarca
##          n missing distinct
##        525         0         12
##
## lowest : El Alto Palancia          El Baix Segura / La Vega Baja El Baix Vinalopò
## highest: La Marina Baixa          La Plana Alta          La Plana Utiel - Requena
## -----
## TM
##          n missing distinct
##        525         0         12
##
## lowest : Agres          Biar          Confrides          Crevillent
## highest: Pobra Tornesa    Puebla de San Miguel    Saler          Sinarcas
## -----
## AltMin
##          n missing distinct      Info      Mean      Gmd      .05      .10
##        525         0         14    0.994    812.9    585.2     0     3
##        .25      .50      .75      .90      .95
##        300      900     1280     1400     1650
##
## Value          0.0  165.0  297.0  478.5  610.5  825.0  891.0 1089.0 1188.0

```

```

## Frequency      70      35      35      35      35      35      70      35      35
## Proportion 0.133 0.067 0.067 0.067 0.067 0.067 0.133 0.067 0.067
##
## Value      1270.5 1336.5 1386.0 1650.0
## Frequency      35      35      35      35
## Proportion 0.067 0.067 0.067 0.067
##
## For the frequency table, variable is rounded to the nearest 16.5
## -----
## AltMax
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      14      0.994      1026      638.9      10      35
##      .25      .50      .75      .90      .95
##      500      1200      1510      1700      1836
##
## Value      10.00      28.26      393.46      484.76      667.36      904.74      1087.34      1196.90
## Frequency      35      35      35      35      35      35      35      35
## Proportion 0.067 0.067 0.067 0.067 0.067 0.067 0.067 0.067
##
## Value      1233.42 1470.80 1507.32 1543.84 1689.92 1836.00
## Frequency      70      35      35      35      35      35
## Proportion 0.133 0.067 0.067 0.067 0.067 0.067
##
## For the frequency table, variable is rounded to the nearest 18.26
## -----
## Superf
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      15      0.996      342      271.4      33.64      35.78
##      .25      .50      .75      .90      .95
##      81.83      358.11      522.32      702.70      745.70
##
## Value      33.6400      55.0018      76.3636      104.8460      211.6550      275.7404      354.0670
## Frequency      70      35      35      35      35      35      35
## Proportion 0.133 0.067 0.067 0.067 0.067 0.067 0.067
##
## Value      411.0318      475.1172      510.7202      517.8408      574.8056      695.8558      745.7000
## Frequency      35      35      35      35      35      35      35
## Proportion 0.067 0.067 0.067 0.067 0.067 0.067 0.067
##
## For the frequency table, variable is rounded to the nearest 7.1206
## -----
## CoordX
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      15      0.996      675902      84611      247400      653710
##      .25      .50      .75      .90      .95
##      661762      710475      729695      737989      741803
##
## Value      247400.0      652810.5      657754.5      692362.7      697306.7      702250.8      707194.8
## Frequency      35      35      70      35      35      35      35
## Proportion 0.067 0.067 0.133 0.067 0.067 0.067 0.067
##
## Value      712138.8      722026.9      726970.9      736859.0      741803.0
## Frequency      35      70      70      35      35
## Proportion 0.067 0.133 0.133 0.067 0.067

```

```

##
## For the frequency table, variable is rounded to the nearest 4944.03
## -----
## CoordY
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      15      0.996  4379218      97937  4219823  4239891
##      .25      .50      .75      .90      .95
##  4281927  4416944  4457593  4469449  4490223
##
## Value      4219823  4238751  4279311  4292831  4357727  4403695  4414511  4430735
## Frequency      35      35      70      35      35      35      35      35
## Proportion  0.067  0.067  0.133  0.067  0.067  0.067  0.067  0.067
##
## Value      4436143  4441551  4455071  4457775  4468591  4490223
## Frequency      35      35      35      35      35      35
## Proportion  0.067  0.067  0.067  0.067  0.067  0.067
##
## For the frequency table, variable is rounded to the nearest 2704
## -----
## TMinMed
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      475      1      9.272      6.922      0.198      1.496
##      .25      .50      .75      .90      .95
##   4.370   9.060   13.780   17.456   19.338
##
## lowest : -3.37 -2.96 -1.96 -1.88 -1.78, highest: 22.48 22.98 23.38 23.55 23.79
## -----
## TMaxMed
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      470      1      19.72      8.305      8.404      9.784
##      .25      .50      .75      .90      .95
##  14.080  19.940  25.430  29.392  31.024
##
## lowest : 2.15  2.85  4.86  4.88  5.03 , highest: 33.44 33.75 33.77 34.89 35.86
## -----
## Precipita
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      269      1      1.502      1.554      0.110      0.174
##      .25      .50      .75      .90      .95
##   0.410   0.960   2.030   3.682   4.596
##
## lowest : 0      0.01  0.02  0.03  0.04 , highest: 9.04  9.62  9.92 10.56 11.68
## -----
## Hylastesater
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      47      0.723      7.949      14.66      0.0      0.0
##      .25      .50      .75      .90      .95
##   0.0      0.0      2.0      12.0      27.6
##
## lowest :  0    1    2    3    4, highest: 128 320 461 484 571
## -----
## Hylastesattenuatus
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      68      0.558      19.01      36.06      0.0      0.0

```

```

##      .25      .50      .75      .90      .95
##      0.0      0.0      0.0     20.2     99.8
##
## lowest :    0    1    2    3    4, highest: 414 516 551 878 1558
## -----
## Hylurgusmiklitzii
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      112    0.905    34.18    59.04      0.0      0.0
##      .25      .50      .75      .90      .95
##      0.0      1.0     19.0    84.6    160.6
##
## lowest :    0    1    2    3    4, highest: 558 562 961 1134 1420
## -----
## Ipssexdentatus
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      97    0.718    41.19    76.1      0.0      0.0
##      .25      .50      .75      .90      .95
##      0.0      0.0      3.0    85.4    233.0
##
## lowest :    0    1    2    3    4, highest: 963 1053 1080 1653 1665
## -----
## Tomicusdestruens
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      22    0.345    0.9829    1.888      0.0      0.0
##      .25      .50      .75      .90      .95
##      0.0      0.0      0.0      1.0      3.8
##
## lowest :    0    1    2    3    4, highest: 20 23 53 72 93
## -----
## Suma
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      334      1    285.6    363.6      4.0     14.0
##      .25      .50      .75      .90      .95
##      41.0    125.0    350.0    695.6    993.8
##
## lowest :    0    1    2    4    5, highest: 2052 3422 3597 4297 5397
## -----
## AltMed
##      n missing distinct      Info      Mean      Gmd      .05      .10
##      525      0      15    0.996    919.5    606.2       5      19
##      .25      .50      .75      .90      .95
##      400     995     1415    1550    1743
##
## Value      5.00  265.70  387.36  578.54  891.38  926.14  978.28 1013.04
## Frequency      70      35      35      35      35      35      35      35
## Proportion  0.133  0.067  0.067  0.067  0.067  0.067  0.067  0.067
##
## Value      1169.46 1325.88 1412.78 1534.44 1743.00
## Frequency      35      35      70      35      35
## Proportion  0.067  0.067  0.133  0.067  0.067
##
## For the frequency table, variable is rounded to the nearest 17.38
## -----
## SumaSpecies

```

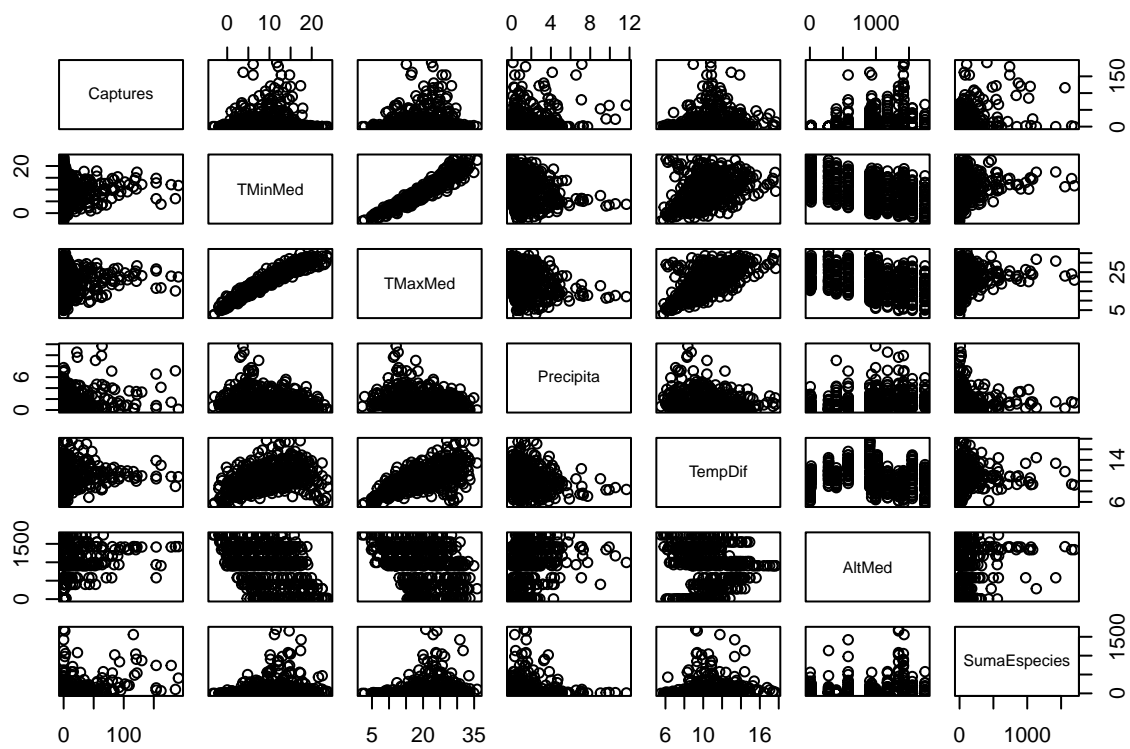
```
##          n missing distinct      Info      Mean      Gmd      .05      .10
##        525         0      198    0.996    103.3    158.7      0.0      0.0
##        .25        .50        .75      .90      .95
##        2.0       20.0     105.0    282.6    530.8
##
## lowest :      0      1      2      3      4, highest: 1134 1421 1558 1656 1696
## -----
## TempDif
##          n missing distinct      Info      Mean      Gmd      .05      .10
##        525         0      449         1    10.45     2.489     7.012     7.564
##        .25        .50        .75      .90      .95
##        9.050    10.320    11.770    13.522    14.232
##
## lowest : 5.52  5.81  5.86  6.01  6.08 , highest: 17.04 17.38 17.4  17.62 17.68
## -----
```

Correlaciones entre nuestras variables:

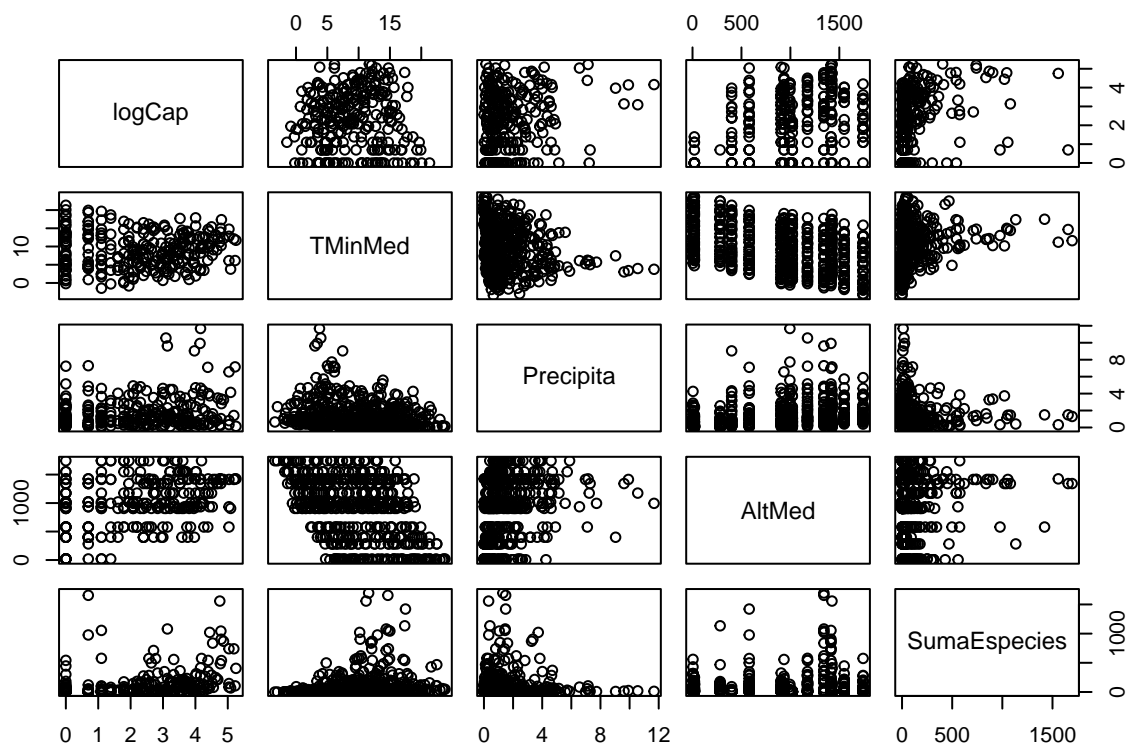
```
round(cor(DatosModelo[,c(6,15,16,17,26,24,25)]),2)
```

```
##          Captures TMinMed TMaxMed Precipita TempDif AltMed SumaEspecies
## Captures          1.00    0.03    0.07     0.20    0.14    0.22         0.35
## TMinMed           0.03    1.00    0.96    -0.16    0.40   -0.42         0.23
## TMaxMed           0.07    0.96    1.00    -0.21    0.65   -0.40         0.23
## Precipita         0.20   -0.16   -0.21     1.00   -0.22    0.23        -0.05
## TempDif           0.14    0.40    0.65    -0.22     1.00   -0.14         0.11
## AltMed            0.22   -0.42   -0.40     0.23   -0.14     1.00         0.10
## SumaEspecies       0.35    0.23    0.23    -0.05    0.11    0.10         1.00
```

```
pairs(DatosModelo[,c(6,15,16,17,26,24,25)])
```



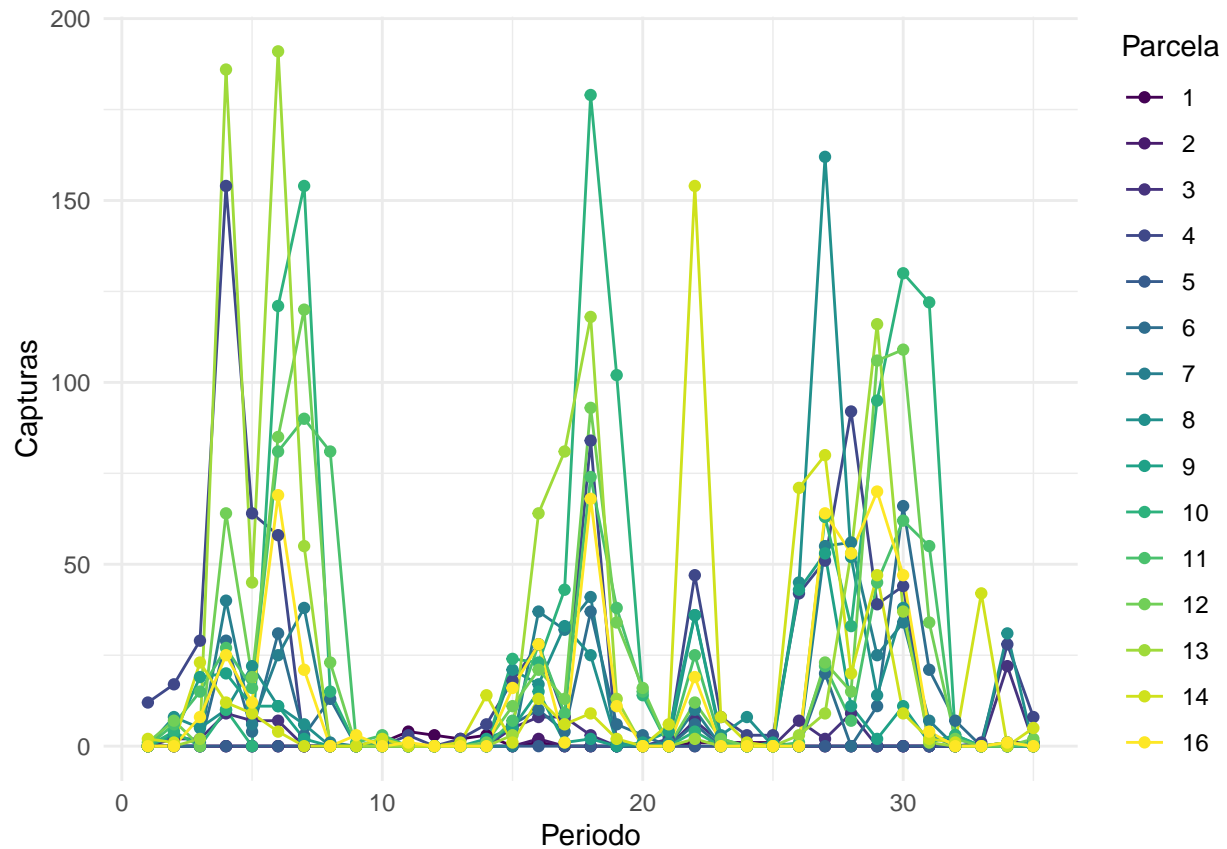
```
DatosModelo$logCap <- log(DatosModelo$Captures)
pairs(DatosModelo[,c(27,15,17,24,25)])
```



Serie temporal de las capturas de nuestra especie de interés:

```
ggplot(DatosModelo, aes(x = Periodo, y = Capturas, color = as.factor(Plot))) +
  geom_line() +
  geom_point() +
  labs(x = "Periodo", y = "Capturas", color = "Parcela") +
  theme_minimal()
```





## Modelos

```
set.seed(22)

Modelo <- function(){
  for (i in 1:14){
    for (j in 1:35){
      Captures[i,j] ~ dpois(lambda[i,j])
      log(lambda[i,j]) <- beta0 + v[i] + g[j]
    }
    v[i] ~ dnorm(0,tauv)
  }
  g[1] ~ dnorm(0,taug)
  for (j in 2:35){
    g[j] ~ dnorm(g[j-1],taug)
  }
  taug <- 1/pow(sdg,2)
  sdg ~ dunif(0,10)
  tauv <- 1/pow(sdv,2)
  sdv ~ dunif(0,10)

  #distribuciones iniciales
  beta0 ~ dflat()
}
```

```

}

#Datos
Datos <- list(Captures = matrix(DatosModeloSIN5$Captures, nrow = 14))

#Iniciales
Iniciales <- function(){
  list(beta0 = rnorm(1), v = rnorm(14), g = rnorm(35),
        sdv = runif(1,0,6), sdv = runif(1,0,3))
}

#Parámetros
Param <- c("beta0", "v", "sdv", "g", "sdg")

```

ResulNulo

```

## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/Rtmp8mEBar/model48446a132d.txt", fit v
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
## n.sims = 1002 iterations saved
##          mean   sd   2.5%   25%   50%   75%   97.5% Rhat n.eff
## beta0      -0.3 0.8   -1.6   -0.9   -0.4    0.2    1.2  1.7     6
## v[1]        0.2 0.1   -0.1    0.1    0.1    0.2    0.4  1.0   250
## v[2]        0.0 0.1   -0.2   -0.1    0.0    0.1    0.2  1.0   210
## v[3]       -0.4 0.1   -0.6   -0.5   -0.4   -0.3   -0.2  1.0   230
## v[4]        0.0 0.1   -0.2    0.0    0.0    0.1    0.3  1.0   200
## v[5]       -0.4 0.1   -0.6   -0.4   -0.4   -0.3   -0.1  1.0   200
## v[6]        0.2 0.1    0.0    0.1    0.2    0.3    0.4  1.0   250
## v[7]        0.2 0.1    0.0    0.1    0.2    0.2    0.4  1.0   250
## v[8]        0.3 0.1    0.0    0.2    0.3    0.3    0.5  1.0   190
## v[9]        0.0 0.1   -0.2   -0.1    0.0    0.1    0.2  1.0   220
## v[10]       -0.5 0.1   -0.7   -0.6   -0.5   -0.4   -0.3  1.0   290
## v[11]        0.6 0.1    0.3    0.5    0.6    0.6    0.8  1.0   170
## v[12]       -0.3 0.1   -0.6   -0.4   -0.3   -0.3   -0.1  1.0   180
## v[13]        0.5 0.1    0.3    0.5    0.5    0.6    0.8  1.0   280
## v[14]       -0.3 0.1   -0.5   -0.3   -0.3   -0.2    0.0  1.0   240
## sdv         0.4 0.1    0.3    0.3    0.4    0.4    0.6  1.0  1000
## g[1]        0.0 0.7   -1.4   -0.5    0.1    0.6    1.3  1.6     7
## g[2]       -0.9 0.9   -2.6   -1.5   -0.8   -0.2    0.6  1.5     8
## g[3]       -1.6 0.9   -3.5   -2.3   -1.6   -0.9    0.1  1.4     8
## g[4]       -1.8 0.9   -3.7   -2.4   -1.7   -1.1    0.0  1.4     8
## g[5]       -1.8 0.9   -3.7   -2.4   -1.7   -1.1    0.0  1.3    10
## g[6]        0.9 0.8   -0.6    0.3    0.9    1.4    2.2  1.6     6
## g[7]        1.5 0.8    0.0    1.0    1.6    2.1    2.8  1.7     6
## g[8]        3.5 0.7    2.0    2.9    3.5    4.0    4.8  1.7     6
## g[9]        2.6 0.8    1.1    2.1    2.7    3.2    3.9  1.7     6
## g[10]       3.5 0.7    2.0    3.0    3.6    4.1    4.8  1.7     6
## g[11]       2.0 0.8    0.5    1.5    2.0    2.6    3.3  1.6     7
## g[12]       2.1 0.7    0.6    1.6    2.2    2.7    3.4  1.7     7
## g[13]       3.0 0.8    1.5    2.4    3.0    3.6    4.2  1.7     6
## g[14]       2.3 0.7    0.8    1.8    2.4    2.9    3.6  1.7     6
## g[15]       2.8 0.8    1.3    2.3    2.9    3.4    4.1  1.7     6
## g[16]       1.8 0.8    0.3    1.2    1.8    2.3    3.1  1.7     6
## g[17]       3.6 0.7    2.1    3.1    3.6    4.2    4.9  1.7     6

```

```
## g[18]      2.6  0.8    1.1    2.1    2.7    3.2    3.9  1.7    6
## g[19]      1.1  0.8   -0.4    0.6    1.2    1.7    2.4  1.7    6
## g[20]      2.7  0.8    1.2    2.2    2.7    3.3    4.0  1.7    6
## g[21]      3.3  0.7    1.8    2.8    3.4    3.9    4.6  1.7    6
## g[22]      3.8  0.7    2.3    3.3    3.8    4.4    5.0  1.7    6
## g[23]      4.0  0.7    2.5    3.4    4.0    4.5    5.3  1.7    6
## g[24]      3.1  0.7    1.7    2.6    3.2    3.7    4.4  1.7    6
## g[25]      3.0  0.8    1.5    2.5    3.0    3.6    4.3  1.7    6
## g[26]      3.4  0.8    1.9    2.8    3.4    3.9    4.6  1.7    6
## g[27]      3.1  0.7    1.6    2.6    3.1    3.6    4.3  1.7    6
## g[28]      4.2  0.7    2.7    3.7    4.2    4.8    5.5  1.7    6
## g[29]      3.2  0.8    1.7    2.7    3.3    3.8    4.5  1.7    6
## g[30]      3.0  0.8    1.5    2.5    3.0    3.6    4.3  1.7    6
## g[31]      1.8  0.8    0.3    1.3    1.9    2.4    3.2  1.7    6
## g[32]      3.5  0.8    2.0    3.0    3.6    4.1    4.8  1.7    6
## g[33]      3.1  0.7    1.6    2.6    3.1    3.7    4.3  1.7    6
## g[34]      2.5  0.8    0.9    1.9    2.5    3.0    3.7  1.7    6
## g[35]      3.1  0.7    1.7    2.6    3.2    3.7    4.4  1.7    6
## sdg        1.1  0.2    0.9    1.0    1.1    1.2    1.5  1.0  1000
## deviance 14030.7 10.0 14010.0 14020.0 14030.0 14040.0 14050.0 1.0 1000
##
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 46.4 and DIC = 14077.0
## DIC is an estimate of expected predictive error (lower deviance is better).

#Rhat
summary((ResulNulo$summary[, "Rhat"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.000   1.010   1.680   1.457   1.717   1.733
```

El modelo no converge.

```
set.seed(22)

Modelo <- function(){
  for (i in 1:14){
    for (j in 1:35){
      Captures[i,j] ~ dpois(lambda[i,j])
      log(lambda[i,j]) <- beta0 + beta1*TMin[i,j] + beta2*Precipita[i,j] +
        beta3*AltMed[i,j] + beta4*Especies[i,j] + beta5*TDif[i,j] + v[i] + g[j]
    }
    v[i] ~ dnorm(0,tauv)
  }
  g[1] ~ dnorm(0,taug)
  for (j in 2:35){
    g[j] ~ dnorm(g[j-1],taug)
  }
  taug <- 1/pow(sdg,2)
  sdg ~ dunif(0,10)
```

```

    tauv <- 1/pow(sdv,2)
    sdv ~ dunif(0,10)

    #distribuciones iniciales
    beta0 ~ dflat()
    beta1 ~ dflat()
    beta2 ~ dflat()
    beta3 ~ dflat()
    beta4 ~ dflat()
    beta5 ~ dflat()
  }

  #Datos
  Datos <- list(Captures = matrix(DatosModeloSIN5$Captures, nrow = 14),
    TMin = matrix(DatosModeloSIN5$TMinMed, nrow = 14),
    Precipita = matrix(DatosModeloSIN5$Precipita, nrow = 14),
    AltMed = matrix(scale(DatosModeloSIN5$AltMed), nrow = 14),
    Especies = matrix(scale(DatosModeloSIN5$SumaEspecies), nrow = 14),
    TDif = matrix(scale(DatosModeloSIN5$TempDif), nrow = 14))

  #Iniciales
  Iniciales <- function(){
    list(beta0 = rnorm(1), v =rnorm(14), sdv = runif(1,0,3), g = rnorm(35), sdg = runif(1,0,6), beta1 = rnorm(1))
  }

  #Parámetros
  Param <- c("beta0","beta1","beta2","beta3","beta4", "v", "sdv", "g", "sdg")

```

Resul1

```

## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/RtmpaI8gjbv/model6ac6f9d67f8.txt", fit
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
## n.sims = 1002 iterations saved
##
##          mean    sd   2.5%   25%   50%   75%   97.5% Rhat n.eff
## beta0      -0.5  2.7   -7.5   -1.6    0.7    1.4    2.3  1.4    9
## beta1       0.0  0.0    0.0    0.0    0.0    0.0    0.0  1.0 1000
## beta2       0.1  0.0    0.1    0.1    0.1    0.2    0.2  1.0  400
## beta3       0.5  0.1    0.3    0.4    0.5    0.5    0.6  1.0   63
## beta4       0.4  0.0    0.3    0.3    0.4    0.4    0.4  1.0 1000
## v[1]        0.4  1.1   -0.2   -0.1    0.0    0.1    3.9  1.2   30
## v[2]         0.0  1.1   -0.6   -0.4   -0.4   -0.3    3.5  1.2   31
## v[3]         0.0  1.1   -0.6   -0.5   -0.4   -0.3    3.5  1.2   30
## v[4]       -0.1  1.1   -0.7   -0.5   -0.5   -0.3    3.4  1.2   30
## v[5]         0.2  1.1   -0.4   -0.2   -0.2   -0.1    3.7  1.2   31
## v[6]         0.6  1.1    0.0    0.1    0.2    0.3    4.1  1.2   31
## v[7]         0.5  1.1    0.0    0.1    0.2    0.3    4.1  1.2   30
## v[8]         0.4  1.1   -0.2   -0.1    0.0    0.1    3.9  1.2   30
## v[9]         0.3  1.1   -0.3   -0.2   -0.1    0.0    3.8  1.2   30
## v[10]        0.2  1.1   -0.4   -0.3   -0.2   -0.1    3.7  1.2   30
## v[11]        1.1  1.1    0.5    0.6    0.7    0.8    4.7  1.1   28
## v[12]        0.6  1.1    0.0    0.1    0.2    0.3    4.1  1.2   31
## v[13]        0.9  1.1    0.4    0.5    0.6    0.7    4.5  1.1   29
## v[14]        0.3  1.1   -0.3   -0.2   -0.1    0.0    3.9  1.2   30

```

```

## sdv      0.8  1.1    0.3    0.3    0.4    0.5    4.4  1.1   32
## g[1]     1.0  1.9   -1.3   -0.4    0.3    2.0    5.1  1.5    8
## g[2]     0.0  2.0   -2.5   -1.4   -0.6    1.1    4.3  1.5    8
## g[3]    -0.9  2.0   -3.7   -2.4   -1.5    0.3    3.2  1.5    8
## g[4]    -1.3  2.0   -4.3   -2.8   -1.8    0.0    3.0  1.5    8
## g[5]    -1.7  2.0   -4.4   -3.2   -2.2   -0.4    2.5  1.4    8
## g[6]     0.5  2.0   -1.9   -1.0   -0.2    1.7    4.8  1.5    7
## g[7]     1.1  2.0   -1.3   -0.4    0.4    2.3    5.4  1.5    8
## g[8]     3.0  2.0    0.6    1.5    2.2    4.1    7.2  1.7    6
## g[9]     2.4  2.0    0.0    0.8    1.6    3.5    6.5  1.5    8
## g[10]    3.4  2.0    1.1    1.9    2.7    4.5    7.6  1.7    6
## g[11]    1.6  2.1   -0.9    0.1    0.8    2.7    5.9  1.5    8
## g[12]    1.4  2.1   -1.0   -0.1    0.6    2.6    5.8  1.5    8
## g[13]    1.8  2.0   -0.6    0.3    1.1    2.9    6.1  1.5    7
## g[14]    1.0  2.0   -1.5   -0.6    0.2    2.1    5.2  1.5    8
## g[15]    1.6  2.0   -0.9    0.0    0.8    2.7    5.8  1.5    8
## g[16]    0.7  2.0   -1.7   -0.8    0.0    1.8    4.9  1.5    7
## g[17]    2.3  2.0   -0.1    0.7    1.5    3.4    6.5  1.5    8
## g[18]    1.6  2.0   -0.8    0.1    0.8    2.7    5.8  1.5    8
## g[19]    2.0  2.0   -0.3    0.5    1.3    3.1    6.2  1.5    8
## g[20]    3.3  2.0    1.0    1.8    2.6    4.5    7.5  1.7    6
## g[21]    2.9  2.0    0.5    1.3    2.1    3.9    7.1  1.5    7
## g[22]    3.5  2.0    1.1    1.9    2.7    4.6    7.7  1.7    6
## g[23]    3.3  2.0    0.9    1.8    2.5    4.4    7.6  1.7    6
## g[24]    2.8  2.0    0.4    1.3    2.0    3.9    7.0  1.5    8
## g[25]    2.7  2.0    0.4    1.2    2.0    3.9    7.0  1.5    8
## g[26]    2.6  2.0    0.2    1.0    1.8    3.7    6.8  1.5    7
## g[27]    2.7  2.0    0.3    1.2    2.0    3.8    7.0  1.5    7
## g[28]    3.6  2.0    1.2    2.1    2.8    4.7    7.9  1.6    6
## g[29]    3.2  2.0    0.8    1.6    2.4    4.4    7.4  1.7    6
## g[30]    2.3  2.0   -0.1    0.7    1.5    3.4    6.5  1.5    8
## g[31]    1.0  2.0   -1.4   -0.5    0.3    2.1    5.2  1.5    8
## g[32]    2.9  2.0    0.5    1.4    2.1    4.0    7.0  1.7    6
## g[33]    2.8  2.0    0.4    1.3    2.0    3.9    7.0  1.5    7
## g[34]    2.7  2.0    0.4    1.2    2.0    3.8    6.9  1.5    8
## g[35]    3.3  2.0    0.9    1.8    2.5    4.4    7.5  1.7    6
## sdg      1.1  0.2    0.8    0.9    1.0    1.2    1.5  1.1   42
## deviance 10558.3 10.6 10540.0 10550.0 10560.0 10560.0 10580.0 1.0 1000
##
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 50.1 and DIC = 10608.5
## DIC is an estimate of expected predictive error (lower deviance is better).

#Rhat
summary((Resul1$summary[, "Rhat"]))

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.9996  1.1637  1.5003  1.3829  1.5054  1.6817

```

El modelo no converge.

```

set.seed(22)

Modelo <- function(){
  for (i in 1:14){
    for (j in 1:35){
      Captures[i,j] ~ dpois(lambda[i,j])
      log(lambda[i,j]) <- beta0 + beta1*TMin[i,j] + beta2*Precipita[i,j] +
        beta3*AltMed[i,j] + beta4*Especies[i,j] + beta5*TDif[i,j] + v[i]
    }
    v[i] ~ dnorm(0,tauv)
  }
  tauv <- 1/pow(sdv,2)
  sdv ~ dunif(0,10)

  #distribuciones iniciales
  beta0 ~ dflat()
  beta1 ~ dflat()
  beta2 ~ dflat()
  beta3 ~ dflat()
  beta4 ~ dflat()
  beta5 ~ dflat()
}

#Datos
Datos <- list(Captures = matrix(DatosModeloSIN5$Captures,
                                nrow = 14),
              TMin = matrix(DatosModeloSIN5$TMinMed,
                             nrow = 14),
              Precipita = matrix(DatosModeloSIN5$Precipita,
                                  nrow = 14),
              AltMed = matrix(scale(DatosModeloSIN5$AltMed),
                               nrow = 14),
              Especies = matrix(scale(DatosModeloSIN5$SumaEspecies),
                                  nrow = 14),
              TDif = matrix(scale(DatosModeloSIN5$TempDif),
                              nrow = 14))

#Iniciales
Iniciales <- function(){
  list(beta0 = rnorm(1), v = rnorm(14), sdv = runif(1,0,3),
        beta1 = rnorm(1), beta2 = rnorm(1), beta3 = rnorm(1),
        beta4 = rnorm(1), beta5 = rnorm(1))
}

#Parámetros
Param <- c("beta0","beta1","beta2","beta3","beta4", "beta5", "v", "sdv")

```

Resul2

```

## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/RtmpAXk7nM/model15d86f9258a9.txt", fi
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
## n.sims = 1002 iterations saved
##           mean    sd   2.5%   25%   50%   75%   97.5% Rhat n.eff

```

```
## beta0      -5.3 10.6   -25.2   -16.8     2.0     2.1     2.2  9.3     3
## beta1       0.0  0.0     0.0     0.0     0.0     0.0     0.0  1.0  1000
## beta2       0.2  0.0     0.2     0.2     0.2     0.2     0.2  1.0  1000
## beta3       0.4  0.0     0.4     0.4     0.4     0.4     0.5  1.0  1000
## beta4       0.4  0.0     0.4     0.4     0.4     0.4     0.4  1.0  1000
## beta5       0.5  0.0     0.4     0.4     0.5     0.5     0.5  1.0  1000
## v[1]        7.6 10.6     0.0     0.2     0.3    19.0    27.4  9.3     3
## v[2]        7.1 10.6    -0.4    -0.3    -0.2    18.6    27.0  9.3     3
## v[3]        7.1 10.6    -0.5    -0.3    -0.2    18.5    26.9  9.3     3
## v[4]        7.1 10.6    -0.5    -0.3    -0.2    18.5    26.9  9.3     3
## v[5]        7.3 10.6    -0.3    -0.1     0.0    18.7    27.1  9.3     3
## v[6]        7.4 10.6    -0.2     0.0     0.1    18.8    27.2  9.3     3
## v[7]        7.5 10.6    -0.1     0.1     0.2    18.9    27.3  9.3     3
## v[8]        7.2 10.6    -0.3    -0.2     0.0    18.6    27.0  9.3     3
## v[9]        7.0 10.6    -0.5    -0.3    -0.2    18.5    26.9  9.3     3
## v[10]       7.0 10.6    -0.6    -0.4    -0.3    18.4    26.8  9.3     3
## v[11]       8.1 10.6     0.5     0.7     0.8    19.5    27.9 17.0     3
## v[12]       7.5 10.6    -0.1     0.1     0.2    18.9    27.4  9.3     3
## v[13]       7.9 10.6     0.4     0.5     0.6    19.4    27.8  9.3     3
## v[14]       7.3 10.6    -0.3    -0.1     0.0    18.7    27.1  9.3     3
## sdv         3.5  4.4     0.2     0.3     0.4     9.7    10.0 16.6     3
## deviance 14309.3   7.1 14300.0 14300.0 14310.0 14310.0 14320.0  1.0   430
##
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 18.6 and DIC = 14327.8
## DIC is an estimate of expected predictive error (lower deviance is better).
```

```
#Rhat
summary(Rhat <- (Resul2$summary[, "Rhat"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.000   3.070   9.276   7.708   9.281  17.026
```

El modelo no converge.

```
set.seed(22)

Modelo <- function(){
  for (i in 1:14){
    for (j in 1:35){
      Captures[i,j] ~ dpois(lambda[i,j])
      log(lambda[i,j]) <- beta0 + beta1*TMin[i,j] + beta2*Precipita[i,j] +
        beta3*AltMed[i,j] + beta4*Especies[i,j] + v[i] + g[j]
    }
    v[i] ~ dnorm(0,tauv)
  }
  g[1] ~ dnorm(0,taug)
  for (j in 2:35){
    g[j] ~ dnorm(g[j-1],taug)
  }
}
```

```

taug <- 1/pow(sdg,2)
sdg ~ dunif(0,10)
tauv <- 1/pow(sdv,2)
sdv ~ dunif(0,10)

#distribuciones iniciales
beta0 ~ dflat()
beta1 ~ dflat()
beta2 ~ dflat()
beta3 ~ dflat()
beta4 ~ dflat()
}

#Datos
Datos <- list(Captures = matrix(DatosModeloSIN5$Captures,
                                nrow = 14),
              TMin = matrix(DatosModeloSIN5$TMinMed,
                             nrow = 14),
              Precipita = matrix(DatosModeloSIN5$Precipita,
                                  nrow = 14),
              AltMed = matrix(scale(DatosModeloSIN5$AltMed),
                               nrow = 14),
              Especies = matrix(scale(DatosModeloSIN5$SumaEspecies),
                                  nrow = 14))

#Iniciales
Iniciales <- function(){
  list(beta0 = rnorm(1), v = rnorm(14), sdv = runif(1,0,3),
        g = rnorm(35), sdg = runif(1,0,6), beta1 = rnorm(1), beta2 = rnorm(1),
        beta3 = rnorm(1), beta4 = rnorm(1))
}

#Parámetros
Param <- c("beta0","beta1","beta2","beta3","beta4", "v", "sdv", "g", "sdg")

```

Resul3

```

## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/RtmpAXk7nM/model15d8706c2c28.txt", fi
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
## n.sims = 1002 iterations saved
##          mean   sd   2.5%   25%   50%   75%   97.5% Rhat n.eff
## beta0      -0.9  0.9   -2.3   -1.5   -0.9   -0.3    1.2  1.7     6
## beta1       0.0  0.0    0.0    0.0    0.0    0.0    0.0  1.0   1000
## beta2       0.1  0.0    0.1    0.1    0.1    0.1    0.1  1.0   1000
## beta3       0.4  0.1    0.2    0.3    0.4    0.4    0.5  1.0    510
## beta4       0.4  0.0    0.3    0.3    0.4    0.4    0.4  1.0   1000
## v[1]        0.4  0.1    0.1    0.3    0.4    0.4    0.6  1.0    380
## v[2]       -0.2  0.1   -0.5   -0.3   -0.2   -0.2    0.0  1.0    560
## v[3]       -0.3  0.1   -0.6   -0.4   -0.3   -0.3   -0.1  1.0    390
## v[4]       -0.6  0.1   -0.9   -0.7   -0.6   -0.6   -0.4  1.0    770
## v[5]       -0.3  0.1   -0.5   -0.3   -0.3   -0.2    0.0  1.0    600
## v[6]        0.1  0.1   -0.1    0.1    0.1    0.2    0.4  1.0    490
## v[7]        0.2  0.1   -0.1    0.1    0.2    0.3    0.4  1.0    490

```



```

## v[8]      0.2 0.1 0.0 0.1 0.2 0.3 0.4 1.0 670
## v[9]     -0.2 0.1 -0.4 -0.3 -0.2 -0.1 0.0 1.0 650
## v[10]    -0.3 0.1 -0.5 -0.4 -0.3 -0.2 0.0 1.0 450
## v[11]     0.7 0.1 0.4 0.6 0.7 0.7 0.9 1.0 350
## v[12]     0.0 0.1 -0.2 -0.1 0.0 0.1 0.3 1.0 390
## v[13]     0.5 0.1 0.3 0.5 0.5 0.6 0.8 1.0 360
## v[14]    -0.1 0.1 -0.3 -0.2 -0.1 0.0 0.2 1.0 200
## sdv      0.4 0.1 0.3 0.3 0.4 0.5 0.6 1.0 1000
## g[1]      0.9 0.8 -1.1 0.4 0.8 1.5 2.4 1.6 7
## g[2]      0.0 1.0 -2.0 -0.6 0.0 0.6 1.8 1.5 7
## g[3]     -0.9 1.0 -3.1 -1.5 -0.9 -0.2 1.0 1.5 8
## g[4]     -1.1 1.0 -3.4 -1.8 -1.1 -0.4 0.8 1.5 8
## g[5]     -1.4 1.0 -3.5 -2.0 -1.3 -0.7 0.5 1.5 8
## g[6]      1.2 0.9 -0.9 0.6 1.2 1.9 2.7 1.7 6
## g[7]      1.8 0.9 -0.3 1.3 1.9 2.5 3.3 1.7 6
## g[8]      3.6 0.9 1.5 3.1 3.6 4.3 5.1 1.7 6
## g[9]      2.8 0.9 0.6 2.2 2.8 3.4 4.2 1.6 7
## g[10]     3.7 0.9 1.5 3.1 3.7 4.3 5.1 1.7 6
## g[11]     1.8 0.9 -0.5 1.2 1.8 2.4 3.3 1.7 6
## g[12]     1.9 0.9 -0.3 1.3 1.9 2.6 3.4 1.7 6
## g[13]     2.6 0.9 0.4 2.0 2.5 3.2 4.1 1.7 6
## g[14]     2.0 0.9 -0.1 1.5 2.0 2.7 3.6 1.7 6
## g[15]     2.6 0.9 0.4 2.0 2.6 3.2 4.0 1.7 6
## g[16]     2.1 0.9 0.0 1.6 2.1 2.7 3.6 1.7 6
## g[17]     3.9 0.9 1.7 3.3 3.9 4.5 5.3 1.7 6
## g[18]     3.0 0.9 0.9 2.5 3.0 3.6 4.5 1.7 7
## g[19]     1.7 0.9 -0.3 1.1 1.7 2.3 3.3 1.7 6
## g[20]     3.2 0.9 1.1 2.6 3.2 3.8 4.7 1.7 6
## g[21]     2.7 0.9 0.5 2.2 2.7 3.4 4.2 1.6 7
## g[22]     3.5 0.9 1.3 2.9 3.5 4.1 5.0 1.7 6
## g[23]     3.4 0.9 1.2 2.8 3.4 4.0 4.9 1.7 7
## g[24]     2.9 0.9 0.7 2.3 2.9 3.5 4.3 1.6 7
## g[25]     2.8 0.9 0.6 2.3 2.8 3.5 4.3 1.6 7
## g[26]     2.3 0.9 0.1 1.7 2.3 2.9 3.8 1.7 6
## g[27]     2.6 0.9 0.4 2.1 2.6 3.3 4.1 1.7 6
## g[28]     3.5 0.9 1.3 3.0 3.5 4.2 5.0 1.7 6
## g[29]     3.0 0.9 0.8 2.5 3.0 3.6 4.5 1.7 7
## g[30]     2.3 0.9 0.1 1.8 2.3 3.0 3.8 1.7 6
## g[31]     1.6 0.9 -0.4 1.1 1.6 2.2 3.1 1.7 6
## g[32]     3.8 0.9 1.8 3.3 3.9 4.5 5.3 1.7 6
## g[33]     3.3 0.9 1.2 2.7 3.3 3.9 4.7 1.7 6
## g[34]     2.7 0.9 0.5 2.1 2.7 3.3 4.1 1.7 6
## g[35]     3.2 0.9 1.1 2.7 3.3 3.9 4.7 1.7 7
## sdg      1.1 0.2 0.8 1.0 1.1 1.2 1.4 1.0 1000
## deviance 12028.2 10.6 12010.0 12020.0 12030.0 12030.0 12050.0 1.0 340
##
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 49.7 and DIC = 12078.1
## DIC is an estimate of expected predictive error (lower deviance is better).

```

```
#Rhat
summary(Rhat <- (Resul3$summary[, "Rhat"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.9996  1.0073  1.6252  1.4206  1.6956  1.7400
```

El modelo no converge.

```
set.seed(22)
```

```
Modelo <- function(){
  for (i in 1:14){
    for (j in 1:35){
      Captures[i,j] ~ dpois(lambda[i,j])
      log(lambda[i,j]) <- beta0 + beta1*TMin[i,j] + beta2*Precipita[i,j] +
        beta3*Especies[i,j] + v[i]
    }
    v[i] ~ dnorm(0,tauv)
  }
  tauv <- 1/pow(sdv,2)
  sdv ~ dunif(0,10)
```

```
#distribuciones iniciales
```

```
beta0 ~ dflat()
beta1 ~ dflat()
beta2 ~ dflat()
beta3 ~ dflat()
```

```
  for (i in 1:14) {
    for (j in 1:35) {
      Captures.pred[i,j] ~ dpois(lambda[i,j])
      resid[i,j] <- pow(Captures[i,j] - Captures.pred[i,j],2)
    }
  }
}
```

```
#Datos
```

```
Datos <- list(Captures = matrix(DatosModeloSIN5$Captures,
                                nrow = 14),
              TMin = matrix(DatosModeloSIN5$TMinMed,
                             nrow = 14),
              Precipita = matrix(DatosModeloSIN5$Precipita,
                                  nrow = 14),
              Especies = matrix(scale(DatosModeloSIN5$SumaEspecies),
                                 nrow = 14))
```

```
#Iniciales
```

```
Iniciales <- function(){
  list(beta0 = rnorm(1), v = rnorm(14), sdv = runif(1,0,3),
        beta1 = rnorm(1), beta2 = rnorm(1), beta3 = rnorm(1))
}
```

```
#Parámetros
```

```
Param <- c("beta0","beta1","beta2","beta3", "v", "sdv", "resid")
```

```
Resul4
```

```
## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/RtmpAXk7nM/model15d817bc66db.txt", fi
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
## n.sims = 1002 iterations saved
```

##	mean	sd	2.5%	25%	50%	75%	97.5%	Rhat	n.eff
## beta0	2.1	0.1	1.9	2.0	2.1	2.2	2.3	1	1000
## beta1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1	1000
## beta2	0.2	0.0	0.2	0.2	0.2	0.2	0.2	1	800
## beta3	0.4	0.0	0.4	0.4	0.4	0.4	0.4	1	1000
## v[1]	0.4	0.1	0.2	0.4	0.4	0.5	0.6	1	1000
## v[2]	-0.2	0.1	-0.4	-0.2	-0.2	-0.1	0.1	1	1000
## v[3]	-0.2	0.1	-0.4	-0.3	-0.2	-0.1	0.0	1	1000
## v[4]	-0.3	0.1	-0.5	-0.4	-0.3	-0.2	-0.1	1	1000
## v[5]	-0.1	0.1	-0.3	-0.2	-0.1	0.0	0.1	1	1000
## v[6]	-0.1	0.1	-0.3	-0.1	-0.1	0.0	0.2	1	1000
## v[7]	0.2	0.1	0.0	0.1	0.2	0.3	0.4	1	1000
## v[8]	0.0	0.1	-0.2	-0.1	0.0	0.0	0.2	1	1000
## v[9]	-0.4	0.1	-0.6	-0.5	-0.4	-0.3	-0.2	1	1000
## v[10]	-0.4	0.1	-0.6	-0.4	-0.4	-0.3	-0.1	1	1000
## v[11]	0.7	0.1	0.4	0.6	0.7	0.7	0.9	1	1000
## v[12]	0.0	0.1	-0.2	0.0	0.0	0.1	0.3	1	1000
## v[13]	0.5	0.1	0.3	0.5	0.5	0.6	0.7	1	1000
## v[14]	-0.1	0.1	-0.3	-0.2	-0.1	0.0	0.1	1	1000
## sdv	0.4	0.1	0.2	0.3	0.4	0.4	0.6	1	1000
## resid[1,1]	142.5	84.3	25.0	81.0	121.0	196.0	324.0	1	1000
## resid[1,2]	149.0	88.6	25.0	81.0	144.0	196.0	361.0	1	1000
## resid[1,3]	288.6	141.3	81.0	196.0	256.0	361.0	625.0	1	1000
## resid[1,4]	287.8	141.2	81.0	196.0	256.0	361.0	625.0	1	890
## resid[1,5]	215.1	113.9	49.0	144.0	196.0	289.0	484.0	1	1000
## resid[1,6]	135.6	78.7	36.0	81.0	121.0	169.0	324.0	1	870
## resid[1,7]	65.5	54.4	1.0	25.0	49.0	100.0	196.0	1	490
## resid[1,8]	176.8	96.2	36.0	100.0	169.0	225.0	400.0	1	800
## resid[1,9]	619.5	257.6	225.0	441.0	576.0	784.0	1225.0	1	640
## resid[1,10]	1111.8	249.3	676.0	961.0	1089.0	1296.0	1600.0	1	1000
## resid[1,11]	162.1	91.6	36.0	100.0	144.0	196.0	361.0	1	1000
## resid[1,12]	77.4	64.4	4.0	25.0	64.0	100.0	225.0	1	310
## resid[1,13]	128.2	107.6	4.0	49.0	100.0	169.0	361.0	1	1000
## resid[1,14]	1519.4	512.9	676.0	1156.0	1444.0	1849.0	2601.0	1	910
## resid[1,15]	53.8	59.6	0.0	9.0	36.0	81.0	196.0	1	280
## resid[1,16]	149.7	86.9	25.0	81.0	144.0	196.0	361.0	1	1000
## resid[1,17]	98.7	64.2	4.0	49.0	81.0	144.0	255.2	1	580
## resid[1,18]	15.4	23.6	0.0	1.0	9.0	16.0	81.0	1	340
## resid[1,19]	256.0	130.0	81.0	169.0	225.0	324.0	576.0	1	1000
## resid[1,20]	575.2	165.3	289.0	441.0	576.0	676.0	900.0	1	1000
## resid[1,21]	158.4	90.2	36.0	100.0	144.0	196.0	400.0	1	580
## resid[1,22]	133.4	76.5	9.0	81.0	121.0	169.0	289.0	1	320
## resid[1,23]	6391.2	616.2	5184.0	5929.0	6400.0	6889.0	7569.0	1	1000
## resid[1,24]	1749.0	541.8	784.0	1369.0	1681.0	2116.0	2916.0	1	1000
## resid[1,25]	107.5	74.5	4.0	49.0	100.0	144.0	289.0	1	420

## resid[1,26]	161.1	95.7	36.0	81.0	144.0	217.4	400.0	1	420
## resid[1,27]	19.7	28.2	0.0	1.0	9.0	25.0	100.0	1	1000
## resid[1,28]	7579.9	770.0	6084.0	7056.0	7569.0	8100.0	9025.0	1	1000
## resid[1,29]	384.3	178.7	121.0	256.0	361.0	484.0	784.0	1	1000
## resid[1,30]	129.1	84.7	16.0	64.0	121.0	169.0	324.0	1	370
## resid[1,31]	92.3	69.1	9.0	49.0	81.0	121.0	256.0	1	1000
## resid[1,32]	403.9	201.0	121.0	256.0	361.0	484.0	900.0	1	1000
## resid[1,33]	984.0	249.3	529.0	784.0	961.0	1156.0	1521.0	1	1000
## resid[1,34]	256.3	136.5	64.0	169.0	225.0	324.0	576.0	1	920
## resid[1,35]	63.0	49.0	1.0	25.0	49.0	100.0	169.0	1	210
## resid[2,1]	55.2	41.3	4.0	25.0	49.0	81.0	144.0	1	210
## resid[2,2]	54.1	43.4	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[2,3]	130.8	79.2	25.0	81.0	121.0	169.0	324.0	1	270
## resid[2,4]	58.7	41.4	9.0	25.0	49.0	81.0	169.0	1	560
## resid[2,5]	88.2	61.6	9.0	49.0	81.0	121.0	256.0	1	430
## resid[2,6]	47.1	40.9	1.0	16.0	36.0	64.0	168.4	1	1000
## resid[2,7]	8.3	11.4	0.0	1.0	4.0	9.0	36.0	1	210
## resid[2,8]	78.9	52.8	9.0	36.0	64.0	100.0	225.0	1	470
## resid[2,9]	138.6	82.4	25.0	81.0	121.0	169.0	324.0	1	440
## resid[2,10]	45.1	52.0	0.0	9.0	25.0	64.0	169.0	1	440
## resid[2,11]	85.6	55.5	9.0	49.0	81.0	121.0	225.0	1	930
## resid[2,12]	9.6	13.3	0.0	1.0	4.0	16.0	49.0	1	950
## resid[2,13]	947.1	376.1	289.8	676.0	900.0	1156.0	1764.0	1	1000
## resid[2,14]	71.0	47.6	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[2,15]	378.2	168.7	121.0	256.0	361.0	484.0	784.0	1	920
## resid[2,16]	7.9	11.6	0.0	1.0	4.0	9.0	36.0	1	570
## resid[2,17]	35.3	35.5	0.0	9.0	25.0	49.0	121.0	1	510
## resid[2,18]	669.1	176.8	324.0	529.0	676.0	784.0	1024.0	1	1000
## resid[2,19]	85.1	56.4	9.1	49.0	81.0	121.0	225.0	1	1000
## resid[2,20]	168.5	105.0	25.0	100.0	144.0	225.0	400.0	1	1000
## resid[2,21]	48.5	47.4	1.0	16.0	36.0	64.0	169.0	1	1000
## resid[2,22]	155.0	76.3	25.0	100.0	144.0	196.0	324.0	1	1000
## resid[2,23]	5774.1	1191.1	3600.0	4900.0	5776.0	6561.0	8100.0	1	1000
## resid[2,24]	354.9	174.0	100.0	225.0	324.0	441.0	729.0	1	1000
## resid[2,25]	589.1	240.6	196.0	441.0	576.0	729.0	1156.0	1	430
## resid[2,26]	19.3	27.1	0.0	4.0	9.0	25.0	99.5	1	1000
## resid[2,27]	66.6	51.9	1.0	25.0	49.0	100.0	196.0	1	800
## resid[2,28]	8255.0	749.7	6724.0	7744.0	8281.0	8836.0	9801.0	1	1000
## resid[2,29]	55.1	40.9	4.0	25.0	49.0	81.0	168.4	1	1000
## resid[2,30]	238.9	130.8	64.0	144.0	225.0	324.0	529.0	1	1000
## resid[2,31]	74.7	54.6	9.0	36.0	64.0	100.0	225.0	1	1000
## resid[2,32]	49.6	63.2	0.0	4.0	25.0	81.0	225.0	1	1000
## resid[2,33]	18.1	25.5	0.0	1.0	9.0	25.0	81.0	1	1000
## resid[2,34]	24.3	27.6	0.0	4.0	16.0	36.0	100.0	1	1000
## resid[2,35]	740.6	291.2	289.0	529.0	729.0	900.0	1442.1	1	1000
## resid[3,1]	78.8	51.9	9.0	36.0	64.0	100.0	225.0	1	870
## resid[3,2]	58.4	42.1	4.0	25.0	49.0	81.0	169.0	1	580
## resid[3,3]	60.6	43.8	9.0	25.0	49.0	81.0	169.0	1	560
## resid[3,4]	52.9	38.8	4.0	25.0	49.0	81.0	144.0	1	420
## resid[3,5]	62.3	44.5	9.0	25.0	49.0	81.0	169.0	1	1000
## resid[3,6]	49.8	43.6	1.0	16.0	36.0	64.0	169.0	1	1000
## resid[3,7]	10.1	11.4	0.0	1.0	4.0	16.0	36.0	1	1000
## resid[3,8]	61.8	43.5	9.0	36.0	49.0	81.0	169.0	1	1000
## resid[3,9]	110.8	68.3	16.2	64.0	100.0	144.0	289.0	1	1000

## resid[3,10]	24.4	27.3	0.0	4.0	16.0	36.0	100.0	1	1000
## resid[3,11]	137.4	85.1	25.0	81.0	121.0	196.0	361.0	1	970
## resid[3,12]	21.5	26.5	0.0	4.0	9.0	25.0	81.0	1	1000
## resid[3,13]	60.5	51.5	1.0	16.0	49.0	95.2	169.0	1	1000
## resid[3,14]	61.6	41.6	9.0	25.0	49.0	81.0	169.0	1	1000
## resid[3,15]	43.9	36.5	4.0	16.0	36.0	64.0	143.4	1	470
## resid[3,16]	33.1	35.1	0.0	9.0	25.0	49.0	121.0	1	1000
## resid[3,17]	624.0	136.2	361.0	529.0	625.0	729.0	900.0	1	450
## resid[3,18]	58.8	50.3	1.0	25.0	49.0	81.0	195.3	1	1000
## resid[3,19]	33.2	31.7	1.0	9.0	25.0	49.0	121.0	1	1000
## resid[3,20]	50.1	36.2	4.0	25.0	36.0	64.0	144.0	1	1000
## resid[3,21]	130.8	78.5	25.0	81.0	121.0	169.0	324.0	1	600
## resid[3,22]	1051.2	206.7	676.0	900.0	1024.0	1225.0	1444.0	1	1000
## resid[3,23]	8886.7	1003.6	6889.0	8100.0	8836.0	9604.0	10820.0	1	1000
## resid[3,24]	218.0	128.9	36.0	121.0	196.0	289.0	529.0	1	1000
## resid[3,25]	54.7	41.2	4.0	25.0	49.0	81.0	144.0	1	1000
## resid[3,26]	132.5	78.1	25.0	81.0	121.0	169.0	324.0	1	1000
## resid[3,27]	14.9	19.1	0.0	1.0	9.0	25.0	64.0	1	510
## resid[3,28]	496.0	150.6	196.0	400.0	484.0	612.4	784.0	1	1000
## resid[3,29]	53.5	39.9	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[3,30]	57.5	40.5	4.0	25.0	49.0	81.0	144.0	1	810
## resid[3,31]	200.0	84.5	49.0	144.0	196.0	256.0	399.0	1	1000
## resid[3,32]	18.2	26.2	0.0	1.0	9.0	25.0	100.0	1	640
## resid[3,33]	98.3	68.4	9.0	49.0	81.0	144.0	256.0	1	1000
## resid[3,34]	48.5	37.8	4.0	25.0	36.0	64.0	144.0	1	570
## resid[3,35]	55.8	41.8	4.0	25.0	49.0	81.0	168.4	1	780
## resid[4,1]	162.8	92.0	36.0	100.0	144.0	196.0	400.0	1	1000
## resid[4,2]	68.9	46.3	9.0	36.0	64.0	100.0	169.0	1	1000
## resid[4,3]	54.3	39.1	4.0	25.0	49.0	81.0	144.0	1	860
## resid[4,4]	42.1	35.0	4.0	16.0	36.0	64.0	121.0	1	1000
## resid[4,5]	39.6	31.2	4.0	16.0	36.0	49.0	121.0	1	520
## resid[4,6]	23.7	34.4	0.0	4.0	9.0	25.0	121.0	1	1000
## resid[4,7]	20.7	25.8	0.0	4.0	9.0	25.0	81.0	1	1000
## resid[4,8]	54.2	39.9	4.0	25.0	49.0	81.0	168.4	1	570
## resid[4,9]	33.9	34.2	0.0	9.0	25.0	49.0	121.0	1	1000
## resid[4,10]	12.9	17.7	0.0	1.0	9.0	16.0	64.0	1	640
## resid[4,11]	373.9	118.4	169.0	289.0	361.0	441.0	625.0	1	1000
## resid[4,12]	738.1	169.6	441.0	625.0	729.0	841.0	1087.3	1	1000
## resid[4,13]	13.6	21.4	0.0	1.0	4.0	16.0	81.0	1	650
## resid[4,14]	44.4	35.1	4.0	16.0	36.0	64.0	121.0	1	610
## resid[4,15]	41.0	30.6	4.0	16.0	36.0	49.0	100.0	1	1000
## resid[4,16]	9.8	13.0	0.0	1.0	4.0	16.0	49.0	1	930
## resid[4,17]	246.1	94.2	81.0	169.0	256.0	324.0	441.0	1	1000
## resid[4,18]	96.3	61.8	16.0	49.0	81.0	121.0	256.0	1	1000
## resid[4,19]	42.4	32.1	4.0	16.0	36.0	64.0	121.0	1	1000
## resid[4,20]	22.9	23.2	0.0	4.0	16.0	36.0	81.0	1	660
## resid[4,21]	34.0	43.6	0.0	4.0	16.0	49.0	169.0	1	990
## resid[4,22]	23350.0	1639.9	20160.0	22200.0	23410.0	24340.0	26570.0	1	1000
## resid[4,23]	1936.2	661.2	900.0	1444.0	1849.0	2304.0	3364.0	1	1000
## resid[4,24]	61.0	43.4	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[4,25]	40.7	33.6	4.0	16.0	36.0	64.0	121.0	1	800
## resid[4,26]	1930.8	381.9	1225.0	1681.0	1936.0	2209.0	2704.0	1	1000
## resid[4,27]	3711.1	720.9	2401.0	3249.0	3721.0	4225.0	5184.0	1	1000
## resid[4,28]	17289.2	3535.5	11240.0	14880.0	16900.0	19600.0	25280.0	1	1000

## resid[4,29]	105.1	66.2	16.0	64.0	81.0	144.0	256.0	1	1000
## resid[4,30]	41.9	30.8	4.0	16.0	36.0	64.0	121.0	1	380
## resid[4,31]	13.4	21.0	0.0	1.0	4.0	16.0	64.0	1	1000
## resid[4,32]	11.8	16.5	0.0	1.0	4.0	16.0	64.0	1	1000
## resid[4,33]	80.3	56.5	9.0	36.0	64.0	121.0	225.0	1	430
## resid[4,34]	37.4	32.0	1.0	16.0	25.0	49.0	121.0	1	1000
## resid[4,35]	42.9	35.3	4.0	16.0	36.0	64.0	121.0	1	1000
## resid[5,1]	95.6	67.1	16.0	49.0	81.0	121.0	256.0	1	1000
## resid[5,2]	63.2	45.0	9.0	36.0	49.0	81.0	169.0	1	1000
## resid[5,3]	147.2	89.5	25.0	81.0	121.0	196.0	361.0	1	1000
## resid[5,4]	70.6	48.6	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[5,5]	69.5	49.4	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[5,6]	9.0	13.9	0.0	1.0	4.0	9.0	49.0	1	390
## resid[5,7]	57.8	42.0	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[5,8]	207.1	114.7	49.0	121.0	196.0	256.0	484.0	1	1000
## resid[5,9]	62.1	45.7	9.0	25.0	49.0	81.0	196.0	1	1000
## resid[5,10]	1027.8	209.8	625.0	900.0	1024.0	1156.0	1444.0	1	1000
## resid[5,11]	25.8	30.2	0.0	4.0	16.0	36.0	121.0	1	1000
## resid[5,12]	12.7	18.6	0.0	1.0	4.0	16.0	64.0	1	1000
## resid[5,13]	107.4	66.4	16.0	64.0	100.0	144.0	289.0	1	1000
## resid[5,14]	56.0	40.8	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[5,15]	49.0	39.4	4.0	25.0	36.0	64.0	144.0	1	1000
## resid[5,16]	178.1	78.2	36.0	121.0	169.0	225.0	361.0	1	1000
## resid[5,17]	60.6	45.0	4.0	25.0	49.0	81.0	169.0	1	850
## resid[5,18]	108.7	66.4	25.0	64.0	100.0	144.0	289.0	1	1000
## resid[5,19]	48.8	36.5	4.0	25.0	36.0	64.0	144.0	1	460
## resid[5,20]	1339.8	181.7	961.0	1225.0	1369.0	1444.0	1681.0	1	1000
## resid[5,21]	89.9	60.5	9.0	49.0	81.0	121.0	225.0	1	790
## resid[5,22]	8119.7	623.6	6889.0	7744.0	8100.0	8649.0	9216.0	1	1000
## resid[5,23]	363.9	178.8	100.0	225.0	324.0	441.0	784.0	1	890
## resid[5,24]	48.7	38.4	4.0	25.0	36.0	64.0	144.0	1	680
## resid[5,25]	42.9	35.0	1.0	16.0	36.0	64.0	143.4	1	1000
## resid[5,26]	101.5	56.2	9.0	64.0	100.0	144.0	225.0	1	910
## resid[5,27]	382.0	152.2	100.5	289.0	361.0	484.0	729.0	1	310
## resid[5,28]	888.9	355.3	361.0	625.0	841.0	1089.0	1681.0	1	1000
## resid[5,29]	91.3	58.9	16.0	49.0	81.0	121.0	225.0	1	350
## resid[5,30]	41.7	39.9	1.0	16.0	36.0	64.0	144.0	1	1000
## resid[5,31]	9.2	11.7	0.0	1.0	4.0	16.0	49.0	1	640
## resid[5,32]	49.1	44.2	1.0	16.0	36.0	64.0	169.0	1	640
## resid[5,33]	852.2	209.4	484.0	676.0	841.0	1024.0	1296.0	1	980
## resid[5,34]	47.7	36.4	4.0	25.0	36.0	64.0	144.0	1	810
## resid[5,35]	57.2	41.2	4.0	25.0	49.0	81.0	169.0	1	560
## resid[6,1]	146.8	87.0	25.0	81.0	121.0	196.0	361.0	1	1000
## resid[6,2]	101.7	63.1	16.0	49.0	90.0	144.0	256.0	1	1000
## resid[6,3]	94.0	62.6	9.0	49.0	81.0	121.0	256.0	1	960
## resid[6,4]	66.8	47.5	9.0	36.0	64.0	81.0	169.0	1	1000
## resid[6,5]	139.0	79.0	25.0	81.0	121.0	196.0	324.0	1	670
## resid[6,6]	8.2	12.0	0.0	1.0	4.0	9.0	36.0	1	1000
## resid[6,7]	66.3	46.5	4.1	36.0	64.0	81.0	195.3	1	820
## resid[6,8]	200.0	81.2	64.0	144.0	196.0	256.0	361.0	1	1000
## resid[6,9]	35.6	32.3	1.0	16.0	25.0	49.0	121.0	1	1000
## resid[6,10]	1344.5	273.1	784.0	1156.0	1369.0	1521.0	1936.0	1	450
## resid[6,11]	338.1	128.1	100.0	256.0	324.0	400.0	625.0	1	830
## resid[6,12]	55.3	50.8	1.0	16.0	36.0	81.0	196.0	1	1000

## resid[6,13]	77.8	55.5	9.0	36.0	64.0	100.0	225.0	1	1000
## resid[6,14]	72.2	50.8	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[6,15]	1473.5	310.9	900.0	1225.0	1521.0	1681.0	2116.0	1	360
## resid[6,16]	13.7	16.1	0.0	1.0	9.0	25.0	64.0	1	1000
## resid[6,17]	77.0	55.1	9.0	36.0	64.0	100.0	225.0	1	1000
## resid[6,18]	463.0	136.1	196.0	361.0	484.0	576.0	729.0	1	1000
## resid[6,19]	67.9	48.8	9.0	36.0	64.0	81.0	196.0	1	1000
## resid[6,20]	256.3	194.9	4.0	100.0	225.0	361.0	729.0	1	1000
## resid[6,21]	3396.2	977.1	1602.0	2704.0	3364.0	4096.0	5476.0	1	1000
## resid[6,22]	14.6	19.0	0.0	1.0	9.0	16.0	64.0	1	440
## resid[6,23]	103.8	65.3	16.0	49.0	100.0	144.0	256.0	1	1000
## resid[6,24]	62.8	46.2	4.0	25.0	49.0	81.0	169.0	1	510
## resid[6,25]	876.3	444.8	196.0	529.0	784.0	1156.0	1936.0	1	1000
## resid[6,26]	102.6	137.0	0.0	9.0	49.0	138.2	484.0	1	1000
## resid[6,27]	154.9	131.9	1.0	64.0	121.0	225.0	484.0	1	1000
## resid[6,28]	94.0	59.5	16.0	49.0	81.0	121.0	225.0	1	370
## resid[6,29]	61.3	45.5	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[6,30]	13.0	19.3	0.0	1.0	4.0	16.0	64.0	1	540
## resid[6,31]	32.8	35.1	0.0	9.0	25.0	49.0	121.0	1	1000
## resid[6,32]	80.2	53.4	9.0	39.2	64.0	100.0	225.0	1	880
## resid[6,33]	103.3	67.1	16.0	64.0	81.0	144.0	256.0	1	1000
## resid[6,34]	66.7	45.4	9.0	36.0	49.0	94.9	169.0	1	1000
## resid[6,35]	86.1	126.1	0.0	9.0	36.0	121.0	441.0	1	1000
## resid[7,1]	170.6	99.5	36.0	100.0	144.0	225.0	400.0	1	500
## resid[7,2]	190.5	107.8	36.0	121.0	169.0	256.0	441.0	1	1000
## resid[7,3]	149.5	87.5	25.0	81.0	144.0	196.0	361.0	1	500
## resid[7,4]	92.9	58.4	16.0	49.0	81.0	121.0	225.0	1	1000
## resid[7,5]	175.8	99.7	36.0	100.0	169.0	225.0	400.0	1	1000
## resid[7,6]	106.1	69.0	16.0	64.0	100.0	144.0	256.0	1	780
## resid[7,7]	182.6	105.2	36.0	100.0	169.0	225.0	441.0	1	1000
## resid[7,8]	116.4	72.8	16.0	64.0	100.0	144.0	289.0	1	330
## resid[7,9]	16.7	22.2	0.0	1.0	9.0	25.0	81.0	1	340
## resid[7,10]	6452.8	553.7	5329.0	6084.0	6400.0	6889.0	7569.0	1	1000
## resid[7,11]	137.9	87.0	16.0	81.0	121.0	196.0	361.0	1	1000
## resid[7,12]	216.8	113.4	49.0	121.0	196.0	289.0	484.0	1	410
## resid[7,13]	187.7	100.0	49.0	121.0	169.0	256.0	400.0	1	1000
## resid[7,14]	145.7	86.2	25.0	81.0	121.0	196.0	361.0	1	310
## resid[7,15]	1557.7	318.3	961.0	1369.0	1521.0	1764.0	2209.0	1	1000
## resid[7,16]	69.5	65.0	1.0	25.0	49.0	100.0	225.0	1	1000
## resid[7,17]	224.7	123.5	49.0	144.0	196.0	289.0	529.0	1	860
## resid[7,18]	170.2	92.1	36.0	100.0	144.0	225.0	400.0	1	430
## resid[7,19]	77.5	57.7	9.0	36.0	64.0	100.0	225.0	1	1000
## resid[7,20]	10.7	13.8	0.0	1.0	4.0	16.0	49.0	1	1000
## resid[7,21]	13063.3	1464.6	10200.0	12100.0	13114.5	14099.6	16123.7	1	1000
## resid[7,22]	657.7	258.9	256.0	484.0	625.0	784.0	1225.0	1	1000
## resid[7,23]	602.5	268.5	225.0	400.0	576.0	729.0	1225.0	1	1000
## resid[7,24]	122.0	73.1	16.2	64.0	100.0	169.0	289.0	1	1000
## resid[7,25]	16.3	23.1	0.0	1.0	9.0	25.0	81.0	1	1000
## resid[7,26]	2528.4	883.5	1024.0	1849.0	2500.0	3136.0	4356.0	1	1000
## resid[7,27]	1880.3	613.0	841.0	1444.0	1849.0	2209.0	3249.0	1	330
## resid[7,28]	440.8	198.6	144.0	289.0	400.0	576.0	959.4	1	1000
## resid[7,29]	85.3	58.0	9.0	49.0	81.0	121.0	225.0	1	1000
## resid[7,30]	1628.2	295.1	1089.0	1444.0	1600.0	1849.0	2209.0	1	1000
## resid[7,31]	170.9	91.6	36.3	100.0	144.0	225.0	400.0	1	1000

## resid[7,32]	91.1	71.0	4.0	36.0	81.0	121.0	256.0	1	700
## resid[7,33]	235.7	136.2	49.0	144.0	196.0	324.0	529.0	1	610
## resid[7,34]	123.1	76.1	25.0	64.0	100.0	169.0	324.0	1	1000
## resid[7,35]	1908.4	266.0	1369.0	1764.0	1936.0	2116.0	2401.0	1	1000
## resid[8,1]	174.2	100.0	36.0	100.0	169.0	225.0	400.0	1	170
## resid[8,2]	75.3	55.3	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[8,3]	65.5	46.3	9.0	36.0	49.0	81.0	169.0	1	1000
## resid[8,4]	60.0	43.9	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[8,5]	101.0	63.8	16.0	49.0	100.0	137.9	256.0	1	1000
## resid[8,6]	163.2	92.4	36.0	100.0	144.0	196.0	361.0	1	1000
## resid[8,7]	10.8	16.2	0.0	1.0	4.0	16.0	63.6	1	940
## resid[8,8]	31.0	25.8	1.0	9.0	25.0	49.0	100.0	1	1000
## resid[8,9]	82.3	51.6	4.0	49.0	81.0	121.0	196.0	1	1000
## resid[8,10]	866.4	182.7	485.1	729.0	841.0	961.0	1225.0	1	1000
## resid[8,11]	44.9	55.2	0.0	4.0	25.0	64.0	196.0	1	250
## resid[8,12]	17.5	25.4	0.0	1.0	9.0	25.0	100.0	1	1000
## resid[8,13]	70.8	50.2	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[8,14]	73.6	48.2	9.0	36.0	64.0	100.0	196.0	1	780
## resid[8,15]	185.5	91.3	36.0	121.0	169.0	256.0	361.0	1	1000
## resid[8,16]	271.1	135.0	81.0	169.0	256.0	361.0	576.0	1	1000
## resid[8,17]	652.4	165.0	324.0	529.0	676.0	784.0	961.0	1	600
## resid[8,18]	57.6	43.5	4.1	25.0	49.0	81.0	169.0	1	200
## resid[8,19]	16.3	21.8	0.0	1.0	9.0	25.0	81.0	1	1000
## resid[8,20]	44.8	39.4	1.0	16.0	36.0	64.0	144.0	1	700
## resid[8,21]	356.1	210.7	49.0	196.0	324.0	484.0	841.0	1	1000
## resid[8,22]	24.4	27.4	0.0	4.0	16.0	36.0	100.0	1	1000
## resid[8,23]	73.0	47.9	9.0	36.0	64.0	100.0	195.3	1	610
## resid[8,24]	9.2	14.2	0.0	1.0	4.0	9.0	49.0	1	670
## resid[8,25]	1200.1	225.4	784.0	1024.0	1225.0	1369.0	1600.0	1	1000
## resid[8,26]	1803.2	698.9	677.3	1296.0	1681.0	2209.0	3478.0	1	1000
## resid[8,27]	11.7	16.1	0.0	1.0	4.0	16.0	49.0	1	1000
## resid[8,28]	64.0	44.1	9.0	36.0	49.0	81.0	169.0	1	710
## resid[8,29]	36.7	35.9	1.0	9.0	25.0	49.0	121.0	1	1000
## resid[8,30]	140.7	200.2	0.0	16.0	64.0	196.0	676.0	1	710
## resid[8,31]	15017.2	3226.4	9413.8	12770.0	14880.0	16900.0	22200.0	1	1000
## resid[8,32]	20723.7	890.1	18770.0	20160.0	20740.0	21320.0	22492.5	1	1000
## resid[8,33]	70.6	52.6	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[8,34]	64.4	42.1	4.0	36.0	64.0	81.0	144.0	1	1000
## resid[8,35]	3612.8	374.9	2916.0	3364.0	3600.0	3844.0	4356.0	1	1000
## resid[9,1]	36.9	29.9	1.0	16.0	25.0	49.0	100.0	1	600
## resid[9,2]	43.4	36.5	1.0	16.0	36.0	64.0	144.0	1	1000
## resid[9,3]	37.3	31.5	4.0	16.0	25.0	49.0	121.0	1	1000
## resid[9,4]	19.2	21.8	0.0	4.0	9.0	25.0	81.0	1	1000
## resid[9,5]	1669.1	521.2	784.0	1296.0	1600.0	2025.0	2809.0	1	1000
## resid[9,6]	47.4	36.1	4.0	25.0	36.0	64.0	144.0	1	1000
## resid[9,7]	85.5	57.9	9.0	49.0	81.0	121.0	225.0	1	1000
## resid[9,8]	96.8	50.0	16.0	64.0	100.0	121.0	196.0	1	1000
## resid[9,9]	191.0	100.2	25.0	121.0	169.0	256.0	400.0	1	1000
## resid[9,10]	1330.5	199.5	900.0	1172.9	1369.0	1444.0	1681.0	1	310
## resid[9,11]	47.2	37.5	4.0	16.0	36.0	64.0	144.0	1	220
## resid[9,12]	216.3	113.6	49.0	144.0	196.0	289.0	484.0	1	430
## resid[9,13]	21.4	24.4	0.0	4.0	16.0	25.0	99.5	1	1000
## resid[9,14]	834.8	163.4	484.0	729.0	841.0	961.0	1156.0	1	1000
## resid[9,15]	498.9	148.5	225.0	400.0	484.0	576.0	784.0	1	550



## resid[9,16]	53.7	40.3	4.0	25.0	49.0	81.0	144.0	1	220
## resid[9,17]	81.4	60.8	4.0	36.0	64.0	100.0	225.0	1	430
## resid[9,18]	51.3	45.5	1.0	16.0	36.0	64.0	169.0	1	540
## resid[9,19]	85.3	41.8	16.0	49.0	81.0	121.0	169.0	1	1000
## resid[9,20]	19.3	19.9	0.0	4.0	16.0	25.0	80.6	1	1000
## resid[9,21]	740.4	289.1	289.0	529.0	729.0	900.0	1369.0	1	630
## resid[9,22]	327.2	164.2	100.0	225.0	289.0	400.0	676.0	1	1000
## resid[9,23]	11.1	18.7	0.0	1.0	4.0	16.0	64.0	1	630
## resid[9,24]	341.2	115.2	144.0	256.0	324.0	441.0	576.0	1	560
## resid[9,25]	2583.2	338.0	1936.0	2401.0	2601.0	2809.0	3249.0	1	1000
## resid[9,26]	14639.3	3100.0	9029.7	12374.6	14400.0	16640.0	21602.7	1	1000
## resid[9,27]	302.2	159.2	64.0	196.0	289.0	400.0	676.0	1	280
## resid[9,28]	70.0	47.9	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[9,29]	2859.8	341.5	2209.0	2601.0	2916.0	3136.0	3481.0	1	610
## resid[9,30]	902.1	157.7	576.0	784.0	900.0	1024.0	1156.0	1	880
## resid[9,31]	104.6	68.7	16.0	49.0	81.0	144.0	289.0	1	1000
## resid[9,32]	40.9	45.6	0.0	9.0	25.0	64.0	169.0	1	310
## resid[9,33]	73.2	51.8	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[9,34]	453.3	110.5	225.0	361.0	441.0	529.0	676.0	1	1000
## resid[9,35]	1449.9	224.1	1024.0	1296.0	1444.0	1600.0	1849.0	1	370
## resid[10,1]	44.4	37.6	1.0	16.0	36.0	64.0	144.0	1	1000
## resid[10,2]	35.8	32.9	1.0	16.0	25.0	49.0	121.0	1	740
## resid[10,3]	47.4	37.3	4.0	25.0	36.0	64.0	144.0	1	1000
## resid[10,4]	41.1	33.0	1.1	16.0	36.0	49.0	121.0	1	1000
## resid[10,5]	57.5	44.0	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[10,6]	25.7	24.9	1.0	9.0	16.0	36.0	81.0	1	580
## resid[10,7]	40.5	32.5	4.0	16.0	36.0	49.0	121.0	1	730
## resid[10,8]	488.6	115.5	289.0	400.0	484.0	576.0	729.0	1	1000
## resid[10,9]	8.3	10.8	0.0	1.0	4.0	9.0	36.0	1	1000
## resid[10,10]	55.2	42.3	4.0	25.0	49.0	81.0	169.0	1	1000
## resid[10,11]	26.5	24.8	1.0	9.0	25.0	36.0	100.0	1	760
## resid[10,12]	35.1	29.5	1.0	16.0	25.0	49.0	120.5	1	1000
## resid[10,13]	72.5	52.4	4.1	36.0	64.0	100.0	196.0	1	1000
## resid[10,14]	564.8	142.2	289.0	484.0	576.0	676.0	841.0	1	1000
## resid[10,15]	35.0	41.8	0.0	4.0	25.0	49.0	144.0	1	400
## resid[10,16]	51.2	42.8	1.1	25.0	36.0	64.0	144.0	1	480
## resid[10,17]	12.4	11.9	0.0	4.0	9.0	16.0	36.0	1	490
## resid[10,18]	103.8	58.3	16.0	64.0	100.0	144.0	225.0	1	1000
## resid[10,19]	30.0	26.3	1.0	9.0	25.0	36.0	100.0	1	1000
## resid[10,20]	85.3	63.4	9.0	36.0	64.0	121.0	256.0	1	1000
## resid[10,21]	75.9	54.3	9.0	36.0	64.0	100.0	225.0	1	910
## resid[10,22]	39.4	31.9	4.0	16.0	36.0	49.0	121.0	1	1000
## resid[10,23]	39.0	33.3	1.0	16.0	36.0	49.0	121.0	1	860
## resid[10,24]	13.3	21.7	0.0	1.0	4.0	16.0	81.0	1	1000
## resid[10,25]	1639.1	310.2	1025.6	1444.0	1681.0	1849.0	2209.0	1	1000
## resid[10,26]	149.1	88.1	25.0	81.0	144.0	196.0	361.0	1	1000
## resid[10,27]	42.6	33.5	4.0	16.0	36.0	64.0	144.0	1	1000
## resid[10,28]	47.9	44.0	1.0	16.0	36.0	64.0	169.0	1	1000
## resid[10,29]	5539.6	373.3	4761.0	5329.0	5625.0	5776.0	6241.0	1	760
## resid[10,30]	31.4	29.9	1.0	9.0	25.0	49.0	121.0	1	500
## resid[10,31]	857.0	354.1	289.0	625.0	784.0	1024.0	1764.0	1	900
## resid[10,32]	31.6	29.9	1.0	9.0	25.0	49.0	100.0	1	1000
## resid[10,33]	16.1	22.2	0.0	1.0	9.0	25.0	81.0	1	1000
## resid[10,34]	69.1	51.9	4.1	36.0	64.0	100.0	196.0	1	1000

## resid[10,35]	22.1	25.1	0.0	4.0	16.0	36.0	81.0	1	1000
## resid[11,1]	193.2	114.5	36.0	100.0	169.0	256.0	441.0	1	1000
## resid[11,2]	210.5	113.3	64.0	121.0	196.0	256.0	484.0	1	1000
## resid[11,3]	602.4	242.2	225.0	441.0	576.0	729.0	1156.0	1	1000
## resid[11,4]	397.1	179.4	121.0	256.0	361.0	484.0	841.0	1	1000
## resid[11,5]	1397.6	463.2	625.0	1089.0	1369.0	1681.0	2500.0	1	1000
## resid[11,6]	300.3	140.2	100.0	196.0	289.0	361.0	625.0	1	520
## resid[11,7]	242.4	126.9	49.3	144.0	225.0	324.0	529.0	1	1000
## resid[11,8]	9730.3	1476.7	6889.0	8836.0	9801.0	10820.0	12540.0	1	1000
## resid[11,9]	4347.6	571.9	3138.8	3969.0	4356.0	4761.0	5476.0	1	1000
## resid[11,10]	421.1	181.8	144.0	289.0	400.0	529.0	841.0	1	1000
## resid[11,11]	250.0	128.2	64.0	169.0	225.0	324.0	529.0	1	1000
## resid[11,12]	255.9	125.0	64.4	169.0	225.0	324.0	576.0	1	1000
## resid[11,13]	170.5	131.9	4.0	64.0	144.0	256.0	484.0	1	430
## resid[11,14]	304.6	165.1	36.0	169.0	289.0	400.0	676.0	1	440
## resid[11,15]	657.2	255.8	256.0	484.0	625.0	784.0	1225.0	1	780
## resid[11,16]	221.5	112.6	64.0	144.0	196.0	289.0	484.0	1	540
## resid[11,17]	251.5	126.8	64.0	149.9	225.0	324.0	529.0	1	910
## resid[11,18]	180.6	162.0	4.0	64.0	144.0	256.0	576.0	1	1000
## resid[11,19]	261.5	135.3	64.0	169.0	225.0	324.0	576.0	1	1000
## resid[11,20]	396.8	189.9	100.5	256.0	361.0	529.0	841.0	1	1000
## resid[11,21]	246.8	121.5	64.0	169.0	225.0	324.0	529.0	1	1000
## resid[11,22]	262.2	133.5	64.0	169.0	225.0	351.4	576.0	1	1000
## resid[11,23]	296.0	234.9	9.0	121.0	256.0	400.0	900.0	1	710
## resid[11,24]	1656.4	479.5	784.0	1296.0	1681.0	2025.0	2601.0	1	1000
## resid[11,25]	655.6	268.3	225.7	484.0	625.0	784.0	1225.0	1	1000
## resid[11,26]	257.1	130.1	81.0	169.0	225.0	324.0	576.0	1	1000
## resid[11,27]	278.8	135.7	81.0	196.0	256.0	361.0	576.0	1	710
## resid[11,28]	15058.1	2077.5	11450.0	13690.0	15130.0	16380.0	19040.0	1	1000
## resid[11,29]	6764.3	1005.3	4900.0	6084.0	6724.0	7396.0	8649.0	1	1000
## resid[11,30]	376.8	167.7	121.0	256.0	361.0	484.0	784.0	1	1000
## resid[11,31]	612.6	245.3	225.0	441.0	576.0	729.0	1156.0	1	1000
## resid[11,32]	210.6	103.6	49.0	144.0	196.0	256.0	441.0	1	1000
## resid[11,33]	78.0	93.4	0.0	16.0	49.0	100.0	361.0	1	1000
## resid[11,34]	2139.0	411.7	1369.0	1849.0	2116.0	2401.0	2916.0	1	450
## resid[11,35]	337.9	158.5	100.0	225.0	324.0	441.0	727.6	1	1000
## resid[12,1]	46.8	42.1	1.0	16.0	36.0	64.0	169.0	1	1000
## resid[12,2]	70.8	49.8	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[12,3]	89.7	63.2	9.0	49.0	81.0	121.0	256.0	1	1000
## resid[12,4]	80.0	54.1	9.0	36.0	64.0	100.0	225.0	1	1000
## resid[12,5]	180.4	104.5	36.0	100.0	169.0	225.0	441.0	1	1000
## resid[12,6]	83.1	54.7	9.1	49.0	64.0	121.0	196.0	1	1000
## resid[12,7]	19.4	26.8	0.0	1.8	9.0	25.0	100.0	1	990
## resid[12,8]	2934.7	336.1	2304.0	2704.0	2916.0	3136.0	3481.0	1	1000
## resid[12,9]	76.6	54.3	9.0	36.0	64.0	100.0	196.0	1	1000
## resid[12,10]	208.0	118.1	49.0	121.0	196.0	256.0	527.8	1	1000
## resid[12,11]	75.6	52.8	9.0	36.0	64.0	100.0	196.0	1	900
## resid[12,12]	63.7	46.1	9.0	36.0	49.0	81.0	196.0	1	420
## resid[12,13]	25.0	33.1	0.0	4.0	9.0	36.0	121.0	1	1000
## resid[12,14]	91.8	60.4	9.1	49.0	81.0	121.0	225.0	1	1000
## resid[12,15]	153.6	90.0	36.0	81.0	144.0	196.0	361.0	1	400
## resid[12,16]	73.0	48.1	9.0	36.0	64.0	100.0	196.0	1	840
## resid[12,17]	1407.9	209.8	1024.0	1296.0	1444.0	1521.0	1764.0	1	1000
## resid[12,18]	12.4	15.3	0.0	1.0	9.0	16.0	49.0	1	1000

## resid[12,19]	95.1	62.6	16.0	49.0	81.0	121.0	256.0	1	1000
## resid[12,20]	159.6	93.2	36.0	100.0	144.0	196.0	361.0	1	1000
## resid[12,21]	64.5	42.9	9.0	36.0	49.0	81.0	169.0	1	1000
## resid[12,22]	68.5	46.8	9.0	36.0	64.0	100.0	169.0	1	620
## resid[12,23]	24.5	29.2	0.0	4.0	16.0	36.0	100.0	1	1000
## resid[12,24]	708.4	181.2	361.0	576.0	729.0	841.0	1089.0	1	910
## resid[12,25]	167.7	91.1	36.0	100.0	144.0	225.0	400.0	1	1000
## resid[12,26]	63.6	44.6	9.0	36.0	49.0	81.0	169.0	1	1000
## resid[12,27]	72.7	49.9	9.0	36.0	64.0	100.0	196.0	1	430
## resid[12,28]	1157.2	229.4	729.0	1024.0	1156.0	1296.0	1600.0	1	1000
## resid[12,29]	31.5	27.3	0.0	9.0	25.0	49.0	100.0	1	1000
## resid[12,30]	292.3	146.6	81.0	196.0	256.0	361.0	625.0	1	1000
## resid[12,31]	64.4	45.9	9.0	36.0	49.0	81.0	195.3	1	570
## resid[12,32]	3999.9	351.0	3249.0	3844.0	3969.0	4225.0	4624.0	1	1000
## resid[12,33]	12.5	16.5	0.0	1.0	4.0	16.0	64.0	1	850
## resid[12,34]	10.6	14.3	0.0	1.0	4.0	16.0	49.0	1	1000
## resid[12,35]	136.2	79.6	25.0	81.0	121.0	169.0	324.0	1	1000
## resid[13,1]	141.6	93.3	25.0	81.0	121.0	196.0	400.0	1	1000
## resid[13,2]	254.4	129.1	64.0	169.0	225.0	324.0	574.8	1	360
## resid[13,3]	223.2	115.0	64.0	144.0	210.0	289.0	484.0	1	1000
## resid[13,4]	237.7	122.2	64.0	144.0	225.0	289.0	529.0	1	290
## resid[13,5]	401.9	180.3	121.5	256.0	361.0	484.0	841.0	1	500
## resid[13,6]	181.9	98.6	49.0	121.0	169.0	225.0	441.0	1	1000
## resid[13,7]	306.8	154.7	81.0	196.0	289.0	400.0	676.0	1	1000
## resid[13,8]	1707.3	342.0	1025.6	1444.0	1681.0	1936.0	2401.0	1	1000
## resid[13,9]	224.7	123.3	49.0	144.0	196.0	289.0	529.0	1	1000
## resid[13,10]	92.8	78.5	1.0	36.0	81.0	138.2	289.0	1	350
## resid[13,11]	231.8	130.2	49.0	144.0	196.0	289.0	529.0	1	820
## resid[13,12]	103.3	109.4	1.0	25.0	81.0	144.0	400.0	1	440
## resid[13,13]	28.9	38.5	0.0	4.0	16.0	36.0	144.0	1	1000
## resid[13,14]	262.1	122.9	81.0	169.0	256.0	324.0	529.0	1	1000
## resid[13,15]	232.6	115.2	64.0	144.0	225.0	289.0	529.0	1	320
## resid[13,16]	204.8	109.0	49.0	121.0	196.0	256.0	484.0	1	1000
## resid[13,17]	17578.2	1411.5	14646.0	16640.0	17690.0	18500.0	20160.0	1	1000
## resid[13,18]	77.0	79.4	1.0	25.0	49.0	100.0	289.0	1	1000
## resid[13,19]	260.3	134.9	64.0	169.0	225.0	324.0	576.0	1	1000
## resid[13,20]	248.2	127.5	64.0	144.0	225.0	324.0	529.0	1	750
## resid[13,21]	175.9	98.4	36.0	100.0	169.0	225.0	400.0	1	610
## resid[13,22]	462.0	407.4	4.0	144.0	361.0	676.0	1519.1	1	1000
## resid[13,23]	1015.2	457.0	225.0	676.0	961.0	1296.0	2025.0	1	370
## resid[13,24]	185.4	145.0	4.0	81.0	144.0	256.0	574.8	1	620
## resid[13,25]	321.6	156.6	100.0	225.0	289.0	400.0	676.0	1	1000
## resid[13,26]	188.5	101.8	49.0	121.0	169.0	256.0	441.0	1	1000
## resid[13,27]	2875.0	1020.1	1225.0	2116.0	2809.0	3481.0	5041.0	1	890
## resid[13,28]	27012.3	1786.8	23410.0	25920.0	27230.0	28220.0	30280.0	1	1000
## resid[13,29]	208.6	114.8	49.0	121.0	196.0	256.0	484.0	1	650
## resid[13,30]	428.8	199.4	144.0	289.0	400.0	529.0	900.0	1	1000
## resid[13,31]	214.8	120.2	49.0	121.0	196.0	289.0	527.8	1	1000
## resid[13,32]	837.9	428.4	144.0	529.0	784.0	1089.0	1764.0	1	1000
## resid[13,33]	1289.3	400.0	576.0	1024.0	1296.0	1521.0	2116.0	1	1000
## resid[13,34]	558.8	240.7	196.0	361.0	529.0	676.0	1089.0	1	1000
## resid[13,35]	242.1	126.9	64.0	144.0	225.0	324.0	529.0	1	1000
## resid[14,1]	22.0	26.2	0.0	4.0	16.0	25.0	100.0	1	1000
## resid[14,2]	91.4	58.5	16.0	49.0	81.0	121.0	225.0	1	450

```
## resid[14,3]      64.0   45.9    4.1   25.0   49.0   81.0   195.3    1  1000
## resid[14,4]     127.6   78.3   25.0   81.0  121.0  169.0   324.0    1   340
## resid[14,5]     115.9   68.9   16.0   64.0  100.0  144.0   289.0    1  1000
## resid[14,6]      49.4   36.2    4.0   25.0   49.0   64.0   144.0    1  1000
## resid[14,7]      11.0   12.6    0.0    1.0    9.0   16.0   49.0    1  1000
## resid[14,8]      68.8   48.1    9.0   36.0   64.0  100.0   196.0    1  1000
## resid[14,9]      80.4   57.9    9.0   36.0   64.0  100.0   225.0    1  1000
## resid[14,10]     10.9   17.0    0.0    1.0    4.0   16.0   64.0    1  1000
## resid[14,11]     78.2   51.2    9.0   36.0   64.0  100.0   196.0    1  1000
## resid[14,12]    112.5   71.1   16.0   64.0  100.0  144.0   289.0    1  1000
## resid[14,13]    655.6  188.2  289.0  529.0  676.0  784.0  1024.0    1  1000
## resid[14,14]    128.1   82.2   16.2   64.0  121.0  169.0   324.0    1   370
## resid[14,15]    105.3   69.3   16.0   64.0  100.0  144.0   256.0    1   670
## resid[14,16]     82.4   55.6    9.0   49.0   64.0  100.0   225.0    1  1000
## resid[14,17]   1920.6  247.3 1369.0 1764.0 1936.0 2116.0 2401.0    1   790
## resid[14,18]     40.2   34.9    1.0   16.0   36.0   49.0   121.0    1  1000
## resid[14,19]    100.9   74.0    9.0   49.0   81.0  144.0   288.2    1  1000
## resid[14,20]    133.2   82.4   25.0   81.0  121.0  169.0   361.0    1  1000
## resid[14,21]     68.3   46.3    9.0   36.0   64.0  100.0   196.0    1  1000
## resid[14,22]    683.1  134.7  441.0  576.0  676.0  784.0   900.0    1  1000
## resid[14,23]   5657.0  591.6 4489.0 5184.0 5625.0 6084.0 6724.0    1   930
## resid[14,24]    359.6  169.2  121.0  225.0  324.0  441.0   784.0    1  1000
## resid[14,25]    301.2  151.1   81.0  196.0  289.0  400.0   676.0    1   550
## resid[14,26]     54.4   43.7    4.0   25.0   36.0   81.0   169.0    1  1000
## resid[14,27]     71.2   40.2    4.1   49.0   64.0  100.0   169.0    1  1000
## resid[14,28]   2276.7  251.7 1764.0 2116.0 2304.0 2500.0 2704.0    1   990
## resid[14,29]    162.7   91.8   36.0  100.0  144.0  196.0   400.0    1   490
## resid[14,30]    193.1  104.5   49.0  121.0  169.0  256.0   441.0    1   800
## resid[14,31]     38.6   48.3    0.0    4.0   25.0   49.0   169.0    1   610
## resid[14,32]    156.1   69.3   36.0  100.0  144.0  196.0   289.0    1  1000
## resid[14,33]    152.9   71.7   36.0  100.0  144.0  196.0   289.0    1  1000
## resid[14,34]    432.3  200.5  121.0  289.0  400.0  529.0   900.0    1   620
## resid[14,35]    210.6  110.7   49.0  121.0  196.0  256.0   484.0    1  1000
## deviance      15932.9    6.3 15920.0 15930.0 15930.0 15940.0 15950.0    1  1000
##
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 17.1 and DIC = 15950.0
## DIC is an estimate of expected predictive error (lower deviance is better).
```

```
#Rhat
```

```
summary(Rhat <- (Resul4$summary[, "Rhat"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.9995  1.0005  1.0018  1.0032  1.0042  1.0375
```

```
summary(Resul4$summary[, "n.eff"])
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 170.0   760.0  1000.0   858.4  1000.0  1000.0
```

```
Resul4$DIC
```

```
## [1] 15950
```

```
Resul4$pD
```

```
## [1] 17.136
```

```
#MSE
```

```
mean(Resul4$mean$resid)
```

```
## [1] 799.2438
```

```
set.seed(22)
```

```
Modelo <- function(){  
  for (i in 1:14){  
    for (j in 1:35){  
      Captures[i,j] ~ dpois(lambda[i,j])  
      log(lambda[i,j]) <- beta0 + beta1*TMin[i,j] + beta2*Precipita[i,j] + beta3*AltMed[i,j] + beta4*  
    }  
    v[i] ~ dnorm(0,tauv)  
  }  
  tauv <- 1/pow(sdv,2)  
  sdv ~ dunif(0,10)  
  
  #distribuciones iniciales  
  beta0 ~ dflat()  
  beta1 ~ dflat()  
  beta2 ~ dflat()  
  beta3 ~ dflat()  
  beta4 ~ dflat()  
  
  for (i in 1:14) {  
    for (j in 1:35) {  
      Captures.pred[i,j] ~ dpois(lambda[i,j])  
      resid[i,j] <- pow(Captures[i,j] - Captures.pred[i,j],2)  
    }  
  }  
}  
  
#Datos  
Datos <- list(Captures = matrix(DatosModeloSIN5$Captures, nrow = 14),  
              TMin = matrix(DatosModeloSIN5$TMinMed, nrow = 14),  
              Precipita = matrix(DatosModeloSIN5$Precipita, nrow = 14),  
              AltMed = matrix(scale(DatosModeloSIN5$AltMed), nrow = 14),  
              Especies = matrix(scale(DatosModeloSIN5$SumaEspecies), nrow = 14),  
              Superficie = matrix((DatosModeloSIN5$Superf), nrow = 14))  
  
#Iniciales  
Iniciales <- function(){
```

```
list(beta0 = rnorm(1), v = rnorm(14), sdv = runif(1,0,3),
      beta1 = rnorm(1), beta2 = rnorm(1), beta3 = rnorm(1), beta4 = rnorm(1))
}
```

*#Parámetros*

```
Param <- c("beta0","beta1","beta2","beta3","beta4", "v", "sdv", "resid")
```

Resul5

```
## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/RtmpAXk7nM/model15d826215c6e.txt", fi
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
```

```
## n.sims = 1002 iterations saved
```

##	mean	sd	2.5%	25%	50%	75%	97.5%	Rhat	n.eff
## beta0	-4.1	0.1	-4.4	-4.2	-4.1	-4.1	-3.9	1.0	350
## beta1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	400
## beta2	0.2	0.0	0.2	0.2	0.2	0.2	0.2	1.0	1000
## beta3	0.5	0.0	0.5	0.5	0.5	0.5	0.6	1.0	950
## beta4	0.5	0.0	0.5	0.5	0.5	0.5	0.5	1.0	1000
## v[1]	0.3	0.1	0.1	0.3	0.3	0.4	0.6	1.0	980
## v[2]	-0.3	0.1	-0.5	-0.4	-0.3	-0.2	-0.1	1.0	300
## v[3]	-0.3	0.1	-0.5	-0.4	-0.3	-0.2	0.0	1.0	790
## v[4]	-0.4	0.1	-0.6	-0.4	-0.4	-0.3	-0.1	1.0	560
## v[5]	0.0	0.1	-0.3	-0.1	0.0	0.0	0.2	1.0	320
## v[6]	0.0	0.1	-0.3	-0.1	0.0	0.0	0.2	1.0	380
## v[7]	0.2	0.1	0.0	0.1	0.2	0.3	0.4	1.0	560
## v[8]	0.0	0.1	-0.2	-0.1	0.0	0.0	0.2	1.0	590
## v[9]	-0.4	0.1	-0.7	-0.5	-0.4	-0.4	-0.2	1.0	390
## v[10]	-0.4	0.1	-0.6	-0.5	-0.4	-0.3	-0.1	1.0	880
## v[11]	0.7	0.1	0.5	0.7	0.7	0.8	1.0	1.0	410
## v[12]	0.1	0.1	-0.1	0.0	0.1	0.2	0.4	1.0	900
## v[13]	0.6	0.1	0.4	0.5	0.6	0.7	0.8	1.0	810
## v[14]	-0.1	0.1	-0.3	-0.2	-0.1	0.0	0.1	1.0	1000
## sdv	0.4	0.1	0.3	0.3	0.4	0.4	0.6	1.0	1000
## resid[1,1]	1.7	3.1	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[1,2]	1.9	3.4	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[1,3]	4.5	6.2	0.0	1.0	4.0	4.0	25.0	1.0	1000
## resid[1,4]	305.1	146.8	81.0	196.0	289.0	400.0	625.0	1.0	1000
## resid[1,5]	177.4	97.9	49.0	100.0	169.0	225.0	400.0	1.0	1000
## resid[1,6]	141.1	80.0	25.0	81.0	121.0	169.0	324.0	1.0	330
## resid[1,7]	77.4	68.4	1.0	25.0	64.0	100.0	256.0	1.0	1000
## resid[1,8]	262.6	131.5	64.0	169.0	256.0	324.0	576.0	1.0	1000
## resid[1,9]	962.9	365.8	361.0	729.0	900.0	1156.0	1764.0	1.0	1000
## resid[1,10]	1049.4	238.3	625.0	900.0	1024.0	1225.0	1521.0	1.0	1000
## resid[1,11]	189.1	104.5	49.0	121.0	169.0	256.0	441.0	1.0	1000
## resid[1,12]	116.4	91.2	9.0	49.0	100.0	169.0	360.0	1.0	460
## resid[1,13]	434.3	231.8	100.0	256.0	400.0	576.0	1024.0	1.0	1000
## resid[1,14]	15536.9	3043.8	10400.0	13460.0	15130.0	17420.0	22200.0	1.0	250
## resid[1,15]	924.9	401.8	289.8	625.0	870.0	1156.0	1849.0	1.0	320
## resid[1,16]	208.7	112.0	49.0	121.0	196.0	256.0	484.0	1.0	1000
## resid[1,17]	62.0	52.2	1.0	25.0	49.0	100.0	196.0	1.0	1000
## resid[1,18]	66.3	74.9	0.0	9.0	36.0	100.0	256.0	1.0	1000
## resid[1,19]	3.3	4.6	0.0	0.0	1.0	4.0	16.0	1.0	1000
## resid[1,20]	1228.9	66.7	1089.0	1225.0	1225.0	1296.0	1296.0	1.0	500

## resid[1,21]	168.4	94.1	36.0	100.0	144.0	225.0	400.0	1.0	550
## resid[1,22]	109.2	72.1	4.0	49.0	100.0	162.8	289.0	1.0	1000
## resid[1,23]	5715.8	671.6	4356.0	5184.0	5776.0	6241.0	7056.0	1.0	1000
## resid[1,24]	2300.4	534.5	1296.0	1936.0	2304.0	2704.0	3364.0	1.0	1000
## resid[1,25]	236.7	97.5	64.0	169.0	225.0	289.0	441.0	1.0	1000
## resid[1,26]	6.9	9.5	0.0	1.0	4.0	9.0	36.0	1.0	1000
## resid[1,27]	65.6	27.4	16.0	49.0	64.0	81.0	121.0	1.0	530
## resid[1,28]	10101.5	492.7	9029.7	9801.0	10200.0	10400.0	11030.0	1.0	700
## resid[1,29]	12.1	13.1	0.0	4.0	9.0	16.0	49.0	1.0	660
## resid[1,30]	1.9	3.0	0.0	0.0	1.0	4.0	9.0	1.0	410
## resid[1,31]	19.6	23.2	0.0	4.0	12.5	25.0	81.0	1.0	1000
## resid[1,32]	145.2	89.9	25.0	81.0	121.0	196.0	361.0	1.0	1000
## resid[1,33]	1347.5	236.8	900.0	1156.0	1369.0	1521.0	1849.0	1.0	340
## resid[1,34]	425.4	185.2	144.0	289.0	400.0	529.0	900.0	1.0	1000
## resid[1,35]	40.7	39.5	0.0	9.0	36.0	64.0	144.0	1.0	770
## resid[2,1]	0.8	1.7	0.0	0.0	0.0	1.0	4.0	1.0	500
## resid[2,2]	0.8	0.8	0.0	0.0	1.0	1.0	4.0	1.0	1000
## resid[2,3]	2.5	4.0	0.0	0.0	1.0	4.0	15.8	1.0	860
## resid[2,4]	48.0	36.5	4.0	25.0	36.0	64.0	144.0	1.0	320
## resid[2,5]	51.8	37.9	4.0	25.0	49.0	64.0	144.0	1.0	800
## resid[2,6]	39.3	36.9	1.0	16.0	25.0	49.0	144.0	1.0	1000
## resid[2,7]	8.4	11.6	0.0	1.0	4.0	9.0	36.0	1.0	490
## resid[2,8]	115.9	69.3	16.0	64.0	100.0	144.0	289.0	1.0	1000
## resid[2,9]	179.0	98.5	49.0	100.0	169.0	225.0	439.9	1.0	510
## resid[2,10]	31.4	36.9	0.0	4.0	16.0	49.0	121.0	1.0	1000
## resid[2,11]	86.5	57.5	9.1	49.0	81.0	121.0	225.0	1.0	1000
## resid[2,12]	20.1	28.8	0.0	1.0	9.0	25.0	100.0	1.0	1000
## resid[2,13]	88.6	126.2	0.0	9.0	36.0	121.0	441.0	1.0	1000
## resid[2,14]	367.4	170.5	121.0	256.0	324.0	441.0	784.0	1.0	1000
## resid[2,15]	1570.2	533.0	729.0	1156.0	1521.0	1849.0	2809.0	1.0	1000
## resid[2,16]	9.7	14.6	0.0	1.0	4.0	16.0	49.0	1.0	1000
## resid[2,17]	19.6	27.9	0.0	1.0	9.0	25.0	100.0	1.0	1000
## resid[2,18]	382.3	167.9	81.0	256.0	361.0	484.0	729.0	1.0	1000
## resid[2,19]	1.2	2.2	0.0	0.0	0.0	1.0	9.0	1.0	790
## resid[2,20]	2.0	1.8	0.0	1.0	1.0	4.0	4.0	1.0	1000
## resid[2,21]	40.4	41.1	0.0	9.0	25.0	49.0	144.0	1.0	1000
## resid[2,22]	103.2	67.8	4.0	49.0	90.5	144.0	256.0	1.0	1000
## resid[2,23]	883.1	628.7	36.3	400.0	784.0	1225.0	2398.6	1.0	1000
## resid[2,24]	199.9	109.1	49.0	121.0	196.0	256.0	441.0	1.0	1000
## resid[2,25]	174.2	106.8	36.0	100.0	144.0	225.0	441.0	1.0	1000
## resid[2,26]	26.7	13.3	1.1	16.0	25.0	36.0	49.0	1.0	1000
## resid[2,27]	302.3	64.9	169.0	256.0	324.0	361.0	400.0	1.0	740
## resid[2,28]	10741.0	497.9	9604.0	10400.0	10820.0	11030.0	11660.0	1.0	480
## resid[2,29]	1.8	2.9	0.0	0.0	1.0	4.0	9.0	1.0	1000
## resid[2,30]	5.2	6.8	0.0	1.0	4.0	9.0	25.0	1.0	1000
## resid[2,31]	17.8	19.3	0.0	4.0	9.0	25.0	64.0	1.0	890
## resid[2,32]	15.2	22.1	0.0	1.0	9.0	16.0	81.0	1.0	1000
## resid[2,33]	9.9	12.8	0.0	1.0	4.0	16.0	49.0	1.0	1000
## resid[2,34]	38.0	37.9	0.0	9.0	25.0	49.0	144.0	1.0	1000
## resid[2,35]	831.4	344.3	324.0	576.0	784.0	1024.0	1600.0	1.0	490
## resid[3,1]	1.2	2.1	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[3,2]	1.2	2.2	0.0	0.0	1.0	1.0	8.9	1.0	410
## resid[3,3]	1.5	3.0	0.0	0.0	1.0	1.0	9.0	1.0	790
## resid[3,4]	50.6	38.9	4.0	25.0	36.0	64.0	144.0	1.0	620

## resid[3,5]	35.8	30.6	1.0	16.0	25.0	49.0	121.0	1.0	1000
## resid[3,6]	64.5	52.8	4.0	25.0	49.0	81.0	196.0	1.0	1000
## resid[3,7]	7.7	9.8	0.0	1.0	4.0	9.0	36.0	1.0	610
## resid[3,8]	132.8	79.0	25.0	81.0	121.0	169.0	324.0	1.0	900
## resid[3,9]	166.0	96.5	36.0	100.0	144.0	225.0	361.0	1.0	1000
## resid[3,10]	16.1	19.8	0.0	1.0	9.0	25.0	64.0	1.0	680
## resid[3,11]	181.8	99.6	36.0	121.0	169.0	225.0	441.0	1.0	1000
## resid[3,12]	51.5	48.8	1.0	16.0	36.0	81.0	169.0	1.0	630
## resid[3,13]	41.0	56.9	0.0	4.0	16.0	49.0	196.0	1.0	1000
## resid[3,14]	331.4	159.0	100.0	225.0	289.0	400.0	729.0	1.0	1000
## resid[3,15]	157.2	87.9	36.0	100.0	144.0	196.0	361.0	1.0	1000
## resid[3,16]	76.7	66.3	1.0	25.0	64.0	100.0	225.0	1.0	1000
## resid[3,17]	453.8	141.3	196.0	361.0	441.0	529.0	729.0	1.0	1000
## resid[3,18]	240.3	131.2	49.0	144.0	225.0	324.0	529.0	1.0	1000
## resid[3,19]	2.6	1.6	0.0	1.0	4.0	4.0	4.0	1.0	1000
## resid[3,20]	0.6	1.2	0.0	0.0	0.0	1.0	4.0	1.0	1000
## resid[3,21]	164.4	93.2	36.0	100.0	144.0	225.0	400.0	1.0	1000
## resid[3,22]	817.3	218.1	400.0	676.0	841.0	961.0	1225.0	1.0	1000
## resid[3,23]	4438.2	1115.4	2401.0	3600.0	4356.0	5184.0	6724.0	1.0	1000
## resid[3,24]	124.4	88.1	16.0	64.0	100.0	169.0	360.1	1.0	1000
## resid[3,25]	15.7	16.0	0.0	4.0	9.0	25.0	49.0	1.0	670
## resid[3,26]	10.1	10.8	0.0	4.0	9.0	16.0	36.0	1.0	750
## resid[3,27]	98.4	34.5	36.0	81.0	100.0	121.0	169.0	1.0	740
## resid[3,28]	892.9	115.4	626.2	841.0	900.0	961.0	1089.0	1.0	500
## resid[3,29]	1.8	3.1	0.0	0.0	1.0	1.0	9.0	1.0	560
## resid[3,30]	1.3	2.2	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[3,31]	318.8	80.2	169.0	256.0	324.0	361.0	484.0	1.0	1000
## resid[3,32]	6.2	10.0	0.0	1.0	4.0	9.0	36.0	1.0	1000
## resid[3,33]	60.8	48.5	4.0	25.0	49.0	81.0	169.0	1.0	1000
## resid[3,34]	67.6	49.3	9.0	36.0	49.0	100.0	196.0	1.0	1000
## resid[3,35]	55.9	39.0	4.0	25.0	49.0	81.0	144.0	1.0	380
## resid[4,1]	2.8	4.1	0.0	0.0	1.0	4.0	16.0	1.0	1000
## resid[4,2]	1.8	3.1	0.0	0.0	1.0	1.0	9.0	1.0	500
## resid[4,3]	1.6	2.8	0.0	0.0	1.0	1.0	9.0	1.0	870
## resid[4,4]	27.1	24.9	1.0	9.0	16.0	36.0	100.0	1.0	1000
## resid[4,5]	22.3	20.0	1.0	9.0	16.0	36.0	81.0	1.0	1000
## resid[4,6]	51.4	57.4	0.0	9.0	36.0	81.0	196.0	1.0	370
## resid[4,7]	46.6	42.0	1.0	16.0	36.0	64.0	169.0	1.0	1000
## resid[4,8]	114.4	69.9	16.0	64.0	100.0	144.0	289.0	1.0	890
## resid[4,9]	31.4	33.4	0.0	9.0	25.0	49.0	121.0	1.0	1000
## resid[4,10]	9.6	14.2	0.0	1.0	4.0	16.0	49.0	1.0	1000
## resid[4,11]	277.0	118.8	81.0	196.0	256.0	361.0	529.0	1.0	630
## resid[4,12]	437.7	165.6	144.0	324.0	441.0	529.0	784.0	1.0	1000
## resid[4,13]	80.0	70.6	1.0	25.0	64.0	100.0	256.0	1.0	1000
## resid[4,14]	180.2	100.1	36.0	100.0	169.0	225.0	400.0	1.0	1000
## resid[4,15]	145.3	86.8	25.0	81.0	121.0	196.0	361.0	1.0	600
## resid[4,16]	24.0	32.3	0.0	4.0	16.0	36.0	100.0	1.0	1000
## resid[4,17]	94.0	71.5	1.0	36.0	81.0	121.0	256.0	1.0	1000
## resid[4,18]	289.3	145.8	81.0	196.0	256.0	361.0	625.0	1.0	1000
## resid[4,19]	0.6	1.4	0.0	0.0	0.0	1.0	4.0	1.0	620
## resid[4,20]	0.8	0.7	0.0	1.0	1.0	1.0	1.0	1.0	1000
## resid[4,21]	79.3	75.6	1.0	25.0	64.0	121.0	256.0	1.0	1000
## resid[4,22]	16539.0	1942.0	12770.0	15192.1	16640.0	17892.1	20450.0	1.0	220
## resid[4,23]	11766.5	2506.2	7396.0	10000.0	11660.0	13230.0	17420.0	1.0	1000



## resid[4,24]	20.5	18.5	1.0	9.0	16.0	25.0	64.0	1.0	1000
## resid[4,25]	12.2	12.7	0.0	4.0	9.0	16.0	49.0	1.0	880
## resid[4,26]	3412.8	255.1	2916.0	3249.0	3364.0	3600.0	3844.0	1.0	460
## resid[4,27]	6288.1	580.9	5184.0	5929.0	6241.0	6724.0	7396.0	1.0	670
## resid[4,28]	7218.8	1779.0	4225.0	5929.0	7056.0	8281.0	11030.0	1.0	1000
## resid[4,29]	2.7	4.1	0.0	0.0	1.0	4.0	16.0	1.0	1000
## resid[4,30]	1.3	2.6	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[4,31]	29.2	25.6	0.0	9.0	25.0	49.0	100.0	1.0	570
## resid[4,32]	8.8	11.1	0.0	1.0	4.0	16.0	36.0	1.0	1000
## resid[4,33]	50.3	40.0	4.0	25.0	36.0	64.0	144.0	1.0	1000
## resid[4,34]	42.1	34.4	1.0	16.0	36.0	64.0	121.0	1.0	900
## resid[4,35]	42.7	32.3	4.0	16.0	36.0	64.0	121.0	1.0	260
## resid[5,1]	2.1	3.4	0.0	0.0	1.0	4.0	9.0	1.0	1000
## resid[5,2]	1.9	3.5	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[5,3]	3.6	5.4	0.0	0.0	1.0	4.0	16.0	1.0	1000
## resid[5,4]	50.5	36.1	4.0	25.0	49.0	64.0	144.0	1.0	1000
## resid[5,5]	57.6	43.2	4.0	25.0	49.0	81.0	169.0	1.0	1000
## resid[5,6]	24.0	32.3	0.0	4.0	9.0	36.0	120.5	1.0	1000
## resid[5,7]	143.9	84.3	25.0	81.0	121.0	196.0	361.0	1.0	1000
## resid[5,8]	515.5	231.5	169.0	361.0	484.0	625.0	1024.0	1.0	970
## resid[5,9]	63.8	45.4	9.0	36.0	49.0	81.0	169.0	1.0	710
## resid[5,10]	1004.1	200.6	625.0	841.0	1024.0	1156.0	1369.0	1.0	1000
## resid[5,11]	71.0	57.1	1.1	25.0	64.0	100.0	225.0	1.0	1000
## resid[5,12]	82.5	71.5	1.0	25.0	64.0	121.0	256.0	1.0	510
## resid[5,13]	279.3	137.6	81.0	169.0	256.0	361.0	576.0	1.0	1000
## resid[5,14]	241.9	118.3	64.0	144.0	225.0	324.0	529.0	1.0	1000
## resid[5,15]	209.6	116.6	49.0	121.0	196.0	256.0	484.0	1.0	1000
## resid[5,16]	73.8	60.1	1.0	25.0	64.0	100.0	225.0	1.0	1000
## resid[5,17]	241.6	126.4	64.0	144.0	225.0	324.0	529.0	1.0	1000
## resid[5,18]	330.7	162.1	81.0	225.0	324.0	400.0	727.6	1.0	130
## resid[5,19]	0.7	1.5	0.0	0.0	0.0	1.0	4.0	1.0	1000
## resid[5,20]	1810.9	56.2	1681.0	1764.0	1849.0	1849.0	1849.0	1.0	230
## resid[5,21]	163.2	93.7	36.0	100.0	144.0	225.0	399.0	1.0	690
## resid[5,22]	6484.2	784.8	4900.0	5929.0	6561.0	7056.0	7921.0	1.0	1000
## resid[5,23]	1212.2	423.2	530.1	900.0	1156.0	1444.0	2209.0	1.0	910
## resid[5,24]	16.9	17.1	0.0	4.0	9.0	25.0	64.0	1.0	1000
## resid[5,25]	13.0	15.6	0.0	4.0	9.0	16.0	64.0	1.0	1000
## resid[5,26]	269.1	52.0	169.0	225.0	289.0	314.9	361.0	1.0	380
## resid[5,27]	789.1	133.1	529.0	676.0	784.0	900.0	1024.0	1.0	1000
## resid[5,28]	174.3	100.4	36.0	100.0	144.0	225.0	400.0	1.0	280
## resid[5,29]	2.5	3.8	0.0	0.0	1.0	4.0	15.8	1.0	1000
## resid[5,30]	4.7	3.3	0.0	1.0	4.0	9.0	9.0	1.0	1000
## resid[5,31]	16.0	14.5	0.0	4.0	16.0	25.0	49.0	1.0	1000
## resid[5,32]	32.0	31.3	0.0	9.0	25.0	49.0	121.0	1.0	600
## resid[5,33]	964.8	196.9	576.0	841.0	961.0	1089.0	1296.0	1.0	1000
## resid[5,34]	57.0	42.5	9.0	25.0	49.0	81.0	169.0	1.0	1000
## resid[5,35]	68.6	48.0	9.0	36.0	64.0	100.0	196.0	1.0	1000
## resid[6,1]	3.5	5.1	0.0	1.0	1.0	4.0	16.0	1.0	230
## resid[6,2]	2.6	4.0	0.0	0.0	1.0	4.0	16.0	1.0	540
## resid[6,3]	1.1	2.3	0.0	0.0	1.0	1.0	4.0	1.1	1000
## resid[6,4]	48.4	36.6	4.0	25.0	36.0	64.0	144.0	1.0	1000
## resid[6,5]	123.9	74.9	16.0	64.0	100.0	169.0	324.0	1.0	530
## resid[6,6]	25.9	31.3	0.0	4.0	16.0	36.0	121.0	1.0	970
## resid[6,7]	157.3	86.3	36.0	100.0	144.0	196.0	361.0	1.0	1000

## resid[6,8]	133.0	71.8	16.0	81.0	121.0	169.0	289.0	1.0	1000
## resid[6,9]	33.5	33.0	1.0	9.0	25.0	49.0	121.0	1.0	1000
## resid[6,10]	1287.9	275.1	784.0	1089.0	1296.0	1521.0	1846.8	1.0	360
## resid[6,11]	120.9	93.5	1.0	49.0	100.0	169.0	361.0	1.0	1000
## resid[6,12]	195.5	118.2	36.0	121.0	169.0	256.0	484.0	1.0	1000
## resid[6,13]	148.9	82.7	25.0	81.0	144.0	196.0	324.0	1.0	1000
## resid[6,14]	317.3	149.1	100.0	196.0	289.0	400.0	676.0	1.0	910
## resid[6,15]	305.3	212.6	9.0	144.0	256.0	441.0	784.0	1.0	1000
## resid[6,16]	30.1	41.7	0.0	4.0	16.0	36.0	144.0	1.0	430
## resid[6,17]	261.8	134.0	81.0	169.0	256.0	324.0	576.0	1.0	1000
## resid[6,18]	237.4	116.2	49.0	144.0	225.0	324.0	484.0	1.0	1000
## resid[6,19]	0.9	1.6	0.0	0.0	0.0	1.0	4.0	1.0	770
## resid[6,20]	2496.0	174.9	2116.0	2401.0	2500.0	2601.0	2809.0	1.0	1000
## resid[6,21]	440.7	513.0	1.0	81.0	256.0	625.0	1849.0	1.0	400
## resid[6,22]	74.4	82.6	0.0	16.0	49.0	100.0	289.0	1.0	1000
## resid[6,23]	180.0	104.4	36.0	100.0	169.0	225.0	441.0	1.0	1000
## resid[6,24]	21.3	19.3	1.0	9.0	16.0	25.0	81.0	1.0	1000
## resid[6,25]	131.3	130.1	1.0	36.0	100.0	196.0	484.0	1.0	1000
## resid[6,26]	2072.0	585.5	1089.0	1600.0	2025.0	2401.0	3249.0	1.0	1000
## resid[6,27]	29.9	30.3	0.0	4.0	16.0	49.0	100.0	1.0	1000
## resid[6,28]	9.1	9.8	0.0	1.0	4.0	16.0	36.0	1.0	330
## resid[6,29]	1.8	3.0	0.0	0.0	1.0	1.0	9.0	1.0	630
## resid[6,30]	59.4	16.9	25.0	49.0	64.0	81.0	81.0	1.0	1000
## resid[6,31]	13.3	18.0	0.0	1.0	9.0	16.0	64.0	1.0	1000
## resid[6,32]	56.0	40.9	4.0	25.0	49.0	81.0	144.0	1.0	820
## resid[6,33]	51.3	41.0	4.0	25.0	36.0	64.0	144.0	1.0	690
## resid[6,34]	78.1	53.9	9.0	36.0	64.0	100.0	196.0	1.0	1000
## resid[6,35]	263.5	303.3	1.0	39.2	144.0	361.0	1154.3	1.0	550
## resid[7,1]	4.0	5.4	0.0	1.0	1.0	4.0	16.0	1.0	1000
## resid[7,2]	3.9	5.5	0.0	1.0	1.0	4.0	16.0	1.0	590
## resid[7,3]	1.1	2.1	0.0	0.0	1.0	1.0	9.0	1.0	380
## resid[7,4]	60.0	43.3	9.0	25.0	49.0	81.0	169.0	1.0	400
## resid[7,5]	154.2	92.6	25.2	81.0	144.0	196.0	400.0	1.0	1000
## resid[7,6]	219.5	108.9	49.0	144.0	196.0	289.0	484.0	1.0	1000
## resid[7,7]	379.8	173.6	121.0	256.0	361.0	484.0	784.0	1.0	1000
## resid[7,8]	167.9	98.9	25.2	100.0	144.0	225.0	400.0	1.0	1000
## resid[7,9]	14.8	22.5	0.0	1.0	4.0	16.0	81.0	1.0	1000
## resid[7,10]	6231.8	571.6	5184.0	5776.0	6241.0	6561.0	7225.0	1.0	1000
## resid[7,11]	460.5	216.5	144.0	289.0	441.0	576.0	961.0	1.0	1000
## resid[7,12]	526.4	226.8	169.0	361.0	484.0	676.0	1089.0	1.0	1000
## resid[7,13]	322.4	154.9	100.0	225.0	289.0	400.0	729.0	1.0	620
## resid[7,14]	557.1	232.9	196.0	400.0	529.0	676.0	1089.0	1.0	1000
## resid[7,15]	221.4	193.5	1.0	81.0	169.0	324.0	676.0	1.0	1000
## resid[7,16]	348.0	184.3	81.0	225.0	324.0	441.0	841.0	1.0	550
## resid[7,17]	711.3	297.5	256.8	484.0	676.0	900.0	1369.0	1.0	1000
## resid[7,18]	338.5	154.9	100.0	225.0	324.0	441.0	727.6	1.0	410
## resid[7,19]	2.4	1.7	0.0	1.0	4.0	4.0	4.0	1.0	1000
## resid[7,20]	105.7	16.5	64.0	100.0	100.0	121.0	121.0	1.0	1000
## resid[7,21]	5050.3	1413.2	2500.0	3969.0	5041.0	5929.0	7744.0	1.0	470
## resid[7,22]	1621.0	521.7	729.0	1225.0	1600.0	1936.0	2809.0	1.0	1000
## resid[7,23]	1043.3	399.9	400.0	742.4	1024.0	1296.0	2025.0	1.0	1000
## resid[7,24]	39.1	32.3	4.0	16.0	36.0	49.0	121.0	1.0	1000
## resid[7,25]	7.9	9.6	0.0	1.0	4.0	9.0	36.0	1.0	1000
## resid[7,26]	7005.7	1079.3	5041.0	6241.0	6889.0	7744.0	9216.0	1.0	760

## resid[7,27]	282.9	139.9	64.4	169.0	256.0	361.0	576.0	1.0	1000
## resid[7,28]	21.1	23.0	0.0	4.0	16.0	25.0	81.0	1.0	1000
## resid[7,29]	2.2	3.4	0.0	0.0	1.0	4.0	9.0	1.0	450
## resid[7,30]	2627.8	142.2	2304.0	2500.0	2601.0	2704.0	2809.0	1.0	680
## resid[7,31]	117.6	74.1	25.0	64.0	100.0	144.0	289.0	1.0	1000
## resid[7,32]	44.1	45.2	0.0	9.0	25.0	64.0	169.0	1.0	830
## resid[7,33]	67.5	55.9	1.0	25.0	64.0	100.0	196.0	1.0	1000
## resid[7,34]	132.5	79.7	25.0	81.0	121.0	169.0	324.0	1.0	1000
## resid[7,35]	1806.3	267.9	1296.0	1600.0	1849.0	2025.0	2304.0	1.0	1000
## resid[8,1]	4.0	5.6	0.0	1.0	1.0	4.0	25.0	1.0	1000
## resid[8,2]	0.9	1.4	0.0	0.0	1.0	1.0	4.0	1.0	1000
## resid[8,3]	39.4	30.7	4.0	16.0	36.0	49.0	121.0	1.0	1000
## resid[8,4]	41.5	33.3	4.0	16.0	36.0	49.0	121.0	1.0	720
## resid[8,5]	103.0	66.1	16.0	64.0	81.0	144.0	256.0	1.0	830
## resid[8,6]	374.0	184.0	100.0	256.0	361.0	441.0	841.0	1.0	530
## resid[8,7]	31.5	37.4	0.0	4.0	16.0	49.0	143.4	1.0	740
## resid[8,8]	34.8	26.5	1.0	16.0	25.0	49.0	100.0	1.0	720
## resid[8,9]	79.7	51.0	4.0	36.0	81.0	100.0	196.0	1.0	1000
## resid[8,10]	732.5	187.5	361.0	576.0	729.0	841.0	1089.0	1.0	600
## resid[8,11]	480.9	273.5	100.0	289.0	441.0	625.0	1156.0	1.0	1000
## resid[8,12]	62.1	55.5	1.0	25.0	49.0	81.0	196.0	1.0	1000
## resid[8,13]	274.4	137.4	81.0	169.0	256.0	361.0	625.0	1.0	1000
## resid[8,14]	346.7	165.4	100.0	225.0	324.0	441.0	729.0	1.0	470
## resid[8,15]	76.5	95.2	0.0	9.0	49.0	100.0	361.0	1.0	1000
## resid[8,16]	1018.1	365.1	441.0	729.0	961.0	1225.0	1849.0	1.0	770
## resid[8,17]	322.4	152.0	64.4	225.0	324.0	400.0	625.0	1.0	1000
## resid[8,18]	0.7	1.5	0.0	0.0	0.0	1.0	4.0	1.0	1000
## resid[8,19]	20.3	5.8	9.0	16.0	25.0	25.0	25.0	1.0	1000
## resid[8,20]	2.5	1.7	0.0	1.0	4.0	4.0	4.0	1.0	1000
## resid[8,21]	3028.8	966.6	1444.0	2304.0	2916.0	3600.0	5180.4	1.0	400
## resid[8,22]	64.4	52.4	1.0	25.0	49.0	100.0	196.0	1.0	1000
## resid[8,23]	23.3	20.4	1.0	9.0	16.0	36.0	81.0	1.0	1000
## resid[8,24]	10.0	10.1	0.0	1.0	9.0	16.0	36.0	1.0	1000
## resid[8,25]	1419.4	195.8	1024.0	1296.0	1444.0	1521.0	1764.0	1.0	390
## resid[8,26]	151.8	139.7	1.0	49.0	121.0	225.0	529.0	1.0	1000
## resid[8,27]	77.9	30.2	25.0	49.0	81.0	100.0	144.0	1.0	540
## resid[8,28]	1.7	3.1	0.0	0.0	1.0	1.0	9.0	1.0	670
## resid[8,29]	5.1	3.3	0.0	4.0	4.0	9.0	9.0	1.0	1000
## resid[8,30]	6991.3	998.9	5041.0	6241.0	7056.0	7744.0	9025.0	1.0	1000
## resid[8,31]	31547.2	5739.8	21610.0	27560.0	31330.0	34970.0	43249.7	1.0	1000
## resid[8,32]	21523.7	780.2	19880.0	21030.0	21610.0	22200.0	22800.0	1.0	1000
## resid[8,33]	76.3	51.6	9.0	36.0	64.0	100.0	196.0	1.0	1000
## resid[8,34]	54.9	40.6	1.0	25.0	49.0	81.0	144.0	1.0	600
## resid[8,35]	3210.6	412.5	2401.0	2916.0	3249.0	3481.0	3969.0	1.0	1000
## resid[9,1]	1.1	2.1	0.0	0.0	0.0	1.0	9.0	1.0	1000
## resid[9,2]	0.8	1.0	0.0	0.0	1.0	1.0	4.0	1.0	1000
## resid[9,3]	23.3	21.9	1.0	9.0	16.0	36.0	81.0	1.0	770
## resid[9,4]	15.3	18.8	0.0	1.0	9.0	25.0	64.0	1.0	350
## resid[9,5]	4306.3	1166.5	2304.0	3481.0	4225.0	4900.0	6889.0	1.0	1000
## resid[9,6]	86.4	57.9	9.0	49.0	81.0	121.0	225.0	1.0	1000
## resid[9,7]	126.8	79.4	25.0	64.0	121.0	169.0	324.0	1.0	1000
## resid[9,8]	109.9	52.1	25.0	81.0	100.0	144.0	225.0	1.0	770
## resid[9,9]	86.3	73.2	1.0	25.0	64.0	121.0	256.0	1.0	1000
## resid[9,10]	1208.3	220.6	784.0	1089.0	1225.0	1369.0	1600.0	1.0	220

## resid[9,11]	107.1	67.7	16.2	49.0	100.0	144.0	289.0	1.0	1000
## resid[9,12]	388.7	172.4	121.0	256.0	361.0	484.0	784.0	1.0	1000
## resid[9,13]	122.6	84.9	16.0	64.0	100.0	169.0	324.0	1.0	1000
## resid[9,14]	297.7	156.0	49.4	169.0	289.0	400.0	625.0	1.0	1000
## resid[9,15]	39.7	54.7	0.0	4.0	16.0	49.0	196.0	1.0	1000
## resid[9,16]	143.6	82.4	36.0	81.0	121.0	196.0	324.0	1.0	440
## resid[9,17]	219.1	134.6	36.0	121.0	196.0	289.0	576.0	1.0	1000
## resid[9,18]	6.3	3.1	1.0	4.0	9.0	9.0	9.0	1.0	1000
## resid[9,19]	211.8	19.0	169.0	196.0	225.0	225.0	225.0	1.0	420
## resid[9,20]	107.5	16.6	64.0	100.0	121.0	121.0	121.0	1.0	1000
## resid[9,21]	3051.7	862.8	1600.0	2401.0	2916.0	3600.0	4900.0	1.0	1000
## resid[9,22]	536.6	225.9	196.0	400.0	529.0	676.0	1089.0	1.0	1000
## resid[9,23]	7.7	7.8	0.0	1.0	4.0	9.0	25.0	1.0	1000
## resid[9,24]	460.5	107.0	256.0	400.0	484.0	529.0	676.0	1.0	1000
## resid[9,25]	2891.1	327.2	2209.0	2704.0	2916.0	3136.0	3481.0	1.0	490
## resid[9,26]	5751.8	1463.3	3364.0	4624.0	5625.0	6724.0	8836.0	1.0	1000
## resid[9,27]	14.6	18.2	0.0	4.0	9.0	16.0	64.0	1.0	1000
## resid[9,28]	1.7	3.2	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[9,29]	3879.8	156.0	3600.0	3721.0	3844.0	3969.0	4096.0	1.0	420
## resid[9,30]	1288.3	75.3	1089.0	1225.0	1296.0	1369.0	1369.0	1.0	960
## resid[9,31]	75.0	53.0	9.0	36.0	64.0	100.0	196.0	1.0	650
## resid[9,32]	8.7	12.5	0.0	1.0	4.0	9.0	36.0	1.0	500
## resid[9,33]	72.5	48.5	9.0	36.0	64.0	100.0	196.0	1.0	1000
## resid[9,34]	413.2	110.3	196.0	324.0	400.0	484.0	625.0	1.0	980
## resid[9,35]	1227.1	237.4	784.0	1089.0	1225.0	1369.0	1681.0	1.0	1000
## resid[10,1]	0.8	1.2	0.0	0.0	1.0	1.0	4.0	1.0	770
## resid[10,2]	0.8	0.8	0.0	0.0	1.0	1.0	1.0	1.1	430
## resid[10,3]	31.3	27.6	1.0	9.0	25.0	45.8	100.0	1.0	680
## resid[10,4]	36.0	31.9	1.0	16.0	25.0	49.0	121.0	1.0	960
## resid[10,5]	76.0	51.5	9.0	36.0	64.0	100.0	196.0	1.0	620
## resid[10,6]	47.1	38.1	4.0	16.0	36.0	64.0	144.0	1.0	490
## resid[10,7]	45.4	36.6	4.0	16.0	36.0	64.0	144.0	1.0	1000
## resid[10,8]	496.4	108.1	289.0	441.0	484.0	576.0	729.0	1.0	950
## resid[10,9]	8.6	12.4	0.0	1.0	4.0	9.0	36.0	1.0	770
## resid[10,10]	97.9	63.4	16.0	49.0	81.0	144.0	256.0	1.0	1000
## resid[10,11]	54.8	42.3	4.0	25.0	49.0	81.0	168.4	1.0	1000
## resid[10,12]	45.5	36.3	4.0	16.0	36.0	64.0	144.0	1.0	840
## resid[10,13]	365.0	175.3	100.0	225.0	324.0	441.0	784.0	1.0	1000
## resid[10,14]	135.8	101.1	1.1	49.0	121.0	196.0	361.0	1.0	560
## resid[10,15]	892.9	392.5	289.0	625.0	841.0	1089.0	1849.0	1.0	860
## resid[10,16]	153.8	91.6	25.0	81.0	144.0	196.0	361.0	1.0	730
## resid[10,17]	8.6	10.0	0.0	1.0	4.0	16.0	36.0	1.0	1000
## resid[10,18]	336.0	29.7	256.8	324.0	361.0	361.0	361.0	1.0	1000
## resid[10,19]	0.7	0.7	0.0	0.0	1.0	1.0	1.0	1.0	1000
## resid[10,20]	1.0	1.4	0.0	0.0	1.0	1.0	4.0	1.0	1000
## resid[10,21]	163.3	87.7	36.0	100.0	144.0	225.0	361.0	1.0	1000
## resid[10,22]	46.8	35.9	4.0	17.9	36.0	64.0	144.0	1.0	1000
## resid[10,23]	97.3	42.1	25.0	64.0	100.0	121.0	169.0	1.0	580
## resid[10,24]	11.2	12.7	0.0	1.0	4.0	16.0	49.0	1.0	680
## resid[10,25]	1781.4	314.5	1156.0	1600.0	1764.0	2025.0	2401.0	1.0	870
## resid[10,26]	21.1	20.1	1.0	9.0	16.0	25.0	81.0	1.0	1000
## resid[10,27]	3.5	5.1	0.0	1.0	1.0	4.0	16.0	1.0	480
## resid[10,28]	1.9	1.7	0.0	1.0	1.0	4.0	4.0	1.0	1000
## resid[10,29]	6409.4	151.1	6084.0	6400.0	6400.0	6561.0	6561.0	1.0	250

## resid[10,30]	1.0	1.6	0.0	0.0	1.0	1.0	4.0	1.0	1000
## resid[10,31]	961.1	379.9	361.0	729.0	900.0	1156.0	1849.0	1.0	350
## resid[10,32]	9.1	11.5	0.0	1.0	4.0	16.0	36.0	1.0	1000
## resid[10,33]	23.6	31.8	0.0	4.0	9.0	36.0	121.0	1.0	1000
## resid[10,34]	101.6	68.9	16.0	49.0	81.0	144.0	289.0	1.0	630
## resid[10,35]	75.0	62.3	4.0	25.0	64.0	100.0	225.0	1.0	860
## resid[11,1]	6.8	5.3	0.0	4.0	4.0	9.0	16.0	1.0	950
## resid[11,2]	1.5	2.8	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[11,3]	538.3	230.4	196.0	361.0	484.0	676.0	1089.0	1.0	700
## resid[11,4]	512.3	212.5	169.0	361.0	484.0	625.0	1024.0	1.0	1000
## resid[11,5]	3167.6	852.7	1681.0	2601.0	3136.0	3600.0	5041.0	1.0	1000
## resid[11,6]	451.8	193.1	144.0	324.0	441.0	576.0	900.0	1.0	1000
## resid[11,7]	311.7	151.8	100.0	196.0	289.0	400.0	676.0	1.0	1000
## resid[11,8]	7943.5	1564.7	4903.5	6889.0	7921.0	9025.0	11030.0	1.0	760
## resid[11,9]	3569.2	596.1	2403.4	3136.0	3600.0	3969.0	4761.0	1.0	520
## resid[11,10]	873.6	338.4	324.0	625.0	841.0	1089.0	1600.0	1.0	1000
## resid[11,11]	420.4	188.4	144.0	289.0	400.0	529.0	898.5	1.0	350
## resid[11,12]	389.6	175.2	121.0	256.0	361.0	484.0	784.0	1.0	1000
## resid[11,13]	955.4	537.2	196.0	576.0	841.0	1225.0	2209.0	1.0	580
## resid[11,14]	1205.6	627.7	256.0	729.0	1089.0	1600.0	2601.0	1.0	570
## resid[11,15]	6208.1	1563.8	3603.0	5041.0	5929.0	7225.0	9604.0	1.0	1000
## resid[11,16]	466.1	204.6	169.0	324.0	441.0	576.0	900.0	1.0	510
## resid[11,17]	446.0	201.5	144.0	289.0	400.0	576.0	900.0	1.0	1000
## resid[11,18]	306.0	53.8	196.0	256.0	324.0	361.0	400.0	1.0	680
## resid[11,19]	1.7	2.2	0.0	0.0	1.0	4.0	4.0	1.0	430
## resid[11,20]	2.8	3.4	0.0	1.0	1.0	4.0	9.0	1.0	930
## resid[11,21]	388.3	175.2	121.0	256.0	361.0	484.0	784.0	1.0	1000
## resid[11,22]	372.8	177.5	121.0	256.0	324.0	484.0	784.0	1.0	670
## resid[11,23]	36.3	56.0	0.0	4.0	16.0	49.0	196.0	1.0	320
## resid[11,24]	2087.4	511.4	1089.0	1764.0	2116.0	2401.0	3136.0	1.0	1000
## resid[11,25]	476.2	202.7	169.0	324.0	441.0	576.0	900.0	1.0	1000
## resid[11,26]	21.1	20.2	1.0	9.0	16.0	25.0	81.0	1.0	990
## resid[11,27]	20.5	18.9	1.0	9.0	16.0	25.0	64.0	1.0	1000
## resid[11,28]	31041.6	1091.1	28900.0	30280.0	30980.0	31680.0	33120.0	1.0	1000
## resid[11,29]	12255.5	591.3	11030.0	11880.0	12320.0	12770.0	13230.0	1.0	570
## resid[11,30]	14.5	15.2	0.0	4.0	9.0	16.0	49.0	1.0	1000
## resid[11,31]	329.4	159.8	81.0	225.0	324.0	400.0	729.0	1.0	1000
## resid[11,32]	80.0	53.8	9.0	36.0	64.0	100.0	225.0	1.0	1000
## resid[11,33]	246.3	200.1	9.0	100.0	196.0	324.0	784.0	1.0	1000
## resid[11,34]	1278.6	406.8	576.0	961.0	1225.0	1521.0	2116.0	1.0	1000
## resid[11,35]	806.8	316.2	324.0	576.0	729.0	961.0	1521.0	1.0	510
## resid[12,1]	5.6	3.3	0.0	4.0	4.0	9.0	9.0	1.0	1000
## resid[12,2]	1.2	2.1	0.0	0.0	1.0	1.0	9.0	1.0	310
## resid[12,3]	89.4	62.2	16.0	49.0	81.0	121.0	256.0	1.0	1000
## resid[12,4]	103.8	65.1	16.0	49.0	100.0	144.0	256.0	1.0	1000
## resid[12,5]	234.5	125.5	49.0	144.0	225.0	289.0	529.0	1.0	1000
## resid[12,6]	111.2	69.1	16.0	64.0	100.0	144.0	289.0	1.0	1000
## resid[12,7]	35.3	41.6	0.0	4.0	25.0	49.0	144.0	1.0	1000
## resid[12,8]	2697.3	355.0	2025.0	2500.0	2704.0	2916.0	3364.0	1.0	1000
## resid[12,9]	140.1	83.9	36.0	81.0	121.0	196.0	361.0	1.0	420
## resid[12,10]	421.4	199.6	121.0	289.0	400.0	529.0	900.0	1.0	230
## resid[12,11]	119.9	75.6	25.0	64.0	100.0	144.0	324.0	1.0	550
## resid[12,12]	95.4	62.6	16.0	49.0	81.0	121.0	256.0	1.0	210
## resid[12,13]	398.6	210.4	100.0	256.0	361.0	529.0	900.0	1.0	1000

## resid[12,14]	724.1	291.9	256.0	529.0	676.0	900.0	1369.0	1.0	1000
## resid[12,15]	1200.1	435.5	484.0	900.0	1156.0	1444.0	2116.0	1.0	1000
## resid[12,16]	133.6	82.2	25.0	81.0	121.0	169.0	324.0	1.0	1000
## resid[12,17]	1190.6	224.7	729.0	1024.0	1190.0	1369.0	1600.0	1.0	730
## resid[12,18]	104.7	17.9	64.0	100.0	100.0	121.0	121.0	1.0	300
## resid[12,19]	1.9	3.2	0.0	0.0	1.0	1.0	9.0	1.0	1000
## resid[12,20]	2.6	4.1	0.0	0.0	1.0	4.0	16.0	1.0	1000
## resid[12,21]	90.0	57.9	16.0	49.0	81.0	121.0	225.0	1.0	1000
## resid[12,22]	95.7	59.8	16.0	49.0	81.0	121.0	225.0	1.0	650
## resid[12,23]	62.7	42.3	1.0	36.0	49.0	81.0	169.0	1.0	810
## resid[12,24]	811.2	176.4	442.0	676.0	784.0	961.0	1156.0	1.0	1000
## resid[12,25]	100.9	63.3	16.0	49.0	81.0	144.0	256.0	1.0	340
## resid[12,26]	5.8	8.3	0.0	1.0	4.0	9.0	25.0	1.0	500
## resid[12,27]	6.4	8.2	0.0	1.0	4.0	9.0	25.0	1.0	320
## resid[12,28]	1878.7	109.8	1602.0	1849.0	1936.0	1936.0	2025.0	1.0	660
## resid[12,29]	135.1	26.8	81.0	121.0	144.0	144.0	169.0	1.0	1000
## resid[12,30]	11.0	11.9	0.0	4.0	9.0	16.0	36.0	1.0	1000
## resid[12,31]	25.3	23.6	1.0	9.0	16.0	36.0	81.0	1.0	830
## resid[12,32]	4381.2	298.2	3721.0	4225.0	4356.0	4624.0	4900.0	1.0	1000
## resid[12,33]	17.0	30.4	0.0	1.0	4.0	16.0	100.0	1.0	1000
## resid[12,34]	35.0	50.7	0.0	4.0	16.0	49.0	169.0	1.0	1000
## resid[12,35]	277.1	136.8	81.0	169.0	256.0	361.0	576.0	1.0	1000
## resid[13,1]	1.8	1.9	0.0	1.0	1.0	4.0	4.0	1.0	1000
## resid[13,2]	3.6	4.7	0.0	1.0	1.0	4.0	16.0	1.0	600
## resid[13,3]	224.5	111.1	64.0	144.0	196.0	289.0	484.0	1.0	1000
## resid[13,4]	313.6	147.5	100.0	196.0	289.0	400.0	625.0	1.0	1000
## resid[13,5]	480.6	207.7	169.0	324.0	441.0	625.0	961.0	1.0	1000
## resid[13,6]	256.9	127.0	64.0	169.0	225.0	324.0	576.0	1.0	1000
## resid[13,7]	504.4	221.4	169.0	332.9	484.0	625.0	1024.0	1.0	420
## resid[13,8]	1360.5	348.6	729.0	1089.0	1369.0	1600.0	2116.0	1.0	1000
## resid[13,9]	425.5	190.9	144.0	289.0	400.0	529.0	841.0	1.0	1000
## resid[13,10]	39.0	49.6	0.0	4.0	25.0	49.0	169.0	1.0	1000
## resid[13,11]	360.8	165.6	121.0	232.4	324.0	441.0	729.0	1.0	1000
## resid[13,12]	376.8	245.8	49.0	196.0	324.0	529.0	961.0	1.0	510
## resid[13,13]	1726.3	731.1	625.0	1225.0	1681.0	2116.0	3481.0	1.0	370
## resid[13,14]	2327.2	688.3	1225.0	1849.0	2209.0	2704.0	3844.0	1.0	1000
## resid[13,15]	1455.1	489.4	676.0	1089.0	1369.0	1764.0	2601.0	1.0	1000
## resid[13,16]	405.3	183.5	121.0	289.0	400.0	529.0	841.0	1.0	900
## resid[13,17]	13775.1	1639.7	10400.0	12770.0	13690.0	14880.0	17160.0	1.0	840
## resid[13,18]	87.1	23.0	49.0	67.9	81.0	100.0	121.0	1.0	1000
## resid[13,19]	2.3	4.4	0.0	0.0	1.0	4.0	16.0	1.0	850
## resid[13,20]	1.4	2.5	0.0	0.0	1.0	1.0	9.0	1.0	640
## resid[13,21]	250.8	129.2	64.0	169.0	225.0	324.0	529.0	1.0	1000
## resid[13,22]	2546.9	1232.8	676.0	1681.0	2304.0	3334.9	5472.3	1.0	620
## resid[13,23]	1571.6	515.5	625.0	1225.0	1521.0	1936.0	2601.0	1.0	1000
## resid[13,24]	92.5	92.2	1.0	25.0	64.0	121.0	324.0	1.0	1000
## resid[13,25]	147.2	82.4	36.0	81.0	144.0	196.0	361.0	1.0	1000
## resid[13,26]	14.6	16.0	0.0	4.0	9.0	16.0	49.0	1.0	1000
## resid[13,27]	24.8	31.9	0.0	4.0	16.0	36.0	121.0	1.0	290
## resid[13,28]	34630.1	801.1	33120.0	34230.0	34600.0	35340.0	36100.0	1.0	1000
## resid[13,29]	8.1	9.5	0.0	1.0	4.0	9.0	36.0	1.0	370
## resid[13,30]	6.9	8.9	0.0	1.0	4.0	9.0	36.0	1.0	220
## resid[13,31]	87.2	60.0	9.0	49.0	81.0	121.0	225.0	1.0	1000
## resid[13,32]	2031.3	542.3	1024.0	1681.0	2025.0	2401.0	3136.0	1.0	1000

```
## resid[13,33] 447.0 282.7 36.0 256.0 400.0 625.0 1089.0 1.0 480
## resid[13,34] 1313.0 436.4 625.0 1024.0 1296.0 1521.0 2304.0 1.0 1000
## resid[13,35] 390.3 179.6 121.0 256.0 361.0 484.0 841.0 1.0 1000
## resid[14,1] 6.5 3.0 1.0 4.0 9.0 9.0 9.0 1.0 1000
## resid[14,2] 1.9 3.3 0.0 0.0 1.0 1.0 9.0 1.0 1000
## resid[14,3] 71.6 53.1 9.0 36.0 64.0 100.0 196.0 1.0 1000
## resid[14,4] 154.5 89.4 25.0 81.0 144.0 196.0 361.0 1.0 1000
## resid[14,5] 89.4 55.0 16.0 49.0 81.0 121.0 225.0 1.0 1000
## resid[14,6] 61.1 46.0 9.0 25.0 49.0 81.0 196.0 1.0 1000
## resid[14,7] 9.2 13.9 0.0 1.0 4.0 9.0 49.0 1.0 1000
## resid[14,8] 109.7 67.6 16.0 64.0 100.0 144.0 256.0 1.0 390
## resid[14,9] 154.5 92.2 25.0 81.0 144.0 196.0 400.0 1.0 1000
## resid[14,10] 13.3 21.8 0.0 1.0 4.0 16.0 81.0 1.0 1000
## resid[14,11] 103.7 64.3 16.0 64.0 100.0 144.0 256.0 1.0 1000
## resid[14,12] 229.1 123.0 49.0 144.0 196.0 289.0 484.0 1.0 260
## resid[14,13] 50.0 69.2 0.0 4.0 25.0 64.0 256.0 1.0 930
## resid[14,14] 960.0 370.1 400.0 676.0 900.0 1156.0 1764.0 1.0 1000
## resid[14,15] 517.6 223.9 169.0 361.0 484.0 625.0 1024.0 1.0 1000
## resid[14,16] 147.5 84.3 25.0 81.0 121.0 196.0 361.0 1.0 1000
## resid[14,17] 1576.9 275.4 1089.0 1369.0 1600.0 1764.0 2116.0 1.0 550
## resid[14,18] 2.3 1.7 0.0 1.0 1.0 4.0 4.0 1.0 1000
## resid[14,19] 8.3 5.3 0.0 4.0 9.0 16.0 16.0 1.0 1000
## resid[14,20] 1.5 2.8 0.0 0.0 1.0 1.0 9.0 1.0 860
## resid[14,21] 87.4 55.6 9.0 49.0 81.0 121.0 225.0 1.0 1000
## resid[14,22] 592.3 143.6 289.8 484.0 576.0 676.0 841.0 1.0 1000
## resid[14,23] 6002.7 576.5 4761.0 5625.0 6084.0 6400.0 7056.0 1.0 1000
## resid[14,24] 209.5 113.0 49.0 121.0 196.0 256.0 484.0 1.0 550
## resid[14,25] 127.4 79.7 25.0 81.0 110.0 169.0 324.0 1.0 1000
## resid[14,26] 2.7 5.0 0.0 0.0 1.0 4.0 16.0 1.0 1000
## resid[14,27] 172.6 35.2 100.0 144.0 169.0 196.0 225.0 1.0 1000
## resid[14,28] 2893.0 122.2 2601.0 2809.0 2916.0 3025.0 3025.0 1.0 1000
## resid[14,29] 6.5 7.9 0.0 1.0 4.0 9.0 25.0 1.0 380
## resid[14,30] 5.2 7.3 0.0 1.0 4.0 9.0 25.0 1.0 490
## resid[14,31] 14.1 18.5 0.0 1.0 9.0 16.0 64.0 1.0 1000
## resid[14,32] 224.8 64.2 100.0 169.0 225.0 256.0 361.0 1.0 610
## resid[14,33] 70.7 56.1 1.1 25.0 64.0 100.0 196.0 1.0 1000
## resid[14,34] 917.0 348.0 361.0 676.0 900.0 1089.0 1681.0 1.0 1000
## resid[14,35] 307.6 160.4 81.0 196.0 289.0 400.0 676.0 1.0 1000
## deviance 19786.2 6.5 19780.0 19780.0 19790.0 19790.0 19800.0 1.0 1000
##
```

```
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 17.5 and DIC = 19803.8
## DIC is an estimate of expected predictive error (lower deviance is better).
```

```
#Rhat
summary(Rhat <- (Resul5$summary[, "Rhat"]))
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.9995 1.0008 1.0023 1.0049 1.0054 1.1053
```

```
summary(Resul5$summary[, "n.eff"])
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    130.0   670.0  1000.0   839.5  1000.0  1000.0
```

```
Resul5$DIC
```

```
## [1] 19803.8
```

```
Resul5$pD
```

```
## [1] 17.536
```

```
#MSE
```

```
mean(Resul5$mean$resid)
```

```
## [1] 937.8168
```

```
set.seed(22)
```

```
Modelo <- function(){
  for (i in 1:14){
    for (j in 1:35){
      Captures[i,j] ~ dpois(lambda[i,j])
      log(lambda[i,j]) <- beta0 + beta1*TMin[i,j] + beta2*Precipita[i,j] +
        beta3*AltMed[i,j] + beta4*Especies[i,j] + v[i]
    }
    v[i] ~ dnorm(0,tauv)
  }
  tauv <- 1/pow(sdv,2)
  sdv ~ dunif(0,10)

  #distribuciones iniciales
  beta0 ~ dflat()
  beta1 ~ dflat()
  beta2 ~ dflat()
  beta3 ~ dflat()
  beta4 ~ dflat()

  for (i in 1:14) {
    for (j in 1:35) {
      Captures.pred[i,j] ~ dpois(lambda[i,j])
      resid[i,j] <- pow(Captures[i,j] - Captures.pred[i,j],2)
    }
  }
}
```

```
#Datos
```

```
Datos <- list(Captures = matrix(DatosModeloSIN5$Captures, nrow = 14),
              TMin = matrix(DatosModeloSIN5$TMinMed, nrow = 14),
```



```

Precipita = matrix(DatosModeloSIN5$Precipita, nrow = 14),
AltMed = matrix(scale(DatosModeloSIN5$AltMed), nrow = 14),
Especies = matrix(scale(DatosModeloSIN5$SumaEspecies), nrow = 14))

#Iniciales
Iniciales <- function(){
  list(beta0 = rnorm(1), v = rnorm(14), sdv = runif(1,0,3),
        beta1 = rnorm(1), beta2 = rnorm(1), beta3 = rnorm(1), beta4 = rnorm(1))
}

#Parámetros
Param <- c("beta0","beta1","beta2","beta3","beta4", "v", "sdv", "resid", "Captures.pred")

```

ResulDef

```

## Inference for Bugs model at "C:/Users/celia/AppData/Local/Temp/RtmpAXk7nM/model15d832377890.txt", fi
## 3 chains, each with 50000 iterations (first 25000 discarded), n.thin = 75
## n.sims = 1002 iterations saved
##
##               mean      sd    2.5%    25%    50%    75%    97.5%
## beta0           1.8    0.1     1.6     1.8     1.8     1.9     2.0
## beta1           0.0    0.0     0.0     0.0     0.0     0.0     0.0
## beta2           0.2    0.0     0.2     0.2     0.2     0.2     0.2
## beta3           0.4    0.0     0.3     0.4     0.4     0.4     0.4
## beta4           0.3    0.0     0.3     0.3     0.3     0.3     0.3
## v[1]            0.4    0.1     0.1     0.3     0.4     0.4     0.6
## v[2]           -0.2    0.1    -0.4    -0.2    -0.2    -0.1     0.1
## v[3]           -0.3    0.1    -0.5    -0.3    -0.3    -0.2    -0.1
## v[4]           -0.3    0.1    -0.5    -0.4    -0.3    -0.2    -0.1
## v[5]           -0.1    0.1    -0.4    -0.2    -0.1    -0.1     0.1
## v[6]            0.0    0.1    -0.2    -0.1     0.0     0.1     0.2
## v[7]            0.1    0.1    -0.1     0.1     0.1     0.2     0.4
## v[8]            0.0    0.1    -0.2     0.0     0.0     0.1     0.2
## v[9]           -0.3    0.1    -0.6    -0.4    -0.3    -0.3    -0.1
## v[10]          -0.4    0.1    -0.6    -0.4    -0.4    -0.3    -0.1
## v[11]           0.7    0.1     0.5     0.6     0.7     0.7     0.9
## v[12]           0.0    0.1    -0.2     0.0     0.0     0.1     0.3
## v[13]           0.5    0.1     0.3     0.5     0.5     0.6     0.8
## v[14]          -0.1    0.1    -0.3    -0.2    -0.1     0.0     0.1
## sdv            0.4    0.1     0.2     0.3     0.4     0.4     0.6
## resid[1,1]      30.4   25.7     1.0    10.8    25.0    36.0    100.0
## resid[1,2]      30.6   25.2     1.0     9.0    25.0    49.0    100.0
## resid[1,3]      74.8   52.5     9.0    36.0    64.0   100.0   196.0
## resid[1,4]     150.0   87.0    36.0    81.0   121.0   196.0   361.0
## resid[1,5]      89.2   58.8    16.0    49.0    81.0   121.0   255.2
## resid[1,6]     100.2   63.4    16.0    49.0    81.0   144.0   256.0
## resid[1,7]      40.2   42.0     0.0     9.0    25.0    49.0   144.0
## resid[1,8]     173.1   96.5    36.0   100.0   169.0   225.0   400.0
## resid[1,9]     625.2  254.6   225.0   441.0   576.0   784.0  1225.0
## resid[1,10]    1144.5  234.4   676.0   961.0  1156.0  1296.0  1600.0
## resid[1,11]     266.8  138.9    64.0   169.0   256.0   324.0   576.0
## resid[1,12]     163.3  119.2    16.0    81.0   144.0   225.0   482.9
## resid[1,13]     455.6  241.9    81.0   289.0   400.0   576.0  1024.0
## resid[1,14]    2923.8  814.7  1600.0  2304.0  2916.0  3481.0  4624.0

```

## resid[1,15]	177.9	127.1	16.0	81.0	144.0	256.0	484.0
## resid[1,16]	87.7	59.5	9.0	49.0	81.0	121.0	255.2
## resid[1,17]	159.2	71.4	36.0	100.0	144.0	196.0	324.0
## resid[1,18]	15.4	20.4	0.0	1.0	9.0	25.0	64.0
## resid[1,19]	146.3	86.7	25.0	81.0	121.0	196.0	361.0
## resid[1,20]	764.7	163.0	442.0	625.0	784.0	900.0	1089.0
## resid[1,21]	178.7	99.8	49.0	100.0	169.0	225.0	400.0
## resid[1,22]	106.1	72.8	4.0	49.0	100.0	144.0	256.0
## resid[1,23]	5775.7	667.7	4489.0	5329.0	5776.0	6241.0	7056.0
## resid[1,24]	1522.2	513.8	625.0	1156.0	1444.0	1849.0	2601.0
## resid[1,25]	78.4	66.7	1.0	25.0	64.0	121.0	225.0
## resid[1,26]	155.3	91.9	25.0	81.0	144.0	196.0	361.0
## resid[1,27]	22.5	33.4	0.0	1.0	9.0	25.0	121.0
## resid[1,28]	7068.1	780.6	5476.0	6561.0	7056.0	7569.0	8649.0
## resid[1,29]	881.9	333.2	324.9	625.0	841.0	1089.0	1681.0
## resid[1,30]	270.2	142.8	64.0	169.0	256.0	361.0	625.0
## resid[1,31]	29.0	29.6	0.0	9.0	16.0	36.0	121.0
## resid[1,32]	130.7	80.2	16.0	64.0	121.0	169.0	324.0
## resid[1,33]	1268.3	247.8	841.0	1089.0	1296.0	1444.0	1764.0
## resid[1,34]	294.5	148.2	81.0	196.0	256.0	361.0	674.7
## resid[1,35]	67.2	50.8	1.0	25.0	64.0	100.0	196.0
## resid[2,1]	13.4	13.8	0.0	4.0	9.0	16.0	49.0
## resid[2,2]	12.5	14.4	0.0	4.0	9.0	16.0	49.0
## resid[2,3]	42.9	33.0	4.0	16.0	36.0	64.0	121.0
## resid[2,4]	32.0	27.4	1.0	16.0	25.0	49.0	100.0
## resid[2,5]	32.7	28.2	1.0	16.0	25.0	49.0	100.0
## resid[2,6]	28.5	28.9	0.0	9.0	16.0	36.0	100.0
## resid[2,7]	8.2	11.3	0.0	1.0	4.0	9.0	36.0
## resid[2,8]	95.7	61.4	16.0	49.0	81.0	121.0	256.0
## resid[2,9]	136.9	81.5	25.0	81.0	121.0	169.0	324.0
## resid[2,10]	22.2	29.1	0.0	4.0	9.0	25.0	100.0
## resid[2,11]	137.2	82.7	25.0	81.0	121.0	169.0	361.0
## resid[2,12]	40.4	48.1	0.0	9.0	25.0	49.0	169.0
## resid[2,13]	256.5	218.5	1.0	100.0	196.0	361.0	841.0
## resid[2,14]	160.3	92.3	36.0	100.0	144.0	225.0	361.0
## resid[2,15]	546.2	227.6	196.0	400.0	484.0	676.0	1089.0
## resid[2,16]	10.2	11.6	0.0	1.0	4.0	16.0	36.0
## resid[2,17]	60.2	44.4	1.0	25.0	49.0	81.0	169.0
## resid[2,18]	719.1	182.6	400.0	576.0	729.0	841.0	1089.0
## resid[2,19]	46.5	36.9	4.0	18.2	36.0	64.0	144.0
## resid[2,20]	52.6	46.0	1.0	16.0	42.5	64.0	169.0
## resid[2,21]	52.7	50.5	1.0	16.0	36.0	81.0	196.0
## resid[2,22]	110.1	68.2	4.1	64.0	100.0	144.0	256.0
## resid[2,23]	4901.9	1149.5	2916.0	4096.0	4900.0	5625.0	7225.0
## resid[2,24]	392.5	184.5	121.0	256.0	361.0	484.0	784.0
## resid[2,25]	509.2	225.8	196.0	324.0	484.0	625.0	1024.0
## resid[2,26]	16.7	22.2	0.0	1.0	9.0	25.0	81.0
## resid[2,27]	50.4	49.6	0.0	16.0	36.0	81.0	169.0
## resid[2,28]	7512.4	834.6	5929.0	6889.0	7569.0	8100.0	9211.2
## resid[2,29]	121.1	72.2	25.0	64.0	100.0	144.0	289.0
## resid[2,30]	354.0	167.7	100.0	225.0	324.0	441.0	729.0
## resid[2,31]	27.1	24.5	0.0	9.0	25.0	36.0	100.0
## resid[2,32]	15.0	19.6	0.0	1.0	9.0	25.0	64.0
## resid[2,33]	8.8	13.2	0.0	1.0	4.0	9.0	49.0

## resid[2,34]	28.8	32.0	0.0	9.0	16.0	36.0	121.0
## resid[2,35]	486.1	208.1	169.0	324.0	441.0	625.0	961.0
## resid[3,1]	16.3	15.9	1.0	4.0	9.0	25.0	49.0
## resid[3,2]	15.8	15.7	0.0	4.0	9.0	25.0	64.0
## resid[3,3]	25.5	24.5	1.0	9.0	16.0	36.0	100.0
## resid[3,4]	24.6	22.7	1.0	9.0	16.0	36.0	81.0
## resid[3,5]	16.4	16.3	0.0	4.0	9.0	25.0	64.0
## resid[3,6]	30.5	31.1	1.0	9.0	25.0	49.0	100.0
## resid[3,7]	10.1	11.6	0.0	1.0	4.0	16.0	36.0
## resid[3,8]	83.2	53.3	9.0	49.0	64.0	121.0	225.0
## resid[3,9]	92.3	60.6	16.0	49.0	81.0	121.0	225.0
## resid[3,10]	10.7	17.1	0.0	1.0	4.0	16.0	64.0
## resid[3,11]	210.1	115.4	49.0	121.0	196.0	289.0	484.0
## resid[3,12]	66.6	55.0	1.0	25.0	49.0	100.0	196.0
## resid[3,13]	52.3	68.6	0.0	9.0	25.0	81.0	256.0
## resid[3,14]	113.1	67.2	16.2	64.0	100.0	144.0	256.0
## resid[3,15]	57.4	42.4	4.0	25.0	49.0	81.0	169.0
## resid[3,16]	12.9	19.6	0.0	1.0	4.0	16.0	81.0
## resid[3,17]	702.2	136.1	441.0	625.0	676.0	784.0	961.0
## resid[3,18]	64.5	55.2	4.0	25.0	49.0	81.0	196.0
## resid[3,19]	12.7	16.5	0.0	1.0	9.0	16.0	63.6
## resid[3,20]	16.1	15.9	0.0	4.0	9.0	25.0	64.0
## resid[3,21]	146.5	82.8	25.0	81.0	144.0	196.0	361.0
## resid[3,22]	928.6	213.8	529.0	784.0	961.0	1089.0	1369.0
## resid[3,23]	7362.7	1109.7	5184.0	6561.0	7396.0	8100.0	9409.0
## resid[3,24]	196.6	120.6	25.2	100.0	169.0	256.0	484.0
## resid[3,25]	43.8	33.1	4.0	16.0	36.0	64.0	121.0
## resid[3,26]	129.5	76.0	25.0	81.0	121.0	169.0	324.0
## resid[3,27]	12.8	16.6	0.0	1.0	9.0	16.0	64.0
## resid[3,28]	302.7	140.7	64.0	196.0	289.0	400.0	576.0
## resid[3,29]	100.1	65.9	16.0	49.0	81.0	144.0	256.0
## resid[3,30]	70.0	48.1	9.0	36.0	64.0	100.0	196.0
## resid[3,31]	307.5	79.1	144.0	256.0	324.0	361.0	441.0
## resid[3,32]	6.2	9.2	0.0	1.0	4.0	9.0	25.0
## resid[3,33]	63.4	50.8	4.0	25.0	49.0	81.0	196.0
## resid[3,34]	44.0	32.4	4.0	25.0	36.0	64.0	121.0
## resid[3,35]	34.2	29.3	1.0	16.0	25.0	49.0	100.0
## resid[4,1]	35.8	28.4	4.0	16.0	25.0	49.0	100.0
## resid[4,2]	25.3	22.3	1.0	9.0	16.0	36.0	81.0
## resid[4,3]	25.3	20.9	1.0	9.0	25.0	36.0	81.0
## resid[4,4]	14.3	14.3	0.0	4.0	9.0	16.0	49.0
## resid[4,5]	12.2	13.7	0.0	4.0	9.0	16.0	49.0
## resid[4,6]	14.6	21.5	0.0	1.0	9.0	16.0	64.0
## resid[4,7]	24.7	25.3	0.0	4.0	16.0	36.0	81.0
## resid[4,8]	76.8	51.9	9.0	36.0	64.0	100.0	225.0
## resid[4,9]	20.6	26.1	0.0	4.0	9.0	25.0	81.0
## resid[4,10]	7.4	11.9	0.0	1.0	4.0	9.0	36.0
## resid[4,11]	236.7	113.7	49.0	144.0	225.0	324.0	484.0
## resid[4,12]	420.1	167.9	121.0	289.0	400.0	529.0	784.0
## resid[4,13]	118.1	91.1	4.1	49.0	100.0	169.0	324.0
## resid[4,14]	65.8	46.6	9.0	36.0	49.0	81.0	169.0
## resid[4,15]	53.1	36.9	9.0	25.0	49.0	76.4	144.0
## resid[4,16]	13.4	15.4	0.0	1.0	9.0	16.0	49.0
## resid[4,17]	271.8	95.0	100.0	196.0	256.0	324.0	441.0

## resid[4,18]	103.8	67.2	16.0	49.0	81.0	144.0	288.1
## resid[4,19]	16.9	17.4	0.0	4.0	12.5	25.0	49.0
## resid[4,20]	6.6	9.4	0.0	1.0	4.0	9.0	36.0
## resid[4,21]	51.4	56.6	0.0	9.0	36.0	81.0	196.0
## resid[4,22]	21862.5	1674.3	18500.0	20740.0	21900.0	23024.6	25272.0
## resid[4,23]	2783.6	839.5	1444.0	2209.0	2704.0	3249.0	4624.0
## resid[4,24]	56.8	41.2	4.0	25.0	49.0	81.0	169.0
## resid[4,25]	34.9	27.8	1.0	16.0	25.0	49.0	100.0
## resid[4,26]	1955.6	401.9	1225.0	1681.0	1936.0	2209.0	2809.0
## resid[4,27]	3428.2	729.3	2116.0	2916.0	3364.0	3844.0	4900.0
## resid[4,28]	15659.9	3090.5	10000.0	13460.0	15380.0	17690.0	21900.0
## resid[4,29]	144.8	87.0	25.0	81.0	121.0	196.0	361.0
## resid[4,30]	56.0	43.1	4.0	25.0	49.0	81.0	169.0
## resid[4,31]	26.1	24.1	0.0	9.0	16.0	36.0	81.0
## resid[4,32]	9.0	11.9	0.0	1.0	4.0	16.0	36.0
## resid[4,33]	56.9	42.0	4.0	25.0	49.0	81.0	169.0
## resid[4,34]	27.7	26.1	1.0	9.0	25.0	36.0	100.0
## resid[4,35]	29.4	24.2	1.0	9.0	25.0	36.0	99.5
## resid[5,1]	26.0	23.0	1.0	9.0	16.0	36.0	81.0
## resid[5,2]	26.3	23.0	1.0	9.0	25.0	36.0	81.0
## resid[5,3]	50.1	37.9	4.0	25.0	36.0	64.0	144.0
## resid[5,4]	20.6	19.9	1.0	9.0	16.0	25.0	64.0
## resid[5,5]	21.6	20.7	1.0	9.0	16.0	25.0	81.0
## resid[5,6]	8.1	11.7	0.0	1.0	4.0	9.0	36.0
## resid[5,7]	78.9	54.9	9.0	36.0	64.0	100.0	225.0
## resid[5,8]	227.6	120.9	64.0	144.0	196.0	289.0	529.0
## resid[5,9]	40.6	31.5	4.0	16.0	36.0	49.0	121.0
## resid[5,10]	1175.5	193.9	784.0	1024.0	1156.0	1296.0	1521.0
## resid[5,11]	78.4	61.2	4.0	36.0	64.0	100.0	225.0
## resid[5,12]	93.4	75.7	1.0	36.0	81.0	121.0	288.2
## resid[5,13]	292.8	151.5	81.0	169.0	256.0	361.0	625.0
## resid[5,14]	71.6	49.4	9.0	36.0	64.0	100.0	196.0
## resid[5,15]	64.9	44.4	9.0	36.0	49.0	81.0	169.0
## resid[5,16]	219.7	79.0	64.0	169.0	225.0	289.0	361.0
## resid[5,17]	61.4	45.4	4.0	25.0	49.0	81.0	169.0
## resid[5,18]	96.0	62.8	16.0	49.0	81.0	121.0	256.0
## resid[5,19]	16.9	17.0	1.0	4.0	9.0	25.0	64.0
## resid[5,20]	1552.0	149.9	1296.0	1444.0	1600.0	1681.0	1849.0
## resid[5,21]	129.0	76.7	25.0	81.0	121.0	169.0	324.0
## resid[5,22]	7297.7	713.6	5929.0	6724.0	7225.0	7744.0	8649.0
## resid[5,23]	598.4	255.9	196.0	400.0	576.0	784.0	1156.0
## resid[5,24]	41.2	33.0	4.0	16.0	36.0	49.0	121.0
## resid[5,25]	34.4	28.6	1.0	16.0	25.0	49.0	100.0
## resid[5,26]	78.7	55.1	1.0	36.0	64.0	121.0	225.0
## resid[5,27]	233.7	131.6	36.0	144.0	225.0	324.0	529.0
## resid[5,28]	1087.6	410.7	441.0	797.9	1024.0	1296.0	2116.0
## resid[5,29]	117.6	77.5	16.0	64.0	100.0	169.0	324.0
## resid[5,30]	57.9	46.3	4.0	25.0	49.0	81.0	169.0
## resid[5,31]	17.4	16.0	0.0	4.0	16.0	25.0	64.0
## resid[5,32]	32.3	32.2	1.0	9.0	25.0	49.0	120.5
## resid[5,33]	1059.0	205.9	625.0	961.0	1089.0	1225.0	1444.0
## resid[5,34]	31.8	26.3	1.0	16.0	25.0	49.0	100.0
## resid[5,35]	34.0	28.2	1.0	16.0	25.0	49.0	121.0
## resid[6,1]	47.3	36.9	4.0	25.0	36.0	64.0	144.0

## resid[6,2]	46.1	34.5	4.0	25.0	36.0	64.0	144.0
## resid[6,3]	27.9	25.6	1.0	9.0	25.0	36.0	100.0
## resid[6,4]	24.3	21.3	1.0	9.0	16.0	36.0	81.0
## resid[6,5]	49.9	38.6	4.0	25.0	36.0	64.0	144.0
## resid[6,6]	12.1	19.3	0.0	1.0	4.0	16.0	64.0
## resid[6,7]	108.1	67.3	16.0	64.0	100.0	144.0	256.0
## resid[6,8]	184.9	78.2	49.0	121.0	169.0	225.0	361.0
## resid[6,9]	27.5	29.3	0.0	9.0	16.0	36.0	120.5
## resid[6,10]	1513.9	278.2	961.0	1296.0	1521.0	1681.0	2025.0
## resid[6,11]	105.8	90.1	1.0	36.0	81.0	144.0	360.1
## resid[6,12]	282.1	153.4	64.0	169.0	256.0	361.0	625.0
## resid[6,13]	201.5	107.8	49.0	121.0	196.0	256.0	482.9
## resid[6,14]	117.2	72.3	25.0	64.0	100.0	144.0	323.1
## resid[6,15]	1177.5	308.8	576.0	961.0	1156.0	1369.0	1849.0
## resid[6,16]	13.4	16.3	0.0	1.0	9.0	16.0	64.0
## resid[6,17]	98.4	59.0	16.0	49.0	81.0	121.0	256.0
## resid[6,18]	480.8	132.7	256.0	400.0	484.0	576.0	729.0
## resid[6,19]	26.8	23.5	1.0	9.0	25.0	36.0	100.0
## resid[6,20]	1022.8	300.9	441.0	841.0	1024.0	1225.0	1600.0
## resid[6,21]	2576.6	885.4	1024.0	1936.0	2500.0	3220.4	4356.0
## resid[6,22]	52.5	63.0	0.0	9.0	25.0	81.0	225.0
## resid[6,23]	178.6	92.6	36.0	100.0	169.0	225.0	400.0
## resid[6,24]	64.4	45.2	9.0	36.0	49.0	81.0	169.0
## resid[6,25]	542.8	331.5	100.0	289.0	484.0	729.0	1369.0
## resid[6,26]	101.7	135.8	0.0	9.0	49.0	144.0	484.0
## resid[6,27]	450.1	271.5	81.0	256.0	400.0	625.0	1089.0
## resid[6,28]	151.2	88.8	25.0	81.0	144.0	196.0	361.0
## resid[6,29]	97.6	61.4	16.0	49.0	81.0	121.0	256.0
## resid[6,30]	32.0	43.3	0.0	4.0	16.0	49.0	144.0
## resid[6,31]	20.1	23.7	0.0	4.0	16.0	25.0	81.0
## resid[6,32]	74.2	52.6	9.0	36.0	64.0	100.0	196.0
## resid[6,33]	63.0	45.9	9.0	36.0	49.0	81.0	169.0
## resid[6,34]	53.4	40.3	4.0	25.0	49.0	64.0	144.0
## resid[6,35]	379.8	266.0	16.0	169.0	324.0	529.0	961.0
## resid[7,1]	55.4	40.0	9.0	25.0	49.0	81.0	144.0
## resid[7,2]	66.3	46.9	9.0	36.0	49.0	100.0	169.0
## resid[7,3]	27.2	26.9	1.0	9.0	16.0	36.0	100.0
## resid[7,4]	28.9	24.3	1.0	9.0	25.0	36.0	100.0
## resid[7,5]	55.1	38.0	4.0	25.0	49.0	81.0	144.0
## resid[7,6]	131.8	82.5	25.0	64.0	121.0	169.0	324.0
## resid[7,7]	204.3	107.4	49.0	121.0	196.0	256.0	484.0
## resid[7,8]	87.2	61.4	9.0	49.0	81.0	121.0	225.0
## resid[7,9]	8.2	13.4	0.0	1.0	4.0	9.0	49.0
## resid[7,10]	6774.5	514.6	5776.0	6400.0	6724.0	7225.0	7744.0
## resid[7,11]	493.3	232.6	121.0	324.0	441.0	625.0	1024.0
## resid[7,12]	578.5	254.5	196.7	400.0	529.0	729.0	1156.0
## resid[7,13]	351.9	170.9	100.0	225.0	324.0	441.0	729.0
## resid[7,14]	177.3	101.1	36.0	100.0	169.0	225.0	441.0
## resid[7,15]	1354.5	320.4	729.0	1156.0	1369.0	1521.0	1936.0
## resid[7,16]	60.9	58.1	1.0	16.0	49.0	81.0	225.0
## resid[7,17]	200.9	112.1	49.0	121.0	169.0	256.0	482.9
## resid[7,18]	108.9	67.7	16.0	64.0	100.0	144.0	289.0
## resid[7,19]	17.1	18.5	0.0	4.0	9.0	25.0	64.0
## resid[7,20]	30.6	23.9	0.0	9.0	25.0	49.0	81.0

## resid[7,21]	11265.8	1527.2	8281.0	10200.0	11240.0	12320.0	14160.0
## resid[7,22]	1119.3	394.4	484.0	841.0	1089.0	1369.0	2025.0
## resid[7,23]	687.6	284.8	256.0	484.0	625.0	841.0	1296.0
## resid[7,24]	97.4	60.9	16.0	49.0	81.0	121.0	256.0
## resid[7,25]	15.8	22.4	0.0	1.0	9.0	25.0	81.0
## resid[7,26]	2124.8	866.3	784.0	1521.0	2025.0	2601.0	4096.0
## resid[7,27]	2403.1	721.1	1296.0	1936.0	2304.0	2809.0	4096.0
## resid[7,28]	441.5	203.3	144.0	289.0	400.0	576.0	900.0
## resid[7,29]	112.7	71.8	16.0	64.0	100.0	144.0	289.0
## resid[7,30]	1430.9	305.7	841.0	1225.0	1444.0	1600.0	2025.0
## resid[7,31]	111.9	70.3	16.0	64.0	100.0	144.0	289.0
## resid[7,32]	46.1	48.0	0.0	9.0	36.0	64.0	169.0
## resid[7,33]	61.3	54.6	1.0	16.0	49.0	81.0	196.0
## resid[7,34]	72.1	48.7	9.0	36.0	64.0	100.0	196.0
## resid[7,35]	2022.2	251.1	1521.0	1849.0	2025.0	2209.0	2500.0
## resid[8,1]	75.4	52.4	9.0	36.0	64.0	100.0	196.0
## resid[8,2]	24.3	24.2	1.0	9.0	16.0	36.0	81.0
## resid[8,3]	21.7	19.3	1.0	9.0	16.0	36.0	64.0
## resid[8,4]	22.4	20.6	1.0	9.0	16.0	36.0	81.0
## resid[8,5]	49.7	36.9	4.0	25.0	36.0	64.0	144.0
## resid[8,6]	230.4	118.6	49.0	144.0	225.0	289.0	484.0
## resid[8,7]	14.4	22.3	0.0	1.0	9.0	16.0	81.0
## resid[8,8]	39.2	27.7	1.0	16.0	36.0	49.0	100.0
## resid[8,9]	99.4	53.9	9.0	64.0	100.0	144.0	224.3
## resid[8,10]	845.9	185.5	484.0	729.0	841.0	961.0	1225.0
## resid[8,11]	452.8	262.9	100.0	256.0	400.0	576.0	1156.0
## resid[8,12]	107.2	82.3	4.0	49.0	81.0	144.0	324.0
## resid[8,13]	110.8	70.4	16.0	64.0	100.0	144.0	289.0
## resid[8,14]	127.0	76.1	25.0	64.0	121.0	169.0	289.0
## resid[8,15]	79.3	66.0	1.0	25.0	64.0	121.0	225.0
## resid[8,16]	285.9	142.1	81.0	175.4	256.0	361.0	625.0
## resid[8,17]	666.3	163.1	361.0	529.0	676.0	784.0	961.0
## resid[8,18]	22.0	21.4	1.0	9.0	16.0	25.0	81.0
## resid[8,19]	5.1	6.8	0.0	1.0	4.0	9.0	25.0
## resid[8,20]	20.8	22.2	0.0	4.0	16.0	25.0	81.0
## resid[8,21]	903.4	435.9	256.0	576.0	841.0	1156.0	1936.0
## resid[8,22]	73.0	57.2	4.0	36.0	64.0	100.0	225.0
## resid[8,23]	70.9	47.3	9.0	36.0	64.0	100.0	196.0
## resid[8,24]	11.1	16.6	0.0	1.0	4.0	16.0	64.0
## resid[8,25]	1082.9	228.3	676.0	900.0	1089.0	1225.0	1521.0
## resid[8,26]	2706.0	989.7	961.0	2025.0	2601.0	3249.0	5041.0
## resid[8,27]	20.7	33.4	0.0	1.0	9.0	25.0	121.0
## resid[8,28]	101.1	66.6	16.0	49.0	81.0	144.0	256.0
## resid[8,29]	67.8	53.7	4.0	25.0	49.0	100.0	196.0
## resid[8,30]	159.0	219.6	0.0	16.0	81.0	225.0	839.6
## resid[8,31]	6044.3	1509.7	3481.0	5041.0	5929.0	7056.0	9404.1
## resid[8,32]	21290.7	812.4	19600.0	20740.0	21320.0	21900.0	22800.0
## resid[8,33]	52.1	38.5	4.0	25.0	49.0	64.0	144.0
## resid[8,34]	80.0	45.8	9.0	49.0	81.0	121.0	169.0
## resid[8,35]	3600.6	387.8	2809.0	3364.0	3600.0	3844.0	4356.0
## resid[9,1]	18.2	18.5	1.0	4.0	16.0	25.0	64.0
## resid[9,2]	10.6	13.3	0.0	1.0	4.0	16.0	49.0
## resid[9,3]	13.2	13.6	0.0	4.0	9.0	16.0	49.0
## resid[9,4]	7.6	10.8	0.0	1.0	4.0	9.0	36.0

## resid[9,5]	541.2	232.5	196.0	361.0	484.0	676.0	1089.0
## resid[9,6]	64.0	47.9	9.0	27.4	49.0	81.0	195.3
## resid[9,7]	77.2	55.0	9.0	36.0	64.0	100.0	224.2
## resid[9,8]	121.5	53.1	25.0	81.0	121.0	144.0	225.0
## resid[9,9]	238.8	109.5	49.0	169.0	225.0	324.0	484.0
## resid[9,10]	1262.1	200.0	900.0	1156.0	1225.0	1369.0	1681.0
## resid[9,11]	156.7	94.1	25.0	81.0	144.0	196.0	400.0
## resid[9,12]	473.5	212.6	144.0	324.0	441.0	576.0	961.0
## resid[9,13]	38.8	37.3	1.0	9.0	25.0	49.0	143.4
## resid[9,14]	669.9	167.9	361.0	576.0	676.0	784.0	961.0
## resid[9,15]	267.7	133.3	49.0	169.0	256.0	361.0	529.0
## resid[9,16]	58.4	42.5	4.0	25.0	49.0	81.0	169.0
## resid[9,17]	62.2	52.4	1.0	25.0	49.0	81.0	196.0
## resid[9,18]	11.8	16.3	0.0	1.0	4.0	16.0	63.6
## resid[9,19]	117.6	43.3	36.0	81.0	121.0	144.0	196.0
## resid[9,20]	30.3	23.3	0.0	9.0	25.0	49.0	81.0
## resid[9,21]	1133.1	386.7	484.0	841.0	1089.0	1369.0	2025.0
## resid[9,22]	470.0	210.1	144.0	324.0	441.0	576.0	961.0
## resid[9,23]	9.0	13.4	0.0	1.0	4.0	9.0	49.0
## resid[9,24]	316.5	111.7	121.0	225.0	324.0	400.0	529.0
## resid[9,25]	2309.3	364.8	1600.0	2025.0	2304.0	2500.0	3025.0
## resid[9,26]	12975.2	2872.5	8281.0	11030.0	12770.0	14640.0	19040.0
## resid[9,27]	393.6	194.2	100.5	256.0	361.0	484.0	841.0
## resid[9,28]	97.0	62.7	16.0	49.0	81.0	121.0	256.0
## resid[9,29]	2479.7	377.4	1764.0	2209.0	2500.0	2704.0	3249.0
## resid[9,30]	665.5	173.2	324.0	529.0	676.0	784.0	1024.0
## resid[9,31]	72.6	52.1	9.0	36.0	64.0	100.0	196.0
## resid[9,32]	12.1	19.1	0.0	1.0	4.0	16.0	64.0
## resid[9,33]	50.1	37.6	4.0	25.0	36.0	64.0	144.0
## resid[9,34]	452.9	104.9	256.0	361.0	441.0	529.0	676.0
## resid[9,35]	1363.6	232.9	900.0	1225.0	1369.0	1521.0	1764.0
## resid[10,1]	12.6	14.9	0.0	4.0	9.0	16.0	49.0
## resid[10,2]	5.8	8.9	0.0	1.0	4.0	9.0	25.0
## resid[10,3]	16.9	17.1	0.0	4.0	16.0	25.0	64.0
## resid[10,4]	17.1	17.3	1.0	4.0	9.0	25.0	64.0
## resid[10,5]	35.9	28.9	1.0	16.0	25.0	49.0	121.0
## resid[10,6]	30.6	28.0	1.0	9.0	25.0	49.0	100.0
## resid[10,7]	28.9	25.4	1.0	9.0	25.0	36.0	99.5
## resid[10,8]	540.5	110.0	324.0	484.0	529.0	625.0	729.0
## resid[10,9]	8.7	11.6	0.0	1.0	4.0	9.0	36.0
## resid[10,10]	72.0	49.7	9.0	36.0	64.0	100.0	196.0
## resid[10,11]	79.5	55.7	9.0	36.0	64.0	100.0	225.0
## resid[10,12]	63.4	45.0	9.0	25.0	49.0	81.0	169.0
## resid[10,13]	114.4	71.4	16.2	64.0	100.0	144.0	256.0
## resid[10,14]	438.8	137.9	169.0	324.0	441.0	529.0	729.0
## resid[10,15]	178.7	128.2	16.0	81.0	144.0	256.0	484.0
## resid[10,16]	41.7	34.1	4.0	16.0	36.0	64.0	121.0
## resid[10,17]	20.6	15.5	0.0	9.0	16.0	36.0	49.0
## resid[10,18]	189.2	62.2	64.0	144.0	196.0	225.0	324.0
## resid[10,19]	13.8	15.6	0.0	4.0	9.0	16.0	63.6
## resid[10,20]	55.0	42.7	4.0	25.0	49.0	81.0	169.0
## resid[10,21]	128.5	77.4	25.0	81.0	121.0	169.0	324.0
## resid[10,22]	49.9	39.0	4.0	25.0	36.0	64.0	144.0
## resid[10,23]	43.9	35.7	1.0	16.0	36.0	64.0	144.0

## resid[10,24]	15.9	24.9	0.0	1.0	9.0	25.0	81.0
## resid[10,25]	1379.8	311.3	784.0	1156.0	1369.0	1600.0	2025.0
## resid[10,26]	205.3	114.2	49.0	121.0	196.0	256.0	484.0
## resid[10,27]	46.5	35.7	4.0	16.0	36.0	64.0	144.0
## resid[10,28]	72.0	52.9	4.0	36.0	64.0	100.0	196.0
## resid[10,29]	5177.7	437.8	4228.2	4900.0	5184.0	5476.0	5929.0
## resid[10,30]	99.5	66.3	16.0	49.0	81.0	138.2	256.0
## resid[10,31]	309.2	162.3	64.0	196.0	289.0	400.0	676.0
## resid[10,32]	11.3	14.0	0.0	1.0	9.0	16.0	49.0
## resid[10,33]	8.2	12.1	0.0	1.0	4.0	9.0	36.0
## resid[10,34]	56.0	42.6	4.0	25.0	49.0	81.0	169.0
## resid[10,35]	37.2	37.9	1.0	9.0	25.0	49.0	144.0
## resid[11,1]	25.7	27.7	0.0	4.0	16.0	36.0	100.0
## resid[11,2]	35.5	32.6	1.0	16.0	25.0	49.0	121.0
## resid[11,3]	200.9	110.8	49.0	121.0	169.0	256.0	482.9
## resid[11,4]	198.9	110.2	49.0	121.0	196.0	256.0	441.0
## resid[11,5]	742.7	291.0	289.0	529.0	729.0	900.0	1369.0
## resid[11,6]	234.1	119.1	64.0	144.0	225.0	289.0	527.8
## resid[11,7]	173.1	92.7	36.0	100.0	169.0	225.0	400.0
## resid[11,8]	12158.1	1534.7	9216.0	11240.0	12100.0	13230.0	15130.0
## resid[11,9]	4204.9	555.3	3136.0	3844.0	4225.0	4624.0	5329.0
## resid[11,10]	566.2	242.2	196.0	400.0	529.0	729.0	1156.0
## resid[11,11]	505.7	210.6	169.0	361.0	484.0	625.0	961.0
## resid[11,12]	461.8	207.4	144.0	324.0	441.0	576.0	961.0
## resid[11,13]	63.9	76.3	0.0	9.0	36.0	100.0	288.2
## resid[11,14]	56.1	72.0	0.0	9.0	25.0	81.0	256.0
## resid[11,15]	1736.1	569.1	841.0	1369.0	1681.0	2116.0	2916.0
## resid[11,16]	155.3	92.0	36.0	100.0	144.0	196.0	399.0
## resid[11,17]	138.0	84.1	25.0	81.0	121.0	169.0	361.0
## resid[11,18]	21.2	29.0	0.0	1.0	9.0	25.0	100.0
## resid[11,19]	142.6	89.7	25.0	81.0	121.0	196.0	361.0
## resid[11,20]	238.3	134.0	49.0	144.0	225.0	324.0	576.0
## resid[11,21]	343.4	156.3	100.0	225.0	324.0	441.0	727.6
## resid[11,22]	318.8	152.2	100.0	225.0	289.0	400.0	676.0
## resid[11,23]	189.4	167.3	1.0	64.0	144.0	256.0	625.0
## resid[11,24]	1282.3	458.9	484.0	961.0	1296.0	1600.0	2304.0
## resid[11,25]	1043.5	370.2	441.0	784.0	1024.0	1225.0	1849.0
## resid[11,26]	319.2	152.2	81.4	196.0	289.0	400.0	676.0
## resid[11,27]	297.6	147.0	81.0	196.0	289.0	389.9	625.0
## resid[11,28]	12993.3	2033.2	9029.7	11660.0	13000.0	14400.0	16900.0
## resid[11,29]	4666.7	998.5	2809.0	3969.0	4624.0	5329.0	6719.9
## resid[11,30]	1041.7	368.1	441.0	784.0	1024.0	1296.0	1849.0
## resid[11,31]	251.2	127.5	64.0	169.0	225.0	324.0	574.8
## resid[11,32]	88.8	59.6	16.0	49.0	81.0	121.0	225.0
## resid[11,33]	30.0	45.9	0.0	4.0	16.0	36.0	144.0
## resid[11,34]	2005.3	441.0	1156.0	1681.0	1980.0	2304.0	2916.0
## resid[11,35]	444.1	200.7	144.0	289.0	400.0	529.0	900.0
## resid[12,1]	5.9	10.5	0.0	1.0	1.0	9.0	36.0
## resid[12,2]	15.6	15.3	0.0	4.0	9.0	25.0	49.0
## resid[12,3]	36.2	29.3	1.0	16.0	25.0	49.0	121.0
## resid[12,4]	47.7	38.2	4.0	25.0	36.0	64.0	144.0
## resid[12,5]	87.5	58.3	16.0	49.0	81.0	121.0	225.0
## resid[12,6]	58.7	44.2	9.0	25.0	49.0	81.0	169.0
## resid[12,7]	9.9	15.7	0.0	1.0	4.0	9.0	49.0



## resid[12,8]	3032.1	350.0	2304.0	2809.0	3025.0	3249.0	3721.0
## resid[12,9]	93.3	59.5	16.0	49.0	81.0	121.0	225.0
## resid[12,10]	206.7	116.2	49.0	121.0	196.0	280.4	441.0
## resid[12,11]	134.3	83.0	25.0	81.0	121.0	169.0	361.0
## resid[12,12]	106.7	65.5	16.0	64.0	100.0	144.0	256.0
## resid[12,13]	64.1	56.6	1.0	25.0	49.0	81.0	196.0
## resid[12,14]	218.3	115.2	49.0	144.0	196.0	289.0	484.0
## resid[12,15]	319.2	159.9	100.0	196.0	289.0	400.0	676.0
## resid[12,16]	40.7	33.6	4.0	16.0	36.0	49.0	121.0
## resid[12,17]	1529.9	193.4	1156.0	1444.0	1521.0	1681.0	1849.0
## resid[12,18]	31.7	24.3	1.0	16.0	25.0	49.0	81.0
## resid[12,19]	58.3	40.3	9.0	25.0	49.0	81.0	169.0
## resid[12,20]	82.3	55.7	9.0	36.0	64.0	121.0	225.0
## resid[12,21]	76.3	51.9	9.0	36.0	64.0	100.0	196.0
## resid[12,22]	81.2	56.0	9.0	36.0	64.0	100.0	225.0
## resid[12,23]	23.4	26.6	0.0	4.0	16.0	36.0	100.0
## resid[12,24]	598.4	189.6	256.0	484.0	576.0	729.0	961.0
## resid[12,25]	222.1	119.8	49.0	144.0	196.0	289.0	529.0
## resid[12,26]	68.6	49.8	9.0	36.0	49.0	100.0	196.0
## resid[12,27]	72.7	51.7	9.0	36.0	64.0	100.0	196.0
## resid[12,28]	969.1	230.0	529.0	841.0	961.0	1089.0	1444.0
## resid[12,29]	14.2	22.1	0.0	1.0	4.0	16.0	80.6
## resid[12,30]	596.7	256.3	225.0	400.0	576.0	729.0	1156.0
## resid[12,31]	27.1	22.9	1.0	9.0	25.0	36.0	81.0
## resid[12,32]	4362.9	298.8	3721.0	4225.0	4356.0	4624.0	4900.0
## resid[12,33]	15.1	18.2	0.0	4.0	9.0	25.0	64.0
## resid[12,34]	12.2	17.3	0.0	1.0	4.0	16.0	63.6
## resid[12,35]	142.5	82.2	25.0	81.0	121.0	196.0	324.0
## resid[13,1]	22.5	25.2	0.0	4.0	16.0	36.0	100.0
## resid[13,2]	56.7	39.9	4.0	25.0	49.0	81.0	168.4
## resid[13,3]	104.3	66.8	16.0	49.0	100.0	144.0	256.0
## resid[13,4]	140.4	80.7	25.0	81.0	121.0	196.0	324.0
## resid[13,5]	175.8	95.6	36.0	100.0	169.0	225.0	400.0
## resid[13,6]	149.3	90.5	36.0	81.0	121.0	196.0	361.0
## resid[13,7]	242.1	123.6	64.0	144.0	225.0	324.0	529.0
## resid[13,8]	1722.1	332.5	1089.0	1521.0	1764.0	1936.0	2304.0
## resid[13,9]	297.3	145.2	81.0	196.0	289.0	361.0	625.0
## resid[13,10]	111.9	85.2	1.0	49.0	100.0	169.0	324.0
## resid[13,11]	439.1	193.0	121.0	289.0	400.0	576.0	841.0
## resid[13,12]	375.7	242.2	36.0	196.0	324.0	529.0	1022.4
## resid[13,13]	88.1	102.2	0.0	16.0	49.0	121.0	361.0
## resid[13,14]	753.0	287.9	289.0	529.0	729.0	900.0	1442.1
## resid[13,15]	469.5	201.1	169.0	324.0	441.0	576.0	900.0
## resid[13,16]	135.7	81.4	25.0	81.0	121.0	169.0	324.0
## resid[13,17]	19542.5	1297.7	16900.0	18770.0	19600.0	20450.0	22200.0
## resid[13,18]	16.7	25.4	0.0	1.0	9.0	16.0	100.0
## resid[13,19]	175.2	101.3	36.0	100.0	169.0	225.0	400.0
## resid[13,20]	110.5	70.8	16.0	64.0	100.0	144.0	289.0
## resid[13,21]	237.7	123.7	64.0	144.0	225.0	289.0	529.0
## resid[13,22]	487.6	417.9	9.0	169.0	400.0	676.0	1521.0
## resid[13,23]	916.6	426.2	225.0	576.0	900.0	1225.0	1849.0
## resid[13,24]	394.1	228.1	64.0	225.0	361.0	529.0	900.0
## resid[13,25]	385.4	177.0	121.0	256.0	361.0	484.0	784.0
## resid[13,26]	220.7	113.3	49.0	144.0	196.0	289.0	484.0

## resid[13,27]	2542.1	932.6	1089.0	1936.0	2401.0	3025.0	4624.0
## resid[13,28]	24175.9	1923.6	20450.0	22800.0	24030.0	25600.0	27890.0
## resid[13,29]	586.4	244.4	225.0	400.0	529.0	729.0	1156.0
## resid[13,30]	812.4	311.8	289.0	587.9	784.0	1024.0	1521.0
## resid[13,31]	89.4	58.5	9.0	49.0	81.0	121.0	225.0
## resid[13,32]	2416.5	560.8	1370.8	2025.0	2401.0	2809.0	3600.0
## resid[13,33]	1386.6	420.2	625.0	1089.0	1369.0	1681.0	2304.0
## resid[13,34]	713.6	282.2	289.0	529.0	676.0	884.9	1369.0
## resid[13,35]	216.3	114.4	49.0	144.0	196.0	289.0	484.0
## resid[14,1]	3.0	4.4	0.0	1.0	1.0	4.0	16.0
## resid[14,2]	23.5	21.3	1.0	9.0	16.0	36.0	81.0
## resid[14,3]	34.4	28.4	1.0	16.0	25.0	49.0	100.0
## resid[14,4]	65.6	47.0	9.0	36.0	49.0	81.0	196.0
## resid[14,5]	37.2	30.2	4.0	16.0	36.0	49.0	121.0
## resid[14,6]	37.9	31.6	1.0	16.0	25.0	49.0	121.0
## resid[14,7]	12.9	13.8	0.0	4.0	9.0	16.0	49.0
## resid[14,8]	79.4	52.1	9.0	36.0	64.0	100.0	225.0
## resid[14,9]	101.3	69.7	9.0	49.0	81.0	144.0	289.0
## resid[14,10]	8.3	11.9	0.0	1.0	4.0	9.0	36.0
## resid[14,11]	127.2	78.3	25.0	64.0	110.0	169.0	324.0
## resid[14,12]	218.1	115.1	49.0	144.0	196.0	289.0	484.0
## resid[14,13]	348.9	166.3	49.0	225.0	324.0	441.0	729.0
## resid[14,14]	292.0	148.0	81.0	169.0	256.0	361.0	625.0
## resid[14,15]	163.1	93.8	36.0	100.0	144.0	225.0	400.0
## resid[14,16]	49.0	36.9	4.0	25.0	36.0	64.0	144.0
## resid[14,17]	2054.9	237.9	1600.0	1936.0	2025.0	2209.0	2500.0
## resid[14,18]	22.8	24.7	0.0	4.0	16.0	36.0	100.0
## resid[14,19]	46.8	44.3	1.0	16.0	36.0	64.0	169.0
## resid[14,20]	46.9	36.1	4.0	16.0	36.0	64.0	144.0
## resid[14,21]	82.4	55.8	9.0	49.0	64.0	121.0	225.0
## resid[14,22]	595.2	139.1	324.0	484.0	576.0	676.0	841.0
## resid[14,23]	5268.3	609.6	4096.0	4900.0	5329.0	5776.0	6400.0
## resid[14,24]	480.8	212.3	169.0	324.0	441.0	625.0	961.0
## resid[14,25]	272.9	147.9	64.0	169.0	256.0	361.0	625.0
## resid[14,26]	56.5	44.9	4.0	25.0	49.0	81.0	169.0
## resid[14,27]	55.4	38.9	1.0	25.0	49.0	81.0	144.0
## resid[14,28]	1881.1	313.5	1225.0	1681.0	1849.0	2116.0	2401.0
## resid[14,29]	366.8	171.9	121.0	256.0	324.0	484.0	784.0
## resid[14,30]	272.0	139.5	64.0	169.0	256.0	361.0	576.0
## resid[14,31]	24.1	23.7	0.0	4.0	16.0	36.0	81.0
## resid[14,32]	221.4	64.3	100.0	169.0	225.0	256.0	361.0
## resid[14,33]	126.7	67.7	16.0	81.0	121.0	169.0	256.0
## resid[14,34]	445.9	207.5	144.0	289.0	400.0	576.0	900.0
## resid[14,35]	153.5	91.8	25.0	81.0	144.0	196.0	400.0
## Captures.pred[1,1]	5.0	2.2	1.0	3.2	5.0	6.0	10.0
## Captures.pred[1,2]	5.1	2.2	1.0	3.0	5.0	7.0	10.0
## Captures.pred[1,3]	8.1	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[1,4]	11.8	3.5	6.0	9.0	11.0	14.0	19.0
## Captures.pred[1,5]	9.0	3.0	4.0	7.0	9.0	11.0	16.0
## Captures.pred[1,6]	9.5	3.1	4.0	7.0	9.0	12.0	16.0
## Captures.pred[1,7]	10.4	3.3	5.0	8.0	10.0	12.0	17.0
## Captures.pred[1,8]	12.7	3.6	6.0	10.0	13.0	15.0	20.0
## Captures.pred[1,9]	24.5	5.1	15.0	21.0	24.0	28.0	35.0
## Captures.pred[1,10]	13.4	3.5	7.0	11.0	13.0	16.0	21.0

## Captures.pred[1,11]	15.8	4.2	8.0	13.0	16.0	18.0	24.0
## Captures.pred[1,12]	18.0	4.4	10.0	15.0	18.0	21.0	28.0
## Captures.pred[1,13]	31.6	5.7	20.0	28.0	31.0	35.0	43.0
## Captures.pred[1,14]	54.5	7.5	41.0	49.0	55.0	60.0	69.0
## Captures.pred[1,15]	22.4	4.8	14.0	19.0	22.0	26.0	32.0
## Captures.pred[1,16]	8.8	3.1	3.0	7.0	9.0	11.0	16.0
## Captures.pred[1,17]	8.7	3.0	3.0	7.0	9.0	11.0	15.0
## Captures.pred[1,18]	12.6	3.7	6.0	10.0	12.0	15.0	20.0
## Captures.pred[1,19]	11.6	3.5	5.0	9.0	11.0	14.0	19.0
## Captures.pred[1,20]	8.5	3.0	3.0	6.0	8.0	11.0	15.0
## Captures.pred[1,21]	12.9	3.7	7.0	10.0	13.0	15.0	20.0
## Captures.pred[1,22]	14.5	3.9	8.0	12.0	14.0	17.0	23.0
## Captures.pred[1,23]	19.1	4.4	11.0	16.0	19.0	22.0	28.0
## Captures.pred[1,24]	42.6	6.7	30.0	38.0	43.0	47.0	56.0
## Captures.pred[1,25]	17.2	4.2	10.0	14.0	17.0	20.0	25.0
## Captures.pred[1,26]	12.9	3.6	6.0	10.0	13.0	15.0	20.0
## Captures.pred[1,27]	13.7	3.9	7.0	11.0	13.0	16.0	22.0
## Captures.pred[1,28]	22.1	4.7	13.0	19.0	22.0	25.0	32.0
## Captures.pred[1,29]	29.2	5.6	18.0	25.0	29.0	33.0	41.0
## Captures.pred[1,30]	17.9	4.3	10.0	15.0	18.0	21.0	27.0
## Captures.pred[1,31]	6.7	2.6	2.0	5.0	6.0	8.0	13.0
## Captures.pred[1,32]	11.9	3.5	5.0	9.0	12.0	14.0	19.0
## Captures.pred[1,33]	11.6	3.5	5.0	9.0	11.0	14.0	18.0
## Captures.pred[1,34]	16.6	4.2	9.0	14.0	16.0	19.0	26.0
## Captures.pred[1,35]	11.6	3.4	5.0	9.0	11.0	14.0	19.0
## Captures.pred[2,1]	3.2	1.8	0.0	2.0	3.0	4.0	7.0
## Captures.pred[2,2]	3.9	2.0	1.0	3.0	4.0	5.0	8.0
## Captures.pred[2,3]	6.1	2.4	2.0	4.0	6.0	8.0	11.0
## Captures.pred[2,4]	5.2	2.3	1.0	4.0	5.0	7.0	10.0
## Captures.pred[2,5]	5.2	2.3	1.0	4.0	5.0	7.0	10.0
## Captures.pred[2,6]	6.7	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[2,7]	7.2	2.8	2.0	5.0	7.0	9.0	13.0
## Captures.pred[2,8]	9.3	3.0	4.0	7.0	9.0	11.0	16.0
## Captures.pred[2,9]	11.2	3.4	5.0	9.0	11.0	13.0	18.0
## Captures.pred[2,10]	11.4	3.3	5.0	9.0	11.0	13.0	18.0
## Captures.pred[2,11]	11.2	3.4	5.0	9.0	11.0	13.0	19.0
## Captures.pred[2,12]	15.0	3.9	8.0	12.0	15.0	17.0	23.0
## Captures.pred[2,13]	51.8	7.3	37.0	47.0	52.0	57.0	67.0
## Captures.pred[2,14]	12.1	3.6	6.0	10.0	12.0	15.0	19.0
## Captures.pred[2,15]	22.9	4.8	14.0	20.0	22.0	26.0	33.0
## Captures.pred[2,16]	5.9	2.4	2.0	4.0	6.0	8.0	11.0
## Captures.pred[2,17]	9.9	3.1	4.0	8.0	10.0	12.0	16.0
## Captures.pred[2,18]	11.4	3.5	5.0	9.0	11.0	14.0	18.0
## Captures.pred[2,19]	6.3	2.6	2.0	4.2	6.0	8.0	12.0
## Captures.pred[2,20]	8.6	3.0	3.0	6.0	8.5	10.0	15.0
## Captures.pred[2,21]	10.4	3.4	4.0	8.0	10.0	13.0	18.0
## Captures.pred[2,22]	13.1	3.5	7.0	11.0	13.0	15.0	21.0
## Captures.pred[2,23]	60.5	8.3	45.0	55.0	60.0	66.0	76.0
## Captures.pred[2,24]	20.3	4.7	12.0	17.0	20.0	23.0	29.0
## Captures.pred[2,25]	23.0	4.9	15.0	19.0	23.0	26.0	33.0
## Captures.pred[2,26]	9.7	3.1	4.0	8.0	10.0	12.0	16.0
## Captures.pred[2,27]	15.1	3.9	8.0	12.0	15.0	18.0	23.0
## Captures.pred[2,28]	22.5	4.8	13.0	19.0	22.0	26.0	32.0
## Captures.pred[2,29]	10.5	3.2	5.0	8.0	10.0	12.0	17.0

## Captures.pred[2,30]	18.3	4.4	10.0	15.0	18.0	21.0	27.0
## Captures.pred[2,31]	5.6	2.4	1.0	4.0	6.0	7.0	11.0
## Captures.pred[2,32]	11.5	3.6	5.0	9.0	11.0	14.0	19.0
## Captures.pred[2,33]	9.0	3.0	4.0	7.0	9.0	11.0	16.0
## Captures.pred[2,34]	7.6	2.8	3.0	6.0	7.0	9.0	14.0
## Captures.pred[2,35]	21.5	4.7	13.0	18.0	21.0	25.0	31.0
## Captures.pred[3,1]	3.6	1.9	1.0	2.0	3.0	5.0	7.0
## Captures.pred[3,2]	3.5	1.9	0.0	2.0	3.0	5.0	8.0
## Captures.pred[3,3]	4.6	2.2	1.0	3.0	4.0	6.0	10.0
## Captures.pred[3,4]	4.4	2.2	1.0	3.0	4.0	6.0	9.0
## Captures.pred[3,5]	3.6	1.9	0.0	2.0	3.0	5.0	8.0
## Captures.pred[3,6]	6.9	2.6	2.0	5.0	7.0	9.0	12.0
## Captures.pred[3,7]	6.0	2.4	2.0	4.0	6.0	8.0	11.0
## Captures.pred[3,8]	8.6	2.9	3.0	7.0	8.0	11.0	15.0
## Captures.pred[3,9]	9.1	3.0	4.0	7.0	9.0	11.0	15.0
## Captures.pred[3,10]	5.2	2.4	1.0	4.0	5.0	7.0	11.0
## Captures.pred[3,11]	14.0	3.9	7.0	11.0	14.0	17.0	22.0
## Captures.pred[3,12]	11.4	3.4	5.0	9.0	11.0	14.0	18.0
## Captures.pred[3,13]	26.1	5.1	16.0	23.0	26.0	30.0	37.0
## Captures.pred[3,14]	10.2	3.1	4.0	8.0	10.0	12.0	16.0
## Captures.pred[3,15]	7.1	2.7	2.0	5.0	7.0	9.0	13.0
## Captures.pred[3,16]	7.3	2.8	3.0	5.0	7.0	9.0	14.0
## Captures.pred[3,17]	6.6	2.6	2.0	5.0	7.0	8.0	12.0
## Captures.pred[3,18]	10.3	3.3	5.0	8.0	10.0	12.0	17.0
## Captures.pred[3,19]	4.8	2.2	1.0	3.0	5.0	6.0	10.0
## Captures.pred[3,20]	3.5	1.9	0.0	2.0	3.0	5.0	8.0
## Captures.pred[3,21]	11.6	3.4	5.0	9.0	12.0	14.0	19.0
## Captures.pred[3,22]	12.7	3.6	6.0	10.0	12.0	15.0	20.0
## Captures.pred[3,23]	36.4	6.5	25.0	32.0	36.0	41.0	50.0
## Captures.pred[3,24]	16.4	4.2	8.0	13.0	16.0	19.0	25.0
## Captures.pred[3,25]	6.2	2.4	2.0	4.0	6.0	8.0	11.0
## Captures.pred[3,26]	10.9	3.3	5.0	9.0	11.0	13.0	18.0
## Captures.pred[3,27]	12.4	3.5	6.0	10.0	12.0	15.0	20.0
## Captures.pred[3,28]	17.1	4.3	10.0	14.0	17.0	20.0	26.0
## Captures.pred[3,29]	9.5	3.2	4.0	7.0	9.0	12.0	16.0
## Captures.pred[3,30]	7.9	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[3,31]	5.6	2.3	2.0	4.0	5.0	7.0	11.0
## Captures.pred[3,32]	6.3	2.5	2.0	4.0	6.0	8.0	11.0
## Captures.pred[3,33]	9.3	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[3,34]	6.2	2.4	2.0	5.0	6.0	8.0	11.0
## Captures.pred[3,35]	5.3	2.4	1.0	4.0	5.0	7.0	10.0
## Captures.pred[4,1]	5.5	2.4	2.0	4.0	5.0	7.0	10.0
## Captures.pred[4,2]	4.5	2.2	1.0	3.0	4.0	6.0	9.0
## Captures.pred[4,3]	4.6	2.1	1.0	3.0	5.0	6.0	9.0
## Captures.pred[4,4]	3.3	1.8	0.0	2.0	3.0	4.0	7.0
## Captures.pred[4,5]	3.0	1.8	0.0	2.0	3.0	4.0	7.0
## Captures.pred[4,6]	10.9	3.3	5.0	9.0	11.0	13.0	17.0
## Captures.pred[4,7]	7.2	2.6	3.0	5.0	7.0	9.0	12.0
## Captures.pred[4,8]	8.3	2.9	3.0	6.0	8.0	10.0	15.0
## Captures.pred[4,9]	6.6	2.8	2.0	5.0	6.0	8.0	12.0
## Captures.pred[4,10]	4.5	2.3	1.0	3.0	4.0	6.0	9.0
## Captures.pred[4,11]	14.1	3.9	7.0	11.0	14.0	17.0	22.0
## Captures.pred[4,12]	17.0	4.3	9.0	14.0	17.0	20.0	26.0
## Captures.pred[4,13]	17.0	4.2	9.0	14.0	17.0	20.0	25.0

## Captures.pred[4,14]	7.6	2.8	3.0	6.0	7.0	9.0	13.0
## Captures.pred[4,15]	6.9	2.5	3.0	5.0	7.0	8.7	12.0
## Captures.pred[4,16]	7.6	2.7	3.0	6.0	8.0	9.0	13.0
## Captures.pred[4,17]	8.8	3.0	4.0	7.0	9.0	11.0	15.0
## Captures.pred[4,18]	9.7	3.2	4.0	7.0	9.0	12.0	17.0
## Captures.pred[4,19]	3.6	1.9	0.0	2.0	3.5	5.0	7.0
## Captures.pred[4,20]	2.9	1.7	0.0	2.0	3.0	4.0	7.0
## Captures.pred[4,21]	15.9	4.0	8.0	13.0	16.0	19.0	24.0
## Captures.pred[4,22]	31.3	5.7	20.0	27.2	31.0	35.0	43.0
## Captures.pred[4,23]	55.2	7.9	41.0	50.0	55.0	60.0	71.0
## Captures.pred[4,24]	7.0	2.7	2.0	5.0	7.0	9.0	13.0
## Captures.pred[4,25]	5.4	2.3	1.0	4.0	5.0	7.0	10.0
## Captures.pred[4,26]	20.0	4.6	11.0	17.0	20.0	23.0	29.0
## Captures.pred[4,27]	34.8	6.3	23.0	31.0	35.0	39.0	47.0
## Captures.pred[4,28]	126.5	12.4	102.0	118.0	126.0	135.0	150.0
## Captures.pred[4,29]	11.5	3.5	5.0	9.0	11.0	14.0	19.0
## Captures.pred[4,30]	7.0	2.7	2.0	5.0	7.0	9.0	13.0
## Captures.pred[4,31]	7.8	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[4,32]	8.0	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[4,33]	8.0	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[4,34]	5.7	2.5	2.0	4.0	6.0	7.0	11.0
## Captures.pred[4,35]	5.0	2.2	1.0	3.0	5.0	6.0	10.0
## Captures.pred[5,1]	4.6	2.2	1.0	3.0	4.0	6.0	9.0
## Captures.pred[5,2]	4.7	2.2	1.0	3.0	5.0	6.0	9.0
## Captures.pred[5,3]	6.6	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[5,4]	4.0	2.1	1.0	3.0	4.0	5.0	8.0
## Captures.pred[5,5]	4.1	2.1	1.0	3.0	4.0	5.0	9.0
## Captures.pred[5,6]	7.4	2.8	3.0	5.0	7.0	9.0	13.0
## Captures.pred[5,7]	8.4	3.0	3.0	6.0	8.0	10.0	15.0
## Captures.pred[5,8]	14.6	3.9	8.0	12.0	14.0	17.0	23.0
## Captures.pred[5,9]	5.9	2.4	2.0	4.0	6.0	7.0	11.0
## Captures.pred[5,10]	7.8	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[5,11]	12.2	3.4	6.0	10.0	12.0	14.0	19.0
## Captures.pred[5,12]	14.8	4.0	7.0	12.0	15.0	17.0	23.0
## Captures.pred[5,13]	16.6	4.3	9.0	13.0	16.0	19.0	25.0
## Captures.pred[5,14]	8.0	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[5,15]	7.6	2.7	3.0	6.0	7.0	9.0	13.0
## Captures.pred[5,16]	7.5	2.8	3.0	5.0	7.0	9.0	14.0
## Captures.pred[5,17]	8.3	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[5,18]	9.3	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[5,19]	3.6	1.9	1.0	2.0	3.0	5.0	8.0
## Captures.pred[5,20]	3.7	1.9	0.0	2.0	3.0	5.0	7.0
## Captures.pred[5,21]	10.9	3.3	5.0	9.0	11.0	13.0	18.0
## Captures.pred[5,22]	16.7	4.2	9.0	14.0	17.0	20.0	25.0
## Captures.pred[5,23]	23.9	5.2	14.0	20.0	24.0	28.0	34.0
## Captures.pred[5,24]	5.9	2.5	2.0	4.0	6.0	7.0	11.0
## Captures.pred[5,25]	6.3	2.5	2.0	5.0	6.0	8.0	11.0
## Captures.pred[5,26]	10.8	3.3	4.0	8.0	11.0	13.0	18.0
## Captures.pred[5,27]	19.4	4.5	11.0	16.0	19.0	22.0	28.0
## Captures.pred[5,28]	32.4	6.0	21.0	28.2	32.0	36.0	46.0
## Captures.pred[5,29]	10.3	3.4	4.0	8.0	10.0	13.0	18.0
## Captures.pred[5,30]	10.0	3.0	5.0	8.0	10.0	12.0	16.0
## Captures.pred[5,31]	5.6	2.4	1.0	4.0	5.0	7.0	11.0
## Captures.pred[5,32]	7.0	2.7	2.0	5.0	7.0	9.0	13.0

## Captures.pred[5,33]	9.6	3.2	4.0	7.0	9.0	11.0	17.0
## Captures.pred[5,34]	5.2	2.3	1.0	4.0	5.0	7.0	10.0
## Captures.pred[5,35]	5.3	2.3	1.0	4.0	5.0	7.0	11.0
## Captures.pred[6,1]	6.4	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[6,2]	6.3	2.5	2.0	5.0	6.0	8.0	12.0
## Captures.pred[6,3]	5.7	2.4	2.0	4.0	6.0	7.0	11.0
## Captures.pred[6,4]	4.5	2.1	1.0	3.0	4.0	6.0	9.0
## Captures.pred[6,5]	6.6	2.5	2.0	5.0	6.0	8.0	12.0
## Captures.pred[6,6]	8.7	3.0	3.0	6.0	9.0	11.0	15.0
## Captures.pred[6,7]	9.9	3.2	4.0	8.0	10.0	12.0	16.0
## Captures.pred[6,8]	8.7	3.0	3.0	7.0	9.0	11.0	15.0
## Captures.pred[6,9]	6.5	2.7	2.0	5.0	6.0	8.0	13.0
## Captures.pred[6,10]	12.3	3.7	6.0	10.0	12.0	15.0	20.0
## Captures.pred[6,11]	21.9	4.8	12.0	19.0	22.0	25.0	32.0
## Captures.pred[6,12]	19.2	4.5	11.0	16.0	19.0	22.0	28.0
## Captures.pred[6,13]	13.7	3.7	7.0	11.0	14.0	16.0	22.0
## Captures.pred[6,14]	10.3	3.2	5.0	8.0	10.0	12.0	18.0
## Captures.pred[6,15]	21.0	4.6	12.0	18.0	21.0	24.0	31.0
## Captures.pred[6,16]	8.9	3.0	3.0	7.0	9.0	11.0	15.0
## Captures.pred[6,17]	9.5	2.9	4.0	7.0	9.0	11.0	16.0
## Captures.pred[6,18]	9.3	3.1	4.0	7.0	9.0	11.0	15.0
## Captures.pred[6,19]	4.7	2.2	1.0	3.0	5.0	6.0	10.0
## Captures.pred[6,20]	21.4	4.9	13.0	18.0	21.0	24.0	32.0
## Captures.pred[6,21]	71.0	9.0	55.0	64.2	71.0	77.0	89.0
## Captures.pred[6,22]	19.6	4.6	11.0	16.0	19.0	23.0	29.0
## Captures.pred[6,23]	12.9	3.5	6.0	10.0	13.0	15.0	20.0
## Captures.pred[6,24]	7.5	2.8	3.0	6.0	7.0	9.0	13.0
## Captures.pred[6,25]	44.2	7.0	32.0	39.0	44.0	49.0	59.0
## Captures.pred[6,26]	84.5	10.1	64.0	78.0	84.0	91.0	103.0
## Captures.pred[6,27]	36.2	6.4	25.0	32.0	36.0	41.0	49.0
## Captures.pred[6,28]	11.8	3.5	5.0	9.0	12.0	14.0	19.0
## Captures.pred[6,29]	9.4	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[6,30]	13.3	3.7	7.0	11.0	13.0	16.0	21.0
## Captures.pred[6,31]	7.6	2.7	3.0	6.0	7.5	9.0	13.0
## Captures.pred[6,32]	8.1	3.0	3.0	6.0	8.0	10.0	14.0
## Captures.pred[6,33]	7.4	2.7	3.0	6.0	7.0	9.0	13.0
## Captures.pred[6,34]	6.8	2.7	2.0	5.0	7.0	8.0	12.0
## Captures.pred[6,35]	46.0	7.4	33.0	41.0	46.0	51.0	61.0
## Captures.pred[7,1]	7.0	2.6	3.0	5.0	7.0	9.0	12.0
## Captures.pred[7,2]	7.6	2.9	3.0	6.0	7.0	10.0	13.0
## Captures.pred[7,3]	5.6	2.4	2.0	4.0	5.0	7.0	11.0
## Captures.pred[7,4]	4.9	2.2	1.0	3.0	5.0	6.0	10.0
## Captures.pred[7,5]	7.0	2.5	2.0	5.0	7.0	9.0	12.0
## Captures.pred[7,6]	10.9	3.5	5.0	8.0	11.0	13.0	18.0
## Captures.pred[7,7]	13.8	3.7	7.0	11.0	14.0	16.0	22.0
## Captures.pred[7,8]	9.8	3.2	4.0	8.0	10.0	12.0	16.0
## Captures.pred[7,9]	7.1	2.6	3.0	5.0	7.0	9.0	13.0
## Captures.pred[7,10]	9.8	3.1	4.0	7.0	10.0	12.0	16.0
## Captures.pred[7,11]	24.6	5.3	14.0	21.0	24.0	28.0	35.0
## Captures.pred[7,12]	23.5	5.1	14.0	20.0	23.0	27.0	34.0
## Captures.pred[7,13]	18.2	4.5	10.0	15.0	18.0	21.0	27.0
## Captures.pred[7,14]	12.8	3.6	6.0	10.0	13.0	15.0	21.0
## Captures.pred[7,15]	19.5	4.4	12.0	17.0	19.0	22.0	29.0
## Captures.pred[7,16]	12.9	3.6	6.0	10.0	13.0	15.0	21.0

## Captures.pred[7,17]	14.6	3.9	8.0	12.0	14.0	17.0	23.0
## Captures.pred[7,18]	9.9	3.2	4.0	8.0	10.0	12.0	17.0
## Captures.pred[7,19]	5.5	2.2	2.0	4.0	5.0	7.0	10.0
## Captures.pred[7,20]	6.0	2.4	2.0	4.0	6.0	8.0	11.0
## Captures.pred[7,21]	48.1	7.3	35.0	43.0	48.0	53.0	63.0
## Captures.pred[7,22]	33.0	5.8	22.0	29.0	33.0	37.0	45.0
## Captures.pred[7,23]	26.7	5.3	17.0	23.0	26.0	30.0	37.0
## Captures.pred[7,24]	9.4	3.0	4.0	7.0	9.0	11.0	16.0
## Captures.pred[7,25]	9.5	3.1	4.0	7.0	9.0	12.0	16.0
## Captures.pred[7,26]	74.9	9.5	56.0	69.0	75.0	81.0	92.0
## Captures.pred[7,27]	48.5	7.2	36.0	44.0	48.0	53.0	64.0
## Captures.pred[7,28]	22.5	4.8	14.0	19.0	22.0	26.0	32.0
## Captures.pred[7,29]	10.1	3.3	4.0	8.0	10.0	12.0	17.0
## Captures.pred[7,30]	15.4	4.2	8.0	13.0	15.0	18.0	24.0
## Captures.pred[7,31]	10.1	3.2	4.0	8.0	10.0	12.0	17.0
## Captures.pred[7,32]	11.8	3.5	6.0	9.0	12.0	14.0	19.0
## Captures.pred[7,33]	12.0	3.5	5.0	9.0	12.0	14.0	19.0
## Captures.pred[7,34]	8.0	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[7,35]	8.1	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[8,1]	8.2	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[8,2]	5.3	2.3	1.0	4.0	5.0	7.0	10.0
## Captures.pred[8,3]	4.2	2.0	1.0	3.0	4.0	6.0	8.0
## Captures.pred[8,4]	4.3	2.1	1.0	3.0	4.0	6.0	9.0
## Captures.pred[8,5]	6.6	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[8,6]	14.7	3.9	7.0	12.0	15.0	17.0	22.0
## Captures.pred[8,7]	9.2	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[8,8]	6.3	2.5	2.0	5.0	6.0	8.0	12.0
## Captures.pred[8,9]	8.5	3.0	3.0	6.0	8.0	10.0	15.0
## Captures.pred[8,10]	10.1	3.3	4.0	8.0	10.0	12.0	17.0
## Captures.pred[8,11]	33.4	6.1	23.0	29.0	33.0	37.0	47.0
## Captures.pred[8,12]	15.6	4.0	8.0	13.0	15.0	18.0	24.0
## Captures.pred[8,13]	10.0	3.3	4.0	8.0	10.0	12.0	17.0
## Captures.pred[8,14]	10.8	3.3	5.0	8.0	11.0	13.0	17.0
## Captures.pred[8,15]	17.2	4.3	10.0	14.0	17.0	20.0	26.0
## Captures.pred[8,16]	16.4	4.2	9.0	13.2	16.0	19.0	25.0
## Captures.pred[8,17]	10.4	3.2	5.0	8.0	10.0	13.0	17.0
## Captures.pred[8,18]	4.2	2.1	1.0	3.0	4.0	5.0	9.0
## Captures.pred[8,19]	4.8	2.2	1.0	3.0	5.0	6.0	10.0
## Captures.pred[8,20]	5.9	2.4	1.0	4.0	6.0	7.0	11.0
## Captures.pred[8,21]	44.2	7.2	31.0	39.0	44.0	49.0	59.0
## Captures.pred[8,22]	11.9	3.4	6.0	10.0	12.0	14.0	19.0
## Captures.pred[8,23]	7.9	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[8,24]	8.4	3.0	3.0	6.0	8.0	10.0	15.0
## Captures.pred[8,25]	12.3	3.5	6.0	10.0	12.0	15.0	19.0
## Captures.pred[8,26]	74.1	9.5	54.0	68.0	74.0	80.0	94.0
## Captures.pred[8,27]	14.5	3.8	8.0	12.0	14.0	17.0	23.0
## Captures.pred[8,28]	9.5	3.2	4.0	7.0	9.0	12.0	16.0
## Captures.pred[8,29]	10.6	3.3	5.0	8.0	10.0	13.0	17.0
## Captures.pred[8,30]	118.4	12.4	96.0	110.0	118.0	127.0	144.0
## Captures.pred[8,31]	77.1	9.6	59.0	71.0	77.0	84.0	97.0
## Captures.pred[8,32]	8.1	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[8,33]	6.8	2.5	2.0	5.0	7.0	8.0	12.0
## Captures.pred[8,34]	7.5	2.8	3.0	5.0	7.0	9.0	13.0
## Captures.pred[8,35]	10.1	3.3	4.0	8.0	10.0	12.0	17.0

## Captures.pred[9,1]	3.8	2.0	1.0	2.0	4.0	5.0	8.0
## Captures.pred[9,2]	3.6	2.0	0.0	2.0	3.0	5.0	8.0
## Captures.pred[9,3]	3.2	1.8	0.0	2.0	3.0	4.0	7.0
## Captures.pred[9,4]	3.9	2.0	0.0	3.0	4.0	5.0	8.0
## Captures.pred[9,5]	22.7	4.9	14.0	19.0	22.0	26.0	33.0
## Captures.pred[9,6]	7.5	2.8	3.0	5.2	7.0	9.0	14.0
## Captures.pred[9,7]	9.2	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[9,8]	6.3	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[9,9]	13.0	3.7	6.0	10.0	13.0	15.0	21.0
## Captures.pred[9,10]	8.6	2.8	3.0	7.0	9.0	10.0	14.0
## Captures.pred[9,11]	12.0	3.6	5.0	9.0	12.0	14.0	20.0
## Captures.pred[9,12]	21.2	4.8	12.0	18.0	21.0	24.0	31.0
## Captures.pred[9,13]	8.5	3.0	3.0	6.0	8.0	10.0	15.0
## Captures.pred[9,14]	11.3	3.4	6.0	9.0	11.0	13.0	18.0
## Captures.pred[9,15]	18.2	4.3	11.0	15.0	18.0	21.0	27.0
## Captures.pred[9,16]	7.2	2.7	2.0	5.0	7.0	9.0	13.0
## Captures.pred[9,17]	10.2	3.3	4.0	8.0	10.0	12.0	17.0
## Captures.pred[9,18]	5.5	2.3	2.0	4.0	5.0	7.0	11.0
## Captures.pred[9,19]	4.4	2.1	1.0	3.0	4.0	6.0	9.0
## Captures.pred[9,20]	6.1	2.5	2.0	4.0	6.0	8.0	11.0
## Captures.pred[9,21]	33.2	5.7	22.0	29.0	33.0	37.0	45.0
## Captures.pred[9,22]	21.1	4.8	12.0	18.0	21.0	24.0	31.0
## Captures.pred[9,23]	7.3	2.7	2.0	5.0	7.0	9.0	13.0
## Captures.pred[9,24]	10.5	3.3	5.0	8.0	10.0	13.0	17.0
## Captures.pred[9,25]	14.1	3.8	7.0	12.0	14.0	17.0	22.0
## Captures.pred[9,26]	113.2	12.5	91.0	105.0	113.0	121.0	138.0
## Captures.pred[9,27]	21.3	4.8	12.0	18.0	21.0	24.0	31.0
## Captures.pred[9,28]	9.4	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[9,29]	14.4	3.9	7.0	12.0	14.0	17.0	22.0
## Captures.pred[9,30]	11.4	3.4	5.0	9.0	11.0	14.0	19.0
## Captures.pred[9,31]	8.0	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[9,32]	9.4	3.2	4.0	7.0	9.0	11.0	16.0
## Captures.pred[9,33]	6.6	2.5	2.0	5.0	6.0	8.0	12.0
## Captures.pred[9,34]	6.9	2.5	2.0	5.0	7.0	9.0	12.0
## Captures.pred[9,35]	10.2	3.2	5.0	8.0	10.0	12.0	17.0
## Captures.pred[10,1]	4.0	2.0	1.0	3.0	4.0	5.0	8.0
## Captures.pred[10,2]	2.7	1.7	0.0	1.0	3.0	4.0	6.0
## Captures.pred[10,3]	3.6	2.0	0.0	2.0	4.0	5.0	8.0
## Captures.pred[10,4]	3.7	1.9	1.0	2.0	3.0	5.0	8.0
## Captures.pred[10,5]	5.5	2.3	1.0	4.0	5.0	7.0	11.0
## Captures.pred[10,6]	5.9	2.5	2.0	4.0	6.0	8.0	11.0
## Captures.pred[10,7]	4.9	2.3	1.0	3.0	5.0	6.0	10.0
## Captures.pred[10,8]	5.9	2.4	2.0	4.0	6.0	7.0	11.0
## Captures.pred[10,9]	6.8	2.7	2.0	5.0	7.0	8.0	13.0
## Captures.pred[10,10]	8.0	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[10,11]	9.4	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[10,12]	7.5	2.8	3.0	5.0	7.0	9.0	13.0
## Captures.pred[10,13]	11.2	3.3	5.0	9.0	11.0	13.0	17.0
## Captures.pred[10,14]	11.3	3.4	5.0	9.0	11.0	14.0	19.0
## Captures.pred[10,15]	19.5	4.7	11.0	16.0	19.0	23.0	29.0
## Captures.pred[10,16]	6.9	2.6	3.0	5.0	7.0	9.0	12.0
## Captures.pred[10,17]	3.9	2.0	1.0	2.0	4.0	5.0	8.0
## Captures.pred[10,18]	5.4	2.4	1.0	4.0	5.0	7.0	11.0
## Captures.pred[10,19]	4.1	2.0	1.0	3.0	4.0	5.0	9.0



## Captures.pred[10,20]	7.9	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[10,21]	10.8	3.4	5.0	9.0	11.0	13.0	18.0
## Captures.pred[10,22]	6.5	2.7	2.0	5.0	6.0	8.0	12.0
## Captures.pred[10,23]	9.1	3.0	3.0	7.0	9.0	11.0	15.0
## Captures.pred[10,24]	11.0	3.4	5.0	9.0	11.0	13.0	18.0
## Captures.pred[10,25]	18.1	4.3	10.0	15.0	18.0	21.0	27.0
## Captures.pred[10,26]	13.8	3.8	7.0	11.0	14.0	16.0	22.0
## Captures.pred[10,27]	6.3	2.5	2.0	4.0	6.0	8.0	12.0
## Captures.pred[10,28]	9.9	3.1	4.0	8.0	10.0	12.0	16.0
## Captures.pred[10,29]	9.1	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[10,30]	10.4	3.2	5.0	8.0	10.0	12.7	17.0
## Captures.pred[10,31]	19.0	4.6	10.0	16.0	19.0	22.0	28.0
## Captures.pred[10,32]	3.7	2.0	0.0	2.0	4.0	5.0	8.0
## Captures.pred[10,33]	8.4	2.8	4.0	6.0	8.0	10.0	14.0
## Captures.pred[10,34]	8.0	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[10,35]	9.3	3.1	4.0	7.0	9.0	11.0	16.0
## Captures.pred[11,1]	8.2	2.8	3.0	6.0	8.0	10.0	14.0
## Captures.pred[11,2]	6.4	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[11,3]	13.7	3.8	7.0	11.0	13.0	16.0	22.0
## Captures.pred[11,4]	13.6	3.8	7.0	11.0	14.0	16.0	21.0
## Captures.pred[11,5]	26.7	5.3	17.0	23.0	27.0	30.0	37.0
## Captures.pred[11,6]	14.8	3.8	8.0	12.0	15.0	17.0	23.0
## Captures.pred[11,7]	12.7	3.5	6.0	10.0	13.0	15.0	20.0
## Captures.pred[11,8]	44.0	7.0	31.0	39.0	44.0	48.0	58.0
## Captures.pred[11,9]	19.3	4.3	11.0	16.0	19.0	22.0	28.0
## Captures.pred[11,10]	23.3	5.1	14.0	20.0	23.0	27.0	34.0
## Captures.pred[11,11]	22.0	4.7	13.0	19.0	22.0	25.0	31.0
## Captures.pred[11,12]	21.0	4.7	12.0	18.0	21.0	24.0	31.0
## Captures.pred[11,13]	34.8	6.1	23.0	31.0	35.0	39.0	47.0
## Captures.pred[11,14]	37.0	6.4	25.0	33.0	37.0	41.0	50.0
## Captures.pred[11,15]	41.1	6.8	29.0	37.0	41.0	46.0	54.0
## Captures.pred[11,16]	11.9	3.6	6.0	10.0	12.0	14.0	20.0
## Captures.pred[11,17]	11.2	3.5	5.0	9.0	11.0	13.0	19.0
## Captures.pred[11,18]	18.8	4.5	11.0	16.0	19.0	22.0	28.0
## Captures.pred[11,19]	13.4	3.7	7.0	11.0	13.0	16.0	21.0
## Captures.pred[11,20]	17.8	4.3	10.0	15.0	18.0	21.0	27.0
## Captures.pred[11,21]	18.0	4.2	10.0	15.0	18.0	21.0	27.0
## Captures.pred[11,22]	17.4	4.2	10.0	15.0	17.0	20.0	26.0
## Captures.pred[11,23]	39.3	6.2	28.0	35.0	39.0	43.0	52.0
## Captures.pred[11,24]	38.8	6.6	26.0	34.0	38.0	43.0	52.0
## Captures.pred[11,25]	31.8	5.7	21.0	28.0	32.0	35.0	43.0
## Captures.pred[11,26]	17.4	4.2	9.0	14.0	17.0	20.0	26.0
## Captures.pred[11,27]	16.7	4.2	9.0	14.0	17.0	19.7	25.0
## Captures.pred[11,28]	72.4	9.0	56.0	66.0	72.0	78.0	91.0
## Captures.pred[11,29]	50.1	7.4	36.0	45.0	50.0	55.0	65.0
## Captures.pred[11,30]	31.8	5.7	21.0	28.0	32.0	36.0	43.0
## Captures.pred[11,31]	15.3	4.0	8.0	13.0	15.0	18.0	24.0
## Captures.pred[11,32]	8.9	3.1	4.0	7.0	9.0	11.0	15.0
## Captures.pred[11,33]	26.6	5.2	17.0	23.0	27.0	30.0	37.0
## Captures.pred[11,34]	23.5	5.0	14.0	20.0	23.5	27.0	34.0
## Captures.pred[11,35]	20.6	4.6	12.0	17.0	20.0	23.0	30.0
## Captures.pred[12,1]	4.0	2.2	1.0	2.0	4.0	5.0	9.0
## Captures.pred[12,2]	3.5	1.9	0.0	2.0	3.0	5.0	7.0
## Captures.pred[12,3]	5.5	2.4	1.0	4.0	5.0	7.0	11.0

## Captures.pred[12,4]	6.4	2.6	2.0	5.0	6.0	8.0	12.0
## Captures.pred[12,5]	8.9	3.0	4.0	7.0	9.0	11.0	15.0
## Captures.pred[12,6]	7.2	2.7	3.0	5.0	7.0	9.0	13.0
## Captures.pred[12,7]	8.2	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[12,8]	9.0	3.2	3.0	7.0	9.0	11.0	16.0
## Captures.pred[12,9]	9.2	3.0	4.0	7.0	9.0	11.0	15.0
## Captures.pred[12,10]	14.8	4.0	8.0	12.0	15.0	17.7	22.0
## Captures.pred[12,11]	11.1	3.5	5.0	9.0	11.0	13.0	19.0
## Captures.pred[12,12]	9.8	3.1	4.0	8.0	10.0	12.0	16.0
## Captures.pred[12,13]	13.2	3.6	6.0	11.0	13.0	15.0	20.0
## Captures.pred[12,14]	14.3	3.8	7.0	12.0	14.0	17.0	22.0
## Captures.pred[12,15]	17.3	4.4	10.0	14.0	17.0	20.0	26.0
## Captures.pred[12,16]	5.9	2.5	2.0	4.0	6.0	7.0	11.0
## Captures.pred[12,17]	6.0	2.5	2.0	4.0	6.0	7.0	11.0
## Captures.pred[12,18]	6.0	2.5	2.0	4.0	6.0	7.0	11.0
## Captures.pred[12,19]	7.2	2.6	3.0	5.0	7.0	9.0	13.0
## Captures.pred[12,20]	8.6	3.0	3.0	6.0	8.0	11.0	15.0
## Captures.pred[12,21]	8.3	2.9	3.0	6.0	8.0	10.0	14.0
## Captures.pred[12,22]	8.5	3.0	3.0	6.0	8.0	10.0	15.0
## Captures.pred[12,23]	12.6	3.5	6.0	10.0	13.0	15.0	20.0
## Captures.pred[12,24]	13.9	4.0	7.0	11.0	14.0	16.0	22.0
## Captures.pred[12,25]	14.4	3.9	7.0	12.0	14.0	17.0	23.0
## Captures.pred[12,26]	7.8	2.9	3.0	6.0	7.0	10.0	14.0
## Captures.pred[12,27]	8.0	3.0	3.0	6.0	8.0	10.0	14.0
## Captures.pred[12,28]	14.1	3.8	7.0	12.0	14.0	16.0	22.0
## Captures.pred[12,29]	13.3	3.8	6.0	11.0	13.0	16.0	21.0
## Captures.pred[12,30]	23.9	5.1	15.0	20.0	24.0	27.0	34.0
## Captures.pred[12,31]	4.7	2.2	1.0	3.0	5.0	6.0	9.0
## Captures.pred[12,32]	5.0	2.3	1.0	3.0	5.0	6.0	10.0
## Captures.pred[12,33]	9.7	3.1	4.0	8.0	10.0	12.0	16.0
## Captures.pred[12,34]	11.0	3.5	5.0	9.0	11.0	13.0	18.0
## Captures.pred[12,35]	11.4	3.4	5.0	9.0	11.0	14.0	18.0
## Captures.pred[13,1]	6.0	2.5	2.0	4.0	6.0	8.0	12.0
## Captures.pred[13,2]	7.1	2.6	2.0	5.0	7.0	9.0	13.0
## Captures.pred[13,3]	9.7	3.2	4.0	7.0	10.0	12.0	16.0
## Captures.pred[13,4]	11.4	3.4	5.0	9.0	11.0	14.0	18.0
## Captures.pred[13,5]	12.8	3.6	6.0	10.0	13.0	15.0	20.0
## Captures.pred[13,6]	11.7	3.6	6.0	9.0	11.0	14.0	19.0
## Captures.pred[13,7]	17.0	4.0	10.0	14.0	17.0	20.0	25.0
## Captures.pred[13,8]	16.7	4.1	10.0	14.0	16.0	19.0	25.0
## Captures.pred[13,9]	16.7	4.2	9.0	14.0	17.0	19.0	25.0
## Captures.pred[13,10]	18.4	4.4	10.0	15.0	18.0	21.0	27.0
## Captures.pred[13,11]	20.4	4.6	11.0	17.0	20.0	24.0	29.0
## Captures.pred[13,12]	38.3	6.3	26.0	34.0	38.0	43.0	52.0
## Captures.pred[13,13]	32.5	5.7	22.0	29.0	32.0	36.0	44.0
## Captures.pred[13,14]	26.9	5.2	17.0	23.0	27.0	30.0	38.0
## Captures.pred[13,15]	21.2	4.6	13.0	18.0	21.0	24.0	30.0
## Captures.pred[13,16]	11.1	3.4	5.0	9.0	11.0	13.0	18.0
## Captures.pred[13,17]	22.3	4.6	13.0	19.0	22.0	25.0	32.0
## Captures.pred[13,18]	13.0	3.5	7.0	11.0	13.0	15.0	21.0
## Captures.pred[13,19]	13.7	3.8	7.0	11.0	14.0	16.0	21.0
## Captures.pred[13,20]	11.0	3.3	5.0	9.0	11.0	13.0	18.0
## Captures.pred[13,21]	14.9	3.9	8.0	12.0	15.0	17.0	23.0
## Captures.pred[13,22]	82.8	9.9	63.0	76.0	83.0	89.0	102.0

```

## Captures.pred[13,23] 51.6 7.3 38.0 46.0 51.0 57.0 66.0
## Captures.pred[13,24] 34.0 5.8 23.0 30.0 34.0 38.0 45.0
## Captures.pred[13,25] 19.1 4.4 11.0 16.0 19.0 22.0 28.0
## Captures.pred[13,26] 14.4 3.8 7.0 12.0 14.0 17.0 22.0
## Captures.pred[13,27] 72.6 9.1 56.0 67.0 72.0 78.0 91.0
## Captures.pred[13,28] 35.6 6.2 24.0 31.0 36.0 40.0 48.0
## Captures.pred[13,29] 23.7 5.0 15.0 20.0 23.0 27.0 34.0
## Captures.pred[13,30] 29.0 5.5 18.0 25.2 29.0 33.0 40.0
## Captures.pred[13,31] 10.0 3.1 4.0 8.0 10.0 12.0 16.0
## Captures.pred[13,32] 31.2 5.8 20.0 27.0 31.0 35.0 43.0
## Captures.pred[13,33] 32.2 5.8 21.0 28.0 32.0 36.0 44.0
## Captures.pred[13,34] 26.2 5.2 17.0 23.0 26.0 29.7 37.0
## Captures.pred[13,35] 15.2 3.9 8.0 13.0 15.0 18.0 23.0
## Captures.pred[14,1] 3.0 1.7 0.0 2.0 3.0 4.0 7.0
## Captures.pred[14,2] 4.4 2.1 1.0 3.0 4.0 6.0 9.0
## Captures.pred[14,3] 5.4 2.4 1.0 4.0 5.0 7.0 10.0
## Captures.pred[14,4] 7.6 2.8 3.0 6.0 7.0 9.0 14.0
## Captures.pred[14,5] 5.6 2.4 2.0 4.0 6.0 7.0 11.0
## Captures.pred[14,6] 5.7 2.4 1.0 4.0 5.0 7.0 11.0
## Captures.pred[14,7] 6.5 2.6 2.0 5.0 6.0 8.0 12.0
## Captures.pred[14,8] 8.4 2.9 3.0 6.0 8.0 10.0 15.0
## Captures.pred[14,9] 11.5 3.4 5.0 9.0 11.0 14.0 19.0
## Captures.pred[14,10] 7.8 2.9 3.0 6.0 7.0 10.0 14.0
## Captures.pred[14,11] 10.8 3.4 5.0 8.0 10.5 13.0 18.0
## Captures.pred[14,12] 14.3 3.8 7.0 12.0 14.0 17.0 22.0
## Captures.pred[14,13] 19.9 4.8 11.0 17.0 20.0 23.0 31.0
## Captures.pred[14,14] 16.6 4.2 9.0 13.0 16.0 19.0 25.0
## Captures.pred[14,15] 12.3 3.6 6.0 10.0 12.0 15.0 20.0
## Captures.pred[14,16] 6.5 2.6 2.0 5.0 6.0 8.0 12.0
## Captures.pred[14,17] 6.7 2.7 2.0 5.0 7.0 8.0 12.0
## Captures.pred[14,18] 6.0 2.6 2.0 4.0 6.0 8.0 12.0
## Captures.pred[14,19] 10.0 3.2 4.0 8.0 10.0 12.0 17.0
## Captures.pred[14,20] 6.3 2.6 2.0 4.0 6.0 8.0 12.0
## Captures.pred[14,21] 8.6 3.0 3.0 7.0 8.0 11.0 15.0
## Captures.pred[14,22] 8.8 2.9 4.0 7.0 9.0 11.0 15.0
## Captures.pred[14,23] 17.5 4.2 10.0 14.0 17.0 20.0 26.0
## Captures.pred[14,24] 21.4 4.8 13.0 18.0 21.0 25.0 31.0
## Captures.pred[14,25] 16.0 4.3 8.0 13.0 16.0 19.0 25.0
## Captures.pred[14,26] 8.0 2.8 3.0 6.0 8.0 10.0 14.0
## Captures.pred[14,27] 8.1 2.9 3.0 6.0 8.0 10.0 14.0
## Captures.pred[14,28] 11.8 3.7 6.0 9.0 12.0 14.0 20.0
## Captures.pred[14,29] 18.6 4.4 11.0 16.0 18.0 22.0 28.0
## Captures.pred[14,30] 16.0 4.2 8.0 13.0 16.0 19.0 24.0
## Captures.pred[14,31] 10.3 3.2 5.0 8.0 10.0 12.0 17.0
## Captures.pred[14,32] 5.3 2.2 1.0 4.0 5.0 7.0 10.0
## Captures.pred[14,33] 10.2 3.3 5.0 8.0 10.0 12.0 17.0
## Captures.pred[14,34] 20.6 4.8 12.0 17.0 20.0 24.0 30.0
## Captures.pred[14,35] 11.9 3.5 5.0 9.0 12.0 14.0 20.0
## deviance 15302.5 6.8 15290.0 15300.0 15300.0 15310.0 15320.0
##
## Rhat n.eff
## beta0 1 800
## beta1 1 1000
## beta2 1 660
## beta3 1 1000

```

## beta4	1	450
## v[1]	1	510
## v[2]	1	690
## v[3]	1	700
## v[4]	1	1000
## v[5]	1	890
## v[6]	1	410
## v[7]	1	330
## v[8]	1	450
## v[9]	1	310
## v[10]	1	1000
## v[11]	1	1000
## v[12]	1	1000
## v[13]	1	890
## v[14]	1	450
## sdv	1	1000
## resid[1,1]	1	1000
## resid[1,2]	1	1000
## resid[1,3]	1	300
## resid[1,4]	1	1000
## resid[1,5]	1	1000
## resid[1,6]	1	1000
## resid[1,7]	1	730
## resid[1,8]	1	710
## resid[1,9]	1	1000
## resid[1,10]	1	1000
## resid[1,11]	1	1000
## resid[1,12]	1	1000
## resid[1,13]	1	970
## resid[1,14]	1	930
## resid[1,15]	1	1000
## resid[1,16]	1	730
## resid[1,17]	1	1000
## resid[1,18]	1	620
## resid[1,19]	1	590
## resid[1,20]	1	460
## resid[1,21]	1	1000
## resid[1,22]	1	690
## resid[1,23]	1	1000
## resid[1,24]	1	670
## resid[1,25]	1	1000
## resid[1,26]	1	1000
## resid[1,27]	1	710
## resid[1,28]	1	1000
## resid[1,29]	1	1000
## resid[1,30]	1	1000
## resid[1,31]	1	1000
## resid[1,32]	1	970
## resid[1,33]	1	1000
## resid[1,34]	1	870
## resid[1,35]	1	500
## resid[2,1]	1	1000
## resid[2,2]	1	1000
## resid[2,3]	1	630

## resid[2,4]	1	1000
## resid[2,5]	1	1000
## resid[2,6]	1	1000
## resid[2,7]	1	600
## resid[2,8]	1	1000
## resid[2,9]	1	1000
## resid[2,10]	1	1000
## resid[2,11]	1	1000
## resid[2,12]	1	780
## resid[2,13]	1	1000
## resid[2,14]	1	430
## resid[2,15]	1	1000
## resid[2,16]	1	240
## resid[2,17]	1	1000
## resid[2,18]	1	1000
## resid[2,19]	1	1000
## resid[2,20]	1	1000
## resid[2,21]	1	1000
## resid[2,22]	1	1000
## resid[2,23]	1	1000
## resid[2,24]	1	1000
## resid[2,25]	1	1000
## resid[2,26]	1	1000
## resid[2,27]	1	900
## resid[2,28]	1	450
## resid[2,29]	1	1000
## resid[2,30]	1	1000
## resid[2,31]	1	1000
## resid[2,32]	1	770
## resid[2,33]	1	430
## resid[2,34]	1	1000
## resid[2,35]	1	1000
## resid[3,1]	1	980
## resid[3,2]	1	640
## resid[3,3]	1	1000
## resid[3,4]	1	1000
## resid[3,5]	1	1000
## resid[3,6]	1	1000
## resid[3,7]	1	1000
## resid[3,8]	1	1000
## resid[3,9]	1	1000
## resid[3,10]	1	1000
## resid[3,11]	1	1000
## resid[3,12]	1	690
## resid[3,13]	1	1000
## resid[3,14]	1	640
## resid[3,15]	1	390
## resid[3,16]	1	1000
## resid[3,17]	1	780
## resid[3,18]	1	1000
## resid[3,19]	1	680
## resid[3,20]	1	1000
## resid[3,21]	1	1000
## resid[3,22]	1	1000

## resid[3,23]	1	500
## resid[3,24]	1	1000
## resid[3,25]	1	1000
## resid[3,26]	1	1000
## resid[3,27]	1	1000
## resid[3,28]	1	230
## resid[3,29]	1	1000
## resid[3,30]	1	360
## resid[3,31]	1	1000
## resid[3,32]	1	470
## resid[3,33]	1	1000
## resid[3,34]	1	1000
## resid[3,35]	1	1000
## resid[4,1]	1	480
## resid[4,2]	1	420
## resid[4,3]	1	230
## resid[4,4]	1	1000
## resid[4,5]	1	1000
## resid[4,6]	1	1000
## resid[4,7]	1	730
## resid[4,8]	1	700
## resid[4,9]	1	1000
## resid[4,10]	1	980
## resid[4,11]	1	700
## resid[4,12]	1	1000
## resid[4,13]	1	1000
## resid[4,14]	1	690
## resid[4,15]	1	480
## resid[4,16]	1	1000
## resid[4,17]	1	1000
## resid[4,18]	1	910
## resid[4,19]	1	1000
## resid[4,20]	1	550
## resid[4,21]	1	1000
## resid[4,22]	1	690
## resid[4,23]	1	320
## resid[4,24]	1	1000
## resid[4,25]	1	1000
## resid[4,26]	1	1000
## resid[4,27]	1	1000
## resid[4,28]	1	160
## resid[4,29]	1	1000
## resid[4,30]	1	660
## resid[4,31]	1	470
## resid[4,32]	1	1000
## resid[4,33]	1	1000
## resid[4,34]	1	1000
## resid[4,35]	1	380
## resid[5,1]	1	640
## resid[5,2]	1	1000
## resid[5,3]	1	770
## resid[5,4]	1	1000
## resid[5,5]	1	570
## resid[5,6]	1	480

## resid[5,7]	1	1000
## resid[5,8]	1	620
## resid[5,9]	1	770
## resid[5,10]	1	860
## resid[5,11]	1	890
## resid[5,12]	1	480
## resid[5,13]	1	390
## resid[5,14]	1	1000
## resid[5,15]	1	1000
## resid[5,16]	1	470
## resid[5,17]	1	530
## resid[5,18]	1	1000
## resid[5,19]	1	970
## resid[5,20]	1	1000
## resid[5,21]	1	400
## resid[5,22]	1	780
## resid[5,23]	1	500
## resid[5,24]	1	490
## resid[5,25]	1	1000
## resid[5,26]	1	850
## resid[5,27]	1	1000
## resid[5,28]	1	190
## resid[5,29]	1	1000
## resid[5,30]	1	380
## resid[5,31]	1	990
## resid[5,32]	1	1000
## resid[5,33]	1	960
## resid[5,34]	1	1000
## resid[5,35]	1	1000
## resid[6,1]	1	1000
## resid[6,2]	1	1000
## resid[6,3]	1	1000
## resid[6,4]	1	1000
## resid[6,5]	1	820
## resid[6,6]	1	770
## resid[6,7]	1	800
## resid[6,8]	1	1000
## resid[6,9]	1	430
## resid[6,10]	1	830
## resid[6,11]	1	1000
## resid[6,12]	1	1000
## resid[6,13]	1	430
## resid[6,14]	1	1000
## resid[6,15]	1	1000
## resid[6,16]	1	1000
## resid[6,17]	1	1000
## resid[6,18]	1	1000
## resid[6,19]	1	1000
## resid[6,20]	1	280
## resid[6,21]	1	700
## resid[6,22]	1	1000
## resid[6,23]	1	1000
## resid[6,24]	1	1000
## resid[6,25]	1	1000

## resid[6,26]	1	1000
## resid[6,27]	1	540
## resid[6,28]	1	1000
## resid[6,29]	1	480
## resid[6,30]	1	1000
## resid[6,31]	1	1000
## resid[6,32]	1	1000
## resid[6,33]	1	1000
## resid[6,34]	1	1000
## resid[6,35]	1	1000
## resid[7,1]	1	410
## resid[7,2]	1	1000
## resid[7,3]	1	1000
## resid[7,4]	1	1000
## resid[7,5]	1	1000
## resid[7,6]	1	1000
## resid[7,7]	1	1000
## resid[7,8]	1	1000
## resid[7,9]	1	330
## resid[7,10]	1	1000
## resid[7,11]	1	1000
## resid[7,12]	1	1000
## resid[7,13]	1	1000
## resid[7,14]	1	1000
## resid[7,15]	1	1000
## resid[7,16]	1	820
## resid[7,17]	1	1000
## resid[7,18]	1	1000
## resid[7,19]	1	1000
## resid[7,20]	1	1000
## resid[7,21]	1	1000
## resid[7,22]	1	1000
## resid[7,23]	1	1000
## resid[7,24]	1	1000
## resid[7,25]	1	1000
## resid[7,26]	1	780
## resid[7,27]	1	1000
## resid[7,28]	1	890
## resid[7,29]	1	820
## resid[7,30]	1	1000
## resid[7,31]	1	210
## resid[7,32]	1	1000
## resid[7,33]	1	560
## resid[7,34]	1	850
## resid[7,35]	1	1000
## resid[8,1]	1	1000
## resid[8,2]	1	1000
## resid[8,3]	1	1000
## resid[8,4]	1	1000
## resid[8,5]	1	1000
## resid[8,6]	1	790
## resid[8,7]	1	370
## resid[8,8]	1	610
## resid[8,9]	1	840



## resid[8,10]	1	430
## resid[8,11]	1	370
## resid[8,12]	1	230
## resid[8,13]	1	1000
## resid[8,14]	1	1000
## resid[8,15]	1	1000
## resid[8,16]	1	1000
## resid[8,17]	1	1000
## resid[8,18]	1	1000
## resid[8,19]	1	1000
## resid[8,20]	1	1000
## resid[8,21]	1	1000
## resid[8,22]	1	950
## resid[8,23]	1	1000
## resid[8,24]	1	1000
## resid[8,25]	1	1000
## resid[8,26]	1	1000
## resid[8,27]	1	1000
## resid[8,28]	1	860
## resid[8,29]	1	1000
## resid[8,30]	1	880
## resid[8,31]	1	1000
## resid[8,32]	1	1000
## resid[8,33]	1	380
## resid[8,34]	1	1000
## resid[8,35]	1	530
## resid[9,1]	1	1000
## resid[9,2]	1	1000
## resid[9,3]	1	1000
## resid[9,4]	1	220
## resid[9,5]	1	590
## resid[9,6]	1	270
## resid[9,7]	1	1000
## resid[9,8]	1	1000
## resid[9,9]	1	270
## resid[9,10]	1	970
## resid[9,11]	1	1000
## resid[9,12]	1	350
## resid[9,13]	1	1000
## resid[9,14]	1	1000
## resid[9,15]	1	1000
## resid[9,16]	1	1000
## resid[9,17]	1	1000
## resid[9,18]	1	1000
## resid[9,19]	1	740
## resid[9,20]	1	1000
## resid[9,21]	1	1000
## resid[9,22]	1	1000
## resid[9,23]	1	1000
## resid[9,24]	1	1000
## resid[9,25]	1	1000
## resid[9,26]	1	1000
## resid[9,27]	1	750
## resid[9,28]	1	220

## resid[9,29]	1	1000
## resid[9,30]	1	1000
## resid[9,31]	1	1000
## resid[9,32]	1	1000
## resid[9,33]	1	630
## resid[9,34]	1	1000
## resid[9,35]	1	630
## resid[10,1]	1	820
## resid[10,2]	1	470
## resid[10,3]	1	920
## resid[10,4]	1	1000
## resid[10,5]	1	940
## resid[10,6]	1	220
## resid[10,7]	1	1000
## resid[10,8]	1	1000
## resid[10,9]	1	1000
## resid[10,10]	1	1000
## resid[10,11]	1	780
## resid[10,12]	1	1000
## resid[10,13]	1	1000
## resid[10,14]	1	1000
## resid[10,15]	1	1000
## resid[10,16]	1	1000
## resid[10,17]	1	1000
## resid[10,18]	1	1000
## resid[10,19]	1	1000
## resid[10,20]	1	1000
## resid[10,21]	1	1000
## resid[10,22]	1	1000
## resid[10,23]	1	1000
## resid[10,24]	1	930
## resid[10,25]	1	1000
## resid[10,26]	1	830
## resid[10,27]	1	1000
## resid[10,28]	1	1000
## resid[10,29]	1	1000
## resid[10,30]	1	1000
## resid[10,31]	1	1000
## resid[10,32]	1	1000
## resid[10,33]	1	1000
## resid[10,34]	1	1000
## resid[10,35]	1	910
## resid[11,1]	1	620
## resid[11,2]	1	1000
## resid[11,3]	1	870
## resid[11,4]	1	1000
## resid[11,5]	1	1000
## resid[11,6]	1	1000
## resid[11,7]	1	1000
## resid[11,8]	1	1000
## resid[11,9]	1	1000
## resid[11,10]	1	1000
## resid[11,11]	1	910
## resid[11,12]	1	1000

## resid[11,13]	1	1000
## resid[11,14]	1	1000
## resid[11,15]	1	1000
## resid[11,16]	1	270
## resid[11,17]	1	380
## resid[11,18]	1	340
## resid[11,19]	1	1000
## resid[11,20]	1	600
## resid[11,21]	1	1000
## resid[11,22]	1	1000
## resid[11,23]	1	490
## resid[11,24]	1	580
## resid[11,25]	1	1000
## resid[11,26]	1	1000
## resid[11,27]	1	1000
## resid[11,28]	1	340
## resid[11,29]	1	550
## resid[11,30]	1	1000
## resid[11,31]	1	1000
## resid[11,32]	1	1000
## resid[11,33]	1	200
## resid[11,34]	1	480
## resid[11,35]	1	240
## resid[12,1]	1	990
## resid[12,2]	1	1000
## resid[12,3]	1	1000
## resid[12,4]	1	1000
## resid[12,5]	1	1000
## resid[12,6]	1	1000
## resid[12,7]	1	670
## resid[12,8]	1	400
## resid[12,9]	1	1000
## resid[12,10]	1	700
## resid[12,11]	1	1000
## resid[12,12]	1	1000
## resid[12,13]	1	1000
## resid[12,14]	1	1000
## resid[12,15]	1	510
## resid[12,16]	1	1000
## resid[12,17]	1	1000
## resid[12,18]	1	1000
## resid[12,19]	1	1000
## resid[12,20]	1	1000
## resid[12,21]	1	1000
## resid[12,22]	1	1000
## resid[12,23]	1	1000
## resid[12,24]	1	330
## resid[12,25]	1	1000
## resid[12,26]	1	1000
## resid[12,27]	1	1000
## resid[12,28]	1	440
## resid[12,29]	1	1000
## resid[12,30]	1	1000
## resid[12,31]	1	1000

## resid[12,32]	1	400
## resid[12,33]	1	1000
## resid[12,34]	1	450
## resid[12,35]	1	1000
## resid[13,1]	1	1000
## resid[13,2]	1	1000
## resid[13,3]	1	1000
## resid[13,4]	1	1000
## resid[13,5]	1	1000
## resid[13,6]	1	860
## resid[13,7]	1	1000
## resid[13,8]	1	1000
## resid[13,9]	1	1000
## resid[13,10]	1	1000
## resid[13,11]	1	370
## resid[13,12]	1	1000
## resid[13,13]	1	1000
## resid[13,14]	1	1000
## resid[13,15]	1	410
## resid[13,16]	1	550
## resid[13,17]	1	400
## resid[13,18]	1	680
## resid[13,19]	1	1000
## resid[13,20]	1	570
## resid[13,21]	1	1000
## resid[13,22]	1	1000
## resid[13,23]	1	1000
## resid[13,24]	1	1000
## resid[13,25]	1	1000
## resid[13,26]	1	1000
## resid[13,27]	1	1000
## resid[13,28]	1	1000
## resid[13,29]	1	290
## resid[13,30]	1	1000
## resid[13,31]	1	1000
## resid[13,32]	1	650
## resid[13,33]	1	1000
## resid[13,34]	1	1000
## resid[13,35]	1	1000
## resid[14,1]	1	1000
## resid[14,2]	1	1000
## resid[14,3]	1	700
## resid[14,4]	1	1000
## resid[14,5]	1	1000
## resid[14,6]	1	890
## resid[14,7]	1	1000
## resid[14,8]	1	1000
## resid[14,9]	1	710
## resid[14,10]	1	1000
## resid[14,11]	1	1000
## resid[14,12]	1	1000
## resid[14,13]	1	840
## resid[14,14]	1	1000
## resid[14,15]	1	1000

## resid[14,16]	1	1000
## resid[14,17]	1	620
## resid[14,18]	1	1000
## resid[14,19]	1	690
## resid[14,20]	1	570
## resid[14,21]	1	850
## resid[14,22]	1	350
## resid[14,23]	1	1000
## resid[14,24]	1	1000
## resid[14,25]	1	1000
## resid[14,26]	1	890
## resid[14,27]	1	1000
## resid[14,28]	1	330
## resid[14,29]	1	1000
## resid[14,30]	1	1000
## resid[14,31]	1	440
## resid[14,32]	1	810
## resid[14,33]	1	1000
## resid[14,34]	1	1000
## resid[14,35]	1	1000
## Captures.pred[1,1]	1	1000
## Captures.pred[1,2]	1	1000
## Captures.pred[1,3]	1	300
## Captures.pred[1,4]	1	1000
## Captures.pred[1,5]	1	1000
## Captures.pred[1,6]	1	1000
## Captures.pred[1,7]	1	610
## Captures.pred[1,8]	1	710
## Captures.pred[1,9]	1	1000
## Captures.pred[1,10]	1	1000
## Captures.pred[1,11]	1	1000
## Captures.pred[1,12]	1	900
## Captures.pred[1,13]	1	880
## Captures.pred[1,14]	1	930
## Captures.pred[1,15]	1	1000
## Captures.pred[1,16]	1	550
## Captures.pred[1,17]	1	1000
## Captures.pred[1,18]	1	1000
## Captures.pred[1,19]	1	590
## Captures.pred[1,20]	1	300
## Captures.pred[1,21]	1	1000
## Captures.pred[1,22]	1	750
## Captures.pred[1,23]	1	1000
## Captures.pred[1,24]	1	530
## Captures.pred[1,25]	1	1000
## Captures.pred[1,26]	1	1000
## Captures.pred[1,27]	1	1000
## Captures.pred[1,28]	1	1000
## Captures.pred[1,29]	1	1000
## Captures.pred[1,30]	1	1000
## Captures.pred[1,31]	1	1000
## Captures.pred[1,32]	1	940
## Captures.pred[1,33]	1	1000
## Captures.pred[1,34]	1	870

```

## Captures.pred[1,35]      1   490
## Captures.pred[2,1]       1  1000
## Captures.pred[2,2]       1   910
## Captures.pred[2,3]       1   700
## Captures.pred[2,4]       1   640
## Captures.pred[2,5]       1  1000
## Captures.pred[2,6]       1  1000
## Captures.pred[2,7]       1  1000
## Captures.pred[2,8]       1  1000
## Captures.pred[2,9]       1  1000
## Captures.pred[2,10]      1  1000
## Captures.pred[2,11]     1  1000
## Captures.pred[2,12]     1  1000
## Captures.pred[2,13]     1  1000
## Captures.pred[2,14]     1   430
## Captures.pred[2,15]     1  1000
## Captures.pred[2,16]     1   490
## Captures.pred[2,17]     1  1000
## Captures.pred[2,18]     1  1000
## Captures.pred[2,19]     1  1000
## Captures.pred[2,20]     1  1000
## Captures.pred[2,21]     1  1000
## Captures.pred[2,22]     1   940
## Captures.pred[2,23]     1  1000
## Captures.pred[2,24]     1  1000
## Captures.pred[2,25]     1  1000
## Captures.pred[2,26]     1  1000
## Captures.pred[2,27]     1   530
## Captures.pred[2,28]     1   560
## Captures.pred[2,29]     1  1000
## Captures.pred[2,30]     1  1000
## Captures.pred[2,31]     1  1000
## Captures.pred[2,32]     1  1000
## Captures.pred[2,33]     1  1000
## Captures.pred[2,34]     1  1000
## Captures.pred[2,35]     1  1000
## Captures.pred[3,1]      1   830
## Captures.pred[3,2]      1   890
## Captures.pred[3,3]      1  1000
## Captures.pred[3,4]      1  1000
## Captures.pred[3,5]      1  1000
## Captures.pred[3,6]      1  1000
## Captures.pred[3,7]      1  1000
## Captures.pred[3,8]      1  1000
## Captures.pred[3,9]      1  1000
## Captures.pred[3,10]     1  1000
## Captures.pred[3,11]     1  1000
## Captures.pred[3,12]     1   410
## Captures.pred[3,13]     1   670
## Captures.pred[3,14]     1   640
## Captures.pred[3,15]     1   320
## Captures.pred[3,16]     1  1000
## Captures.pred[3,17]     1   720
## Captures.pred[3,18]     1  1000

```

```

## Captures.pred[3,19]      1  1000
## Captures.pred[3,20]      1  1000
## Captures.pred[3,21]      1  1000
## Captures.pred[3,22]      1  1000
## Captures.pred[3,23]      1   480
## Captures.pred[3,24]      1  1000
## Captures.pred[3,25]      1  1000
## Captures.pred[3,26]      1  1000
## Captures.pred[3,27]      1  1000
## Captures.pred[3,28]      1   480
## Captures.pred[3,29]      1  1000
## Captures.pred[3,30]      1   360
## Captures.pred[3,31]      1  1000
## Captures.pred[3,32]      1  1000
## Captures.pred[3,33]      1  1000
## Captures.pred[3,34]      1  1000
## Captures.pred[3,35]      1  1000
## Captures.pred[4,1]       1   810
## Captures.pred[4,2]       1   550
## Captures.pred[4,3]       1   290
## Captures.pred[4,4]       1  1000
## Captures.pred[4,5]       1  1000
## Captures.pred[4,6]       1  1000
## Captures.pred[4,7]       1   460
## Captures.pred[4,8]       1   700
## Captures.pred[4,9]       1  1000
## Captures.pred[4,10]      1  1000
## Captures.pred[4,11]      1   690
## Captures.pred[4,12]      1  1000
## Captures.pred[4,13]      1  1000
## Captures.pred[4,14]      1   690
## Captures.pred[4,15]      1   480
## Captures.pred[4,16]      1  1000
## Captures.pred[4,17]      1  1000
## Captures.pred[4,18]      1   910
## Captures.pred[4,19]      1  1000
## Captures.pred[4,20]      1   530
## Captures.pred[4,21]      1  1000
## Captures.pred[4,22]      1   850
## Captures.pred[4,23]      1   320
## Captures.pred[4,24]      1  1000
## Captures.pred[4,25]      1  1000
## Captures.pred[4,26]      1   850
## Captures.pred[4,27]      1  1000
## Captures.pred[4,28]      1   160
## Captures.pred[4,29]      1  1000
## Captures.pred[4,30]      1   640
## Captures.pred[4,31]      1  1000
## Captures.pred[4,32]      1   690
## Captures.pred[4,33]      1   430
## Captures.pred[4,34]      1  1000
## Captures.pred[4,35]      1   670
## Captures.pred[5,1]       1  1000
## Captures.pred[5,2]       1  1000

```

## Captures.pred[5,3]	1	540
## Captures.pred[5,4]	1	940
## Captures.pred[5,5]	1	770
## Captures.pred[5,6]	1	570
## Captures.pred[5,7]	1	1000
## Captures.pred[5,8]	1	620
## Captures.pred[5,9]	1	730
## Captures.pred[5,10]	1	1000
## Captures.pred[5,11]	1	980
## Captures.pred[5,12]	1	410
## Captures.pred[5,13]	1	390
## Captures.pred[5,14]	1	1000
## Captures.pred[5,15]	1	1000
## Captures.pred[5,16]	1	800
## Captures.pred[5,17]	1	550
## Captures.pred[5,18]	1	1000
## Captures.pred[5,19]	1	1000
## Captures.pred[5,20]	1	1000
## Captures.pred[5,21]	1	400
## Captures.pred[5,22]	1	620
## Captures.pred[5,23]	1	500
## Captures.pred[5,24]	1	360
## Captures.pred[5,25]	1	1000
## Captures.pred[5,26]	1	740
## Captures.pred[5,27]	1	1000
## Captures.pred[5,28]	1	190
## Captures.pred[5,29]	1	1000
## Captures.pred[5,30]	1	210
## Captures.pred[5,31]	1	1000
## Captures.pred[5,32]	1	1000
## Captures.pred[5,33]	1	930
## Captures.pred[5,34]	1	1000
## Captures.pred[5,35]	1	1000
## Captures.pred[6,1]	1	1000
## Captures.pred[6,2]	1	1000
## Captures.pred[6,3]	1	1000
## Captures.pred[6,4]	1	1000
## Captures.pred[6,5]	1	510
## Captures.pred[6,6]	1	1000
## Captures.pred[6,7]	1	800
## Captures.pred[6,8]	1	1000
## Captures.pred[6,9]	1	550
## Captures.pred[6,10]	1	740
## Captures.pred[6,11]	1	1000
## Captures.pred[6,12]	1	1000
## Captures.pred[6,13]	1	430
## Captures.pred[6,14]	1	1000
## Captures.pred[6,15]	1	1000
## Captures.pred[6,16]	1	1000
## Captures.pred[6,17]	1	1000
## Captures.pred[6,18]	1	1000
## Captures.pred[6,19]	1	720
## Captures.pred[6,20]	1	230
## Captures.pred[6,21]	1	710



```

## Captures.pred[6,22]      1 1000
## Captures.pred[6,23]      1 1000
## Captures.pred[6,24]      1 1000
## Captures.pred[6,25]      1 1000
## Captures.pred[6,26]      1  460
## Captures.pred[6,27]      1 1000
## Captures.pred[6,28]      1 1000
## Captures.pred[6,29]      1  480
## Captures.pred[6,30]      1 1000
## Captures.pred[6,31]      1 1000
## Captures.pred[6,32]      1 1000
## Captures.pred[6,33]      1 1000
## Captures.pred[6,34]      1 1000
## Captures.pred[6,35]      1 1000
## Captures.pred[7,1]       1  410
## Captures.pred[7,2]       1 1000
## Captures.pred[7,3]       1 1000
## Captures.pred[7,4]       1 1000
## Captures.pred[7,5]       1 1000
## Captures.pred[7,6]       1 1000
## Captures.pred[7,7]       1 1000
## Captures.pred[7,8]       1 1000
## Captures.pred[7,9]       1 1000
## Captures.pred[7,10]      1  520
## Captures.pred[7,11]      1 1000
## Captures.pred[7,12]      1 1000
## Captures.pred[7,13]      1 1000
## Captures.pred[7,14]      1 1000
## Captures.pred[7,15]      1 1000
## Captures.pred[7,16]      1  860
## Captures.pred[7,17]      1 1000
## Captures.pred[7,18]      1 1000
## Captures.pred[7,19]      1  660
## Captures.pred[7,20]      1 1000
## Captures.pred[7,21]      1 1000
## Captures.pred[7,22]      1 1000
## Captures.pred[7,23]      1 1000
## Captures.pred[7,24]      1 1000
## Captures.pred[7,25]      1 1000
## Captures.pred[7,26]      1  830
## Captures.pred[7,27]      1 1000
## Captures.pred[7,28]      1  890
## Captures.pred[7,29]      1  820
## Captures.pred[7,30]      1 1000
## Captures.pred[7,31]      1  210
## Captures.pred[7,32]      1  970
## Captures.pred[7,33]      1 1000
## Captures.pred[7,34]      1  850
## Captures.pred[7,35]      1 1000
## Captures.pred[8,1]       1 1000
## Captures.pred[8,2]       1 1000
## Captures.pred[8,3]       1 1000
## Captures.pred[8,4]       1 1000
## Captures.pred[8,5]       1 1000

```

```

## Captures.pred[8,6]      1   790
## Captures.pred[8,7]      1   500
## Captures.pred[8,8]      1   720
## Captures.pred[8,9]      1   970
## Captures.pred[8,10]     1   550
## Captures.pred[8,11]     1   500
## Captures.pred[8,12]     1   210
## Captures.pred[8,13]     1  1000
## Captures.pred[8,14]     1  1000
## Captures.pred[8,15]     1   850
## Captures.pred[8,16]     1  1000
## Captures.pred[8,17]     1   970
## Captures.pred[8,18]     1  1000
## Captures.pred[8,19]     1   270
## Captures.pred[8,20]     1  1000
## Captures.pred[8,21]     1  1000
## Captures.pred[8,22]     1  1000
## Captures.pred[8,23]     1  1000
## Captures.pred[8,24]     1  1000
## Captures.pred[8,25]     1   590
## Captures.pred[8,26]     1  1000
## Captures.pred[8,27]     1  1000
## Captures.pred[8,28]     1   860
## Captures.pred[8,29]     1  1000
## Captures.pred[8,30]     1  1000
## Captures.pred[8,31]     1  1000
## Captures.pred[8,32]     1  1000
## Captures.pred[8,33]     1   380
## Captures.pred[8,34]     1   930
## Captures.pred[8,35]     1   660
## Captures.pred[9,1]      1   880
## Captures.pred[9,2]      1  1000
## Captures.pred[9,3]      1  1000
## Captures.pred[9,4]      1   290
## Captures.pred[9,5]      1   590
## Captures.pred[9,6]      1   270
## Captures.pred[9,7]      1  1000
## Captures.pred[9,8]      1  1000
## Captures.pred[9,9]      1   390
## Captures.pred[9,10]     1   780
## Captures.pred[9,11]     1  1000
## Captures.pred[9,12]     1   350
## Captures.pred[9,13]     1  1000
## Captures.pred[9,14]     1  1000
## Captures.pred[9,15]     1   600
## Captures.pred[9,16]     1  1000
## Captures.pred[9,17]     1  1000
## Captures.pred[9,18]     1  1000
## Captures.pred[9,19]     1   640
## Captures.pred[9,20]     1  1000
## Captures.pred[9,21]     1  1000
## Captures.pred[9,22]     1  1000
## Captures.pred[9,23]     1  1000
## Captures.pred[9,24]     1  1000

```

```

## Captures.pred[9,25]      1  1000
## Captures.pred[9,26]      1  1000
## Captures.pred[9,27]      1   750
## Captures.pred[9,28]      1   220
## Captures.pred[9,29]      1  1000
## Captures.pred[9,30]      1  1000
## Captures.pred[9,31]      1  1000
## Captures.pred[9,32]      1  1000
## Captures.pred[9,33]      1   380
## Captures.pred[9,34]      1  1000
## Captures.pred[9,35]      1   550
## Captures.pred[10,1]       1   780
## Captures.pred[10,2]       1   560
## Captures.pred[10,3]       1   710
## Captures.pred[10,4]       1  1000
## Captures.pred[10,5]       1  1000
## Captures.pred[10,6]       1   260
## Captures.pred[10,7]       1  1000
## Captures.pred[10,8]       1  1000
## Captures.pred[10,9]       1   690
## Captures.pred[10,10]      1  1000
## Captures.pred[10,11]      1  1000
## Captures.pred[10,12]      1  1000
## Captures.pred[10,13]      1  1000
## Captures.pred[10,14]      1  1000
## Captures.pred[10,15]      1   660
## Captures.pred[10,16]      1   770
## Captures.pred[10,17]      1  1000
## Captures.pred[10,18]      1  1000
## Captures.pred[10,19]      1  1000
## Captures.pred[10,20]      1  1000
## Captures.pred[10,21]      1  1000
## Captures.pred[10,22]      1  1000
## Captures.pred[10,23]      1  1000
## Captures.pred[10,24]      1  1000
## Captures.pred[10,25]      1  1000
## Captures.pred[10,26]      1   830
## Captures.pred[10,27]      1  1000
## Captures.pred[10,28]      1  1000
## Captures.pred[10,29]      1  1000
## Captures.pred[10,30]      1  1000
## Captures.pred[10,31]      1  1000
## Captures.pred[10,32]      1  1000
## Captures.pred[10,33]      1  1000
## Captures.pred[10,34]      1   790
## Captures.pred[10,35]      1  1000
## Captures.pred[11,1]       1   520
## Captures.pred[11,2]       1   680
## Captures.pred[11,3]       1   870
## Captures.pred[11,4]       1  1000
## Captures.pred[11,5]       1  1000
## Captures.pred[11,6]       1  1000
## Captures.pred[11,7]       1  1000
## Captures.pred[11,8]       1  1000

```

```

## Captures.pred[11,9]      1 1000
## Captures.pred[11,10]     1 1000
## Captures.pred[11,11]     1  910
## Captures.pred[11,12]     1 1000
## Captures.pred[11,13]     1 1000
## Captures.pred[11,14]     1 1000
## Captures.pred[11,15]     1 1000
## Captures.pred[11,16]     1  270
## Captures.pred[11,17]     1  380
## Captures.pred[11,18]     1  300
## Captures.pred[11,19]     1 1000
## Captures.pred[11,20]     1  600
## Captures.pred[11,21]     1 1000
## Captures.pred[11,22]     1 1000
## Captures.pred[11,23]     1  820
## Captures.pred[11,24]     1  800
## Captures.pred[11,25]     1 1000
## Captures.pred[11,26]     1 1000
## Captures.pred[11,27]     1 1000
## Captures.pred[11,28]     1  350
## Captures.pred[11,29]     1  520
## Captures.pred[11,30]     1 1000
## Captures.pred[11,31]     1 1000
## Captures.pred[11,32]     1 1000
## Captures.pred[11,33]     1  280
## Captures.pred[11,34]     1  940
## Captures.pred[11,35]     1  240
## Captures.pred[12,1]      1  670
## Captures.pred[12,2]      1 1000
## Captures.pred[12,3]      1 1000
## Captures.pred[12,4]      1 1000
## Captures.pred[12,5]      1 1000
## Captures.pred[12,6]      1 1000
## Captures.pred[12,7]      1  490
## Captures.pred[12,8]      1  520
## Captures.pred[12,9]      1 1000
## Captures.pred[12,10]     1  700
## Captures.pred[12,11]     1 1000
## Captures.pred[12,12]     1 1000
## Captures.pred[12,13]     1 1000
## Captures.pred[12,14]     1 1000
## Captures.pred[12,15]     1  510
## Captures.pred[12,16]     1 1000
## Captures.pred[12,17]     1 1000
## Captures.pred[12,18]     1 1000
## Captures.pred[12,19]     1  960
## Captures.pred[12,20]     1 1000
## Captures.pred[12,21]     1 1000
## Captures.pred[12,22]     1 1000
## Captures.pred[12,23]     1 1000
## Captures.pred[12,24]     1  210
## Captures.pred[12,25]     1 1000
## Captures.pred[12,26]     1 1000
## Captures.pred[12,27]     1 1000

```

```

## Captures.pred[12,28] 1 320
## Captures.pred[12,29] 1 1000
## Captures.pred[12,30] 1 1000
## Captures.pred[12,31] 1 1000
## Captures.pred[12,32] 1 390
## Captures.pred[12,33] 1 1000
## Captures.pred[12,34] 1 1000
## Captures.pred[12,35] 1 1000
## Captures.pred[13,1] 1 1000
## Captures.pred[13,2] 1 1000
## Captures.pred[13,3] 1 1000
## Captures.pred[13,4] 1 1000
## Captures.pred[13,5] 1 1000
## Captures.pred[13,6] 1 860
## Captures.pred[13,7] 1 1000
## Captures.pred[13,8] 1 1000
## Captures.pred[13,9] 1 1000
## Captures.pred[13,10] 1 1000
## Captures.pred[13,11] 1 370
## Captures.pred[13,12] 1 1000
## Captures.pred[13,13] 1 730
## Captures.pred[13,14] 1 1000
## Captures.pred[13,15] 1 410
## Captures.pred[13,16] 1 550
## Captures.pred[13,17] 1 420
## Captures.pred[13,18] 1 350
## Captures.pred[13,19] 1 1000
## Captures.pred[13,20] 1 550
## Captures.pred[13,21] 1 1000
## Captures.pred[13,22] 1 1000
## Captures.pred[13,23] 1 1000
## Captures.pred[13,24] 1 1000
## Captures.pred[13,25] 1 1000
## Captures.pred[13,26] 1 1000
## Captures.pred[13,27] 1 1000
## Captures.pred[13,28] 1 1000
## Captures.pred[13,29] 1 290
## Captures.pred[13,30] 1 1000
## Captures.pred[13,31] 1 1000
## Captures.pred[13,32] 1 840
## Captures.pred[13,33] 1 1000
## Captures.pred[13,34] 1 1000
## Captures.pred[13,35] 1 1000
## Captures.pred[14,1] 1 1000
## Captures.pred[14,2] 1 1000
## Captures.pred[14,3] 1 1000
## Captures.pred[14,4] 1 1000
## Captures.pred[14,5] 1 1000
## Captures.pred[14,6] 1 570
## Captures.pred[14,7] 1 870
## Captures.pred[14,8] 1 1000
## Captures.pred[14,9] 1 630
## Captures.pred[14,10] 1 1000
## Captures.pred[14,11] 1 1000

```

```
## Captures.pred[14,12]      1 1000
## Captures.pred[14,13]      1 1000
## Captures.pred[14,14]      1 1000
## Captures.pred[14,15]      1 1000
## Captures.pred[14,16]      1 1000
## Captures.pred[14,17]      1  590
## Captures.pred[14,18]      1 1000
## Captures.pred[14,19]      1  590
## Captures.pred[14,20]      1  450
## Captures.pred[14,21]      1 1000
## Captures.pred[14,22]      1  370
## Captures.pred[14,23]      1 1000
## Captures.pred[14,24]      1 1000
## Captures.pred[14,25]      1 1000
## Captures.pred[14,26]      1 1000
## Captures.pred[14,27]      1 1000
## Captures.pred[14,28]      1  490
## Captures.pred[14,29]      1 1000
## Captures.pred[14,30]      1 1000
## Captures.pred[14,31]      1  630
## Captures.pred[14,32]      1  850
## Captures.pred[14,33]      1 1000
## Captures.pred[14,34]      1 1000
## Captures.pred[14,35]      1 1000
## deviance                   1 1000
##
## For each parameter, n.eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor (at convergence, Rhat=1).
##
## DIC info (using the rule, pD = Dbar-Dhat)
## pD = 17.8 and DIC = 15320.1
## DIC is an estimate of expected predictive error (lower deviance is better).
```

```
#Rhat
summary(Rhat <- (ResulDef$summary[, "Rhat"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.9995  1.0005  1.0017  1.0028  1.0036  1.0479
```

```
summary(ResulDef$summary[, "n.eff"])
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   160.0   750.0  1000.0   859.1  1000.0  1000.0
```

```
ResulDef$DIC
```

```
## [1] 15320.1
```

```
ResulDef$pD
```

```
## [1] 17.753
```

```
#MSE  
mean(ResulDef$mean$resid)
```

```
## [1] 758.7237
```