

Chapter 1: Programming with R

The screenshot displays the RStudio interface with the following components:

- Source Editor:** Contains R code for creating vectors, a matrix, and a plot.
- Console:** Shows the execution output of the code, including the contents of the created objects.
- Environment Pane:** Lists the objects in the current environment and their values.
- Plots Pane:** Displays a scatter plot of the 'x' and 'y' variables.

```
4 integer_vector <- c(1L, 2L, 12L, 29L)
5 # To see contents of 'integer_vector'
6 integer_vector
7
8 logical_vector <- c(T, TRUE, F, FALSE)
9 # To see contents of 'logical_vector'
10 logical_vector
11
12 character_vector <- c("Apple", "Pear", "Red", "Green", "These are my favorite fruits and colors")
13 # To see content of character_vector
14
15 x <- seq(2, 25, by=1)
16 y <- x^2 + 3
17 plot(x, y)
18
19 values <- seq(1, 12, by=2)
20 values_matrix <- matrix(values, ncol=3, nrow=2)
```

Console Output:

```
> integer_vector
[1] 1 2 12 29
>
> logical_vector <- c(T, TRUE, F, FALSE)
> # To see contents of 'logical_vector'
> logical_vector
[1] TRUE TRUE FALSE FALSE
>
> character_vector <- c("Apple", "Pear", "Red", "Green", "These are my favorite fruits and colors")
> # To see content of character_vector
>
> x <- seq(2, 25, by=1)
> y <- x^2 + 3
> plot(x, y)
> values <- seq(1, 12, by=2)
> values_matrix <- matrix(values, ncol=3, nrow=2)
> View(values_matrix)
```

Environment Pane Data:

Object	Class	Value
values_matrix	num [1:2, 1:3]	1 3 5 7 9 11
character_vector	chr [1:5]	"Apple" "Pear" "Red" "Green" ...
integer_vector	int [1:4]	1 2 12 29
logical_vector	logi [1:4]	TRUE TRUE FALSE FALSE
values	num [1:6]	1 3 5 7 9 11
x	num [1:24]	2 3 4 5 6 7 8 9 10 11 ...
y	num [1:24]	7 12 19 28 39 52 67 84 103 124 ...

Plots Pane: A scatter plot showing the relationship between x and y. The x-axis ranges from 0 to 25, and the y-axis ranges from 0 to 500. The data points form a clear upward-curving parabolic shape, representing the equation $y = x^2 + 3$.

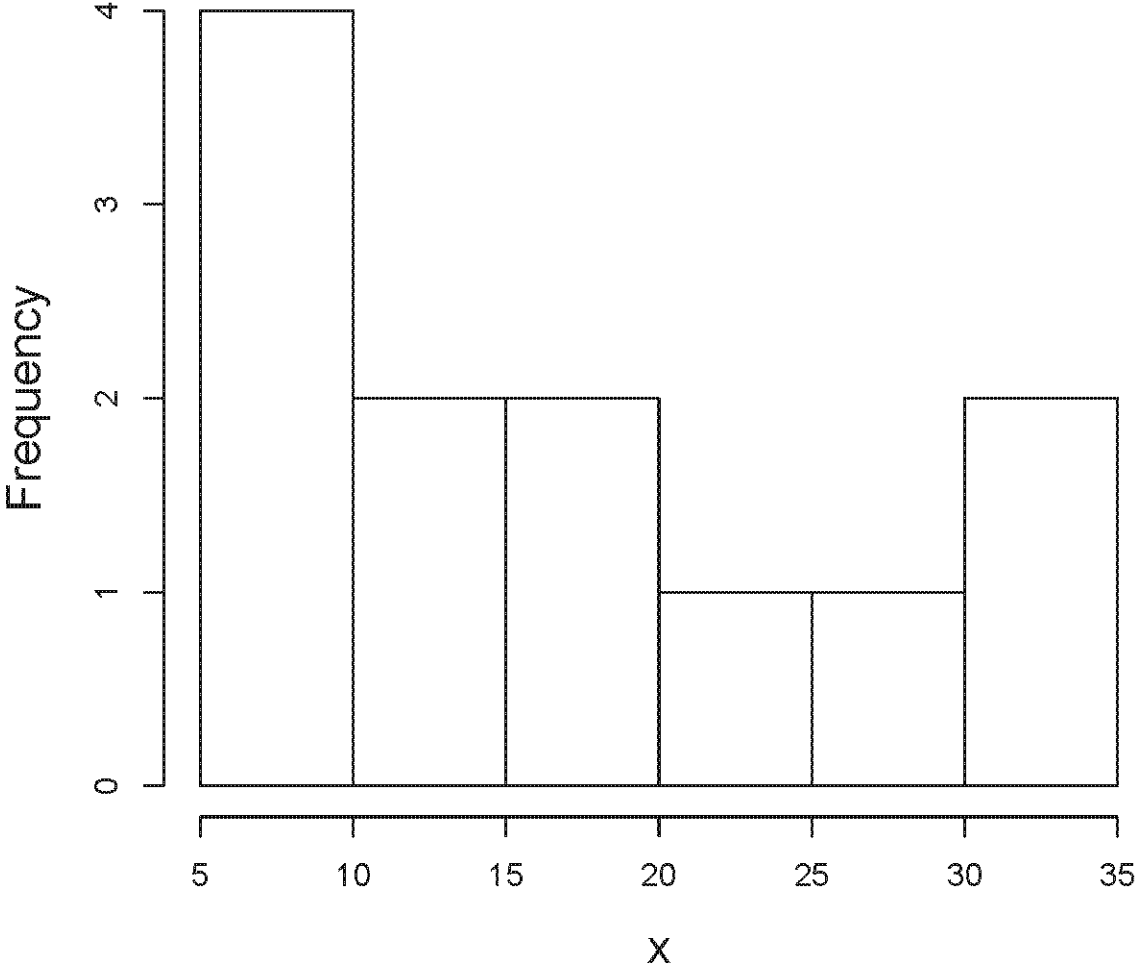
Most Flexible

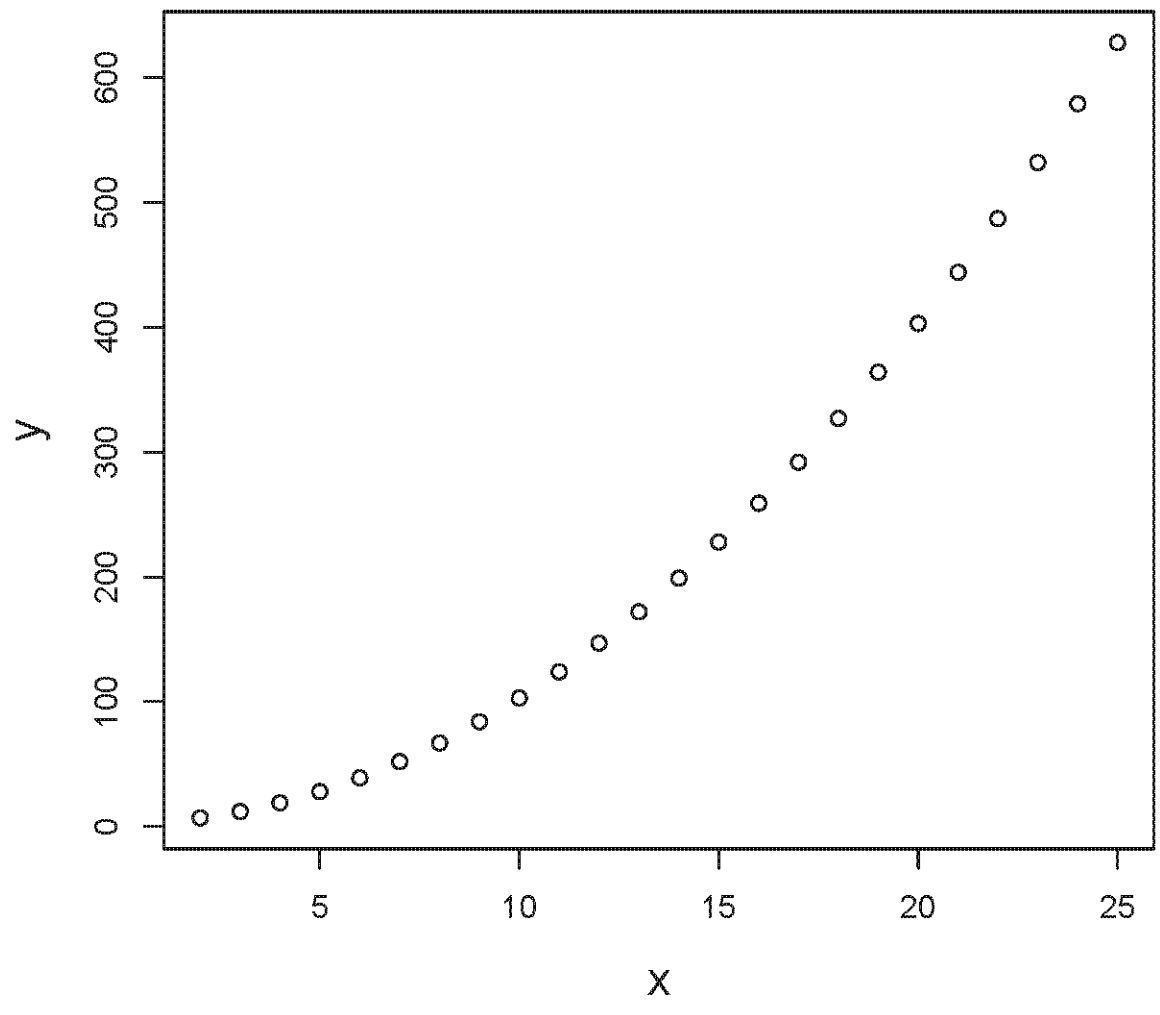


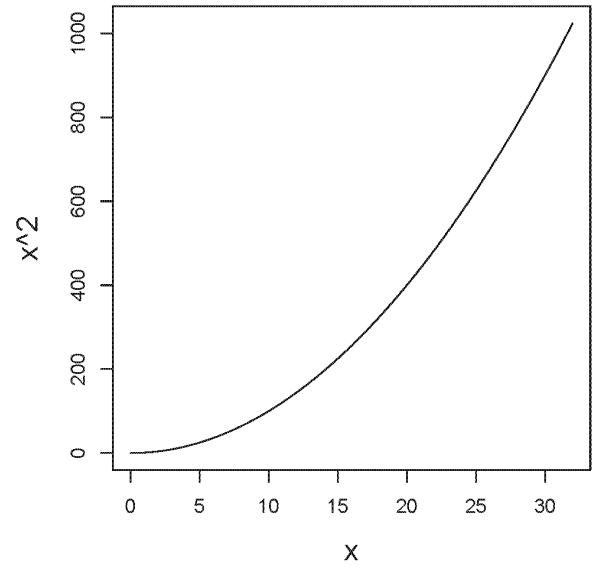
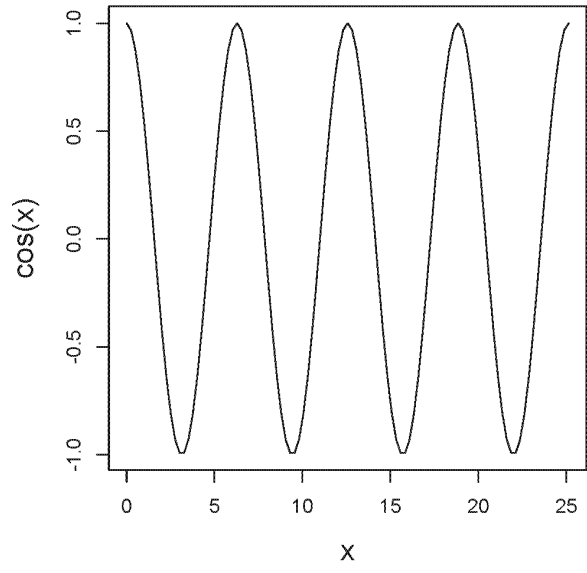
Least Flexible

Character → numeric → integer → logical

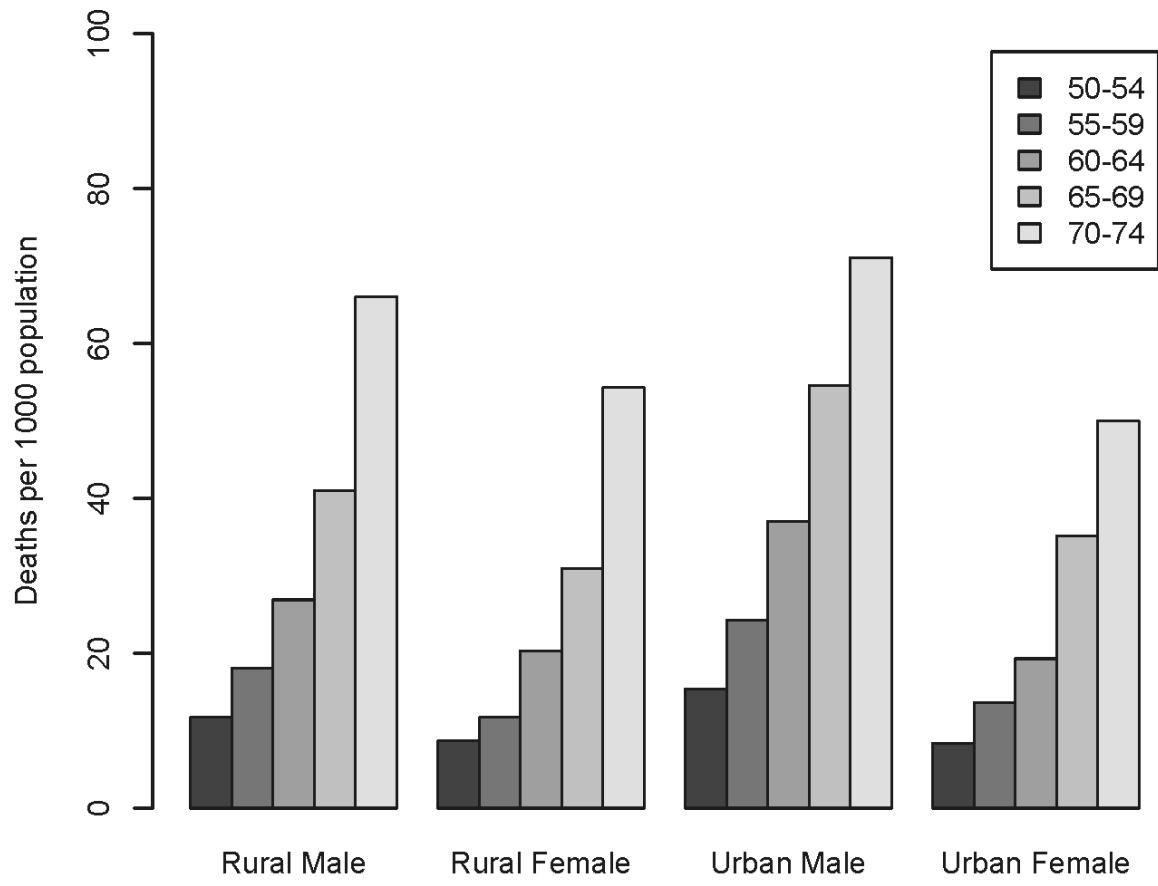
Histogram of x

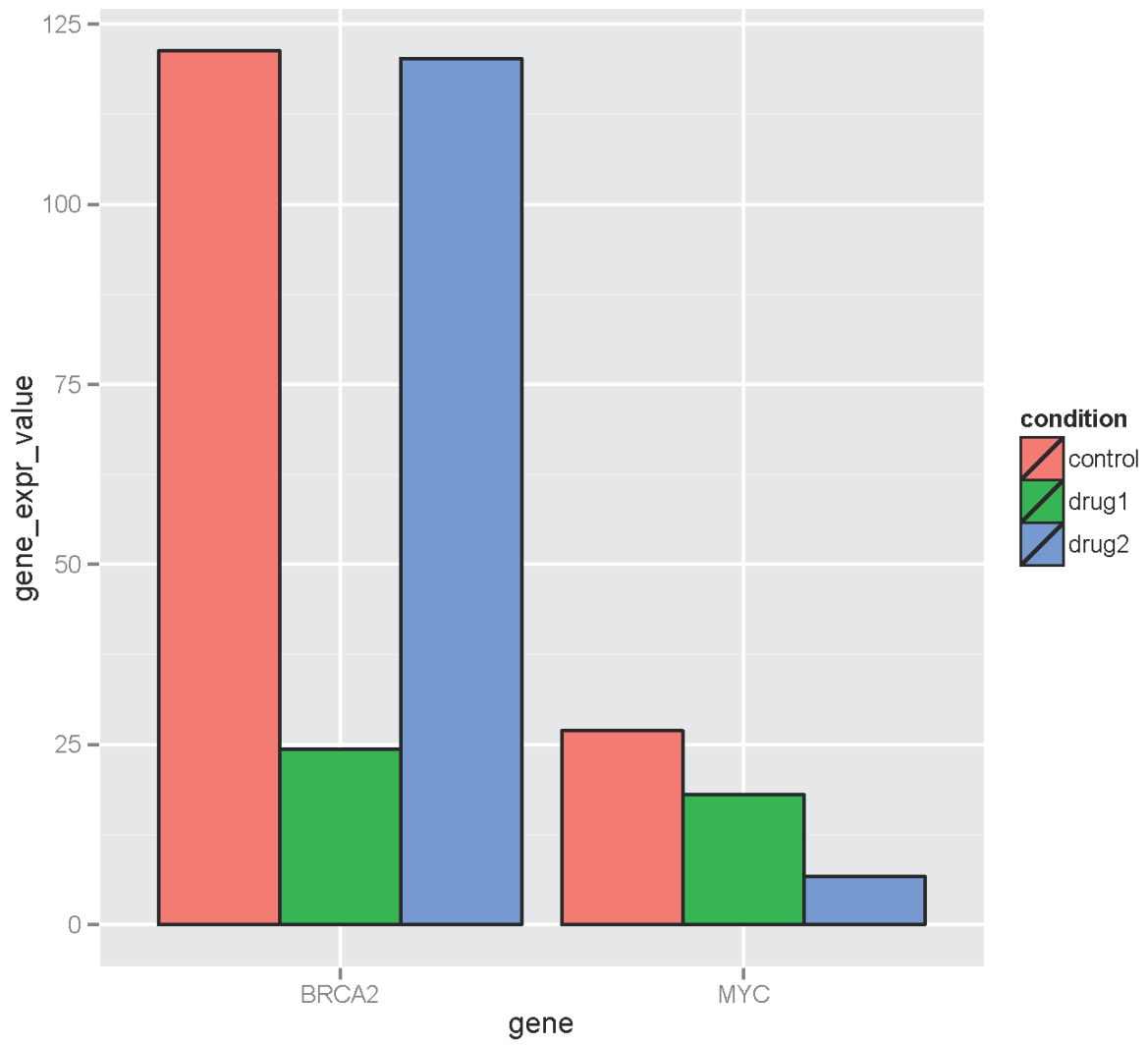


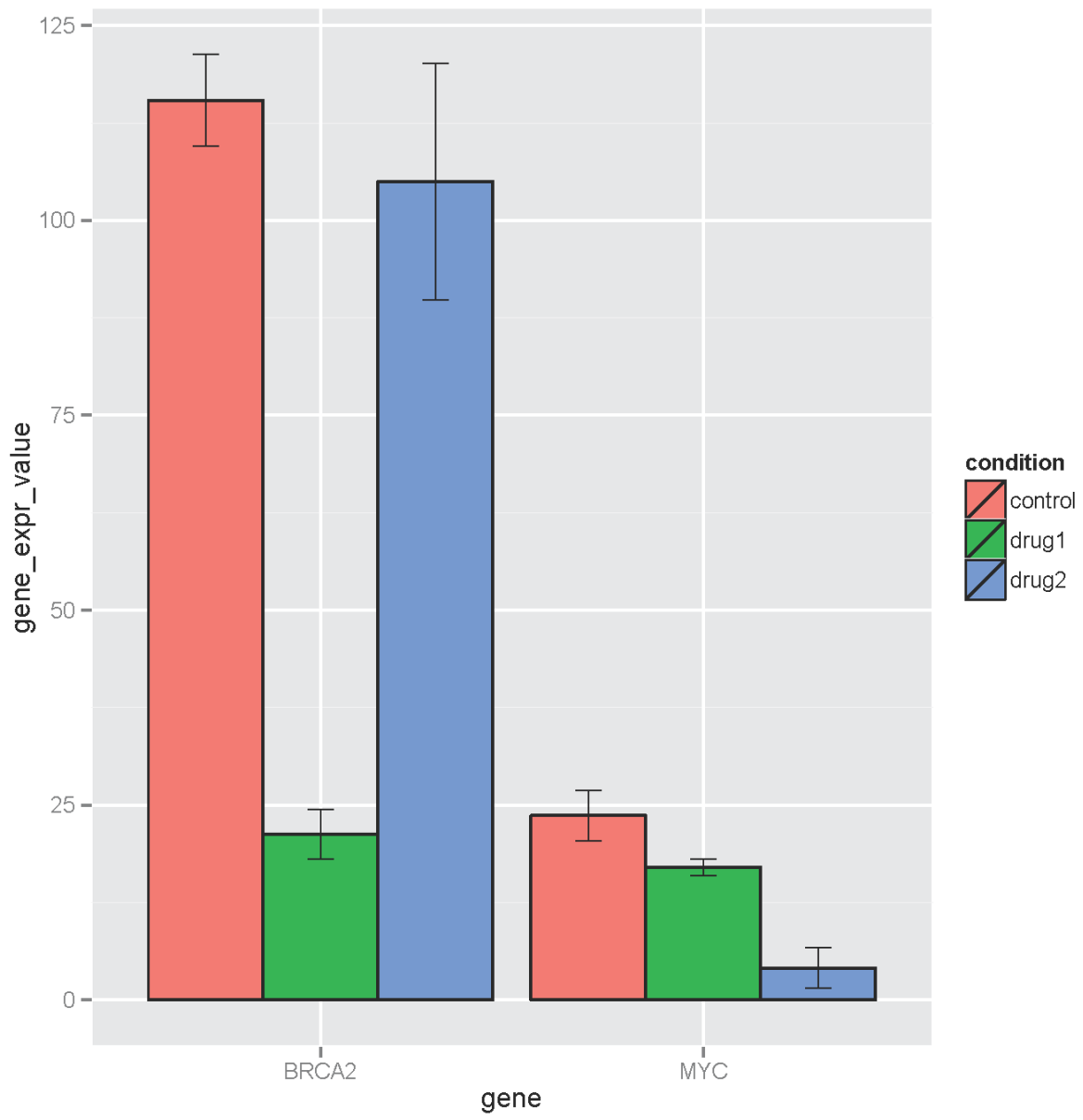




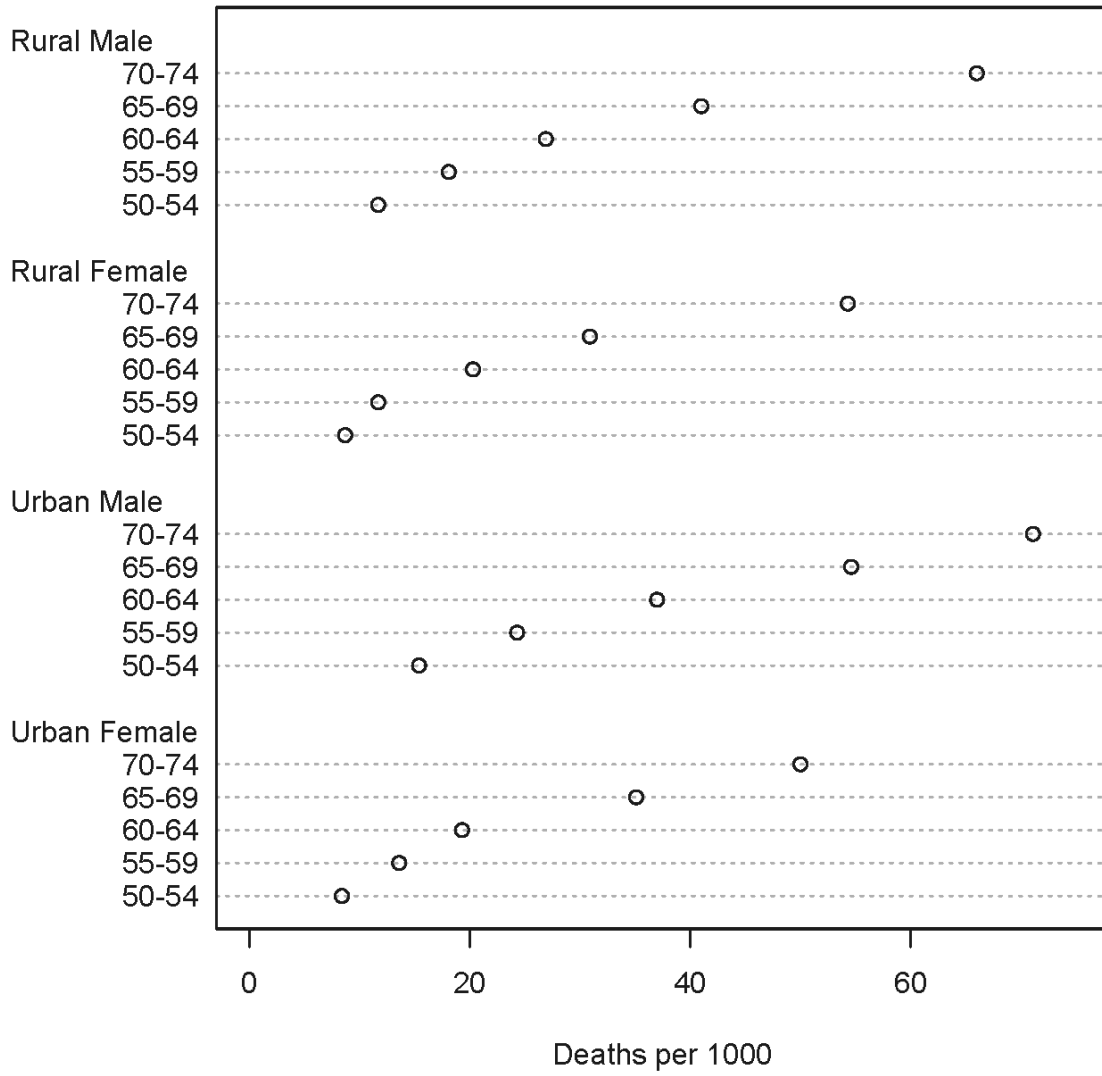
Death rate in VA



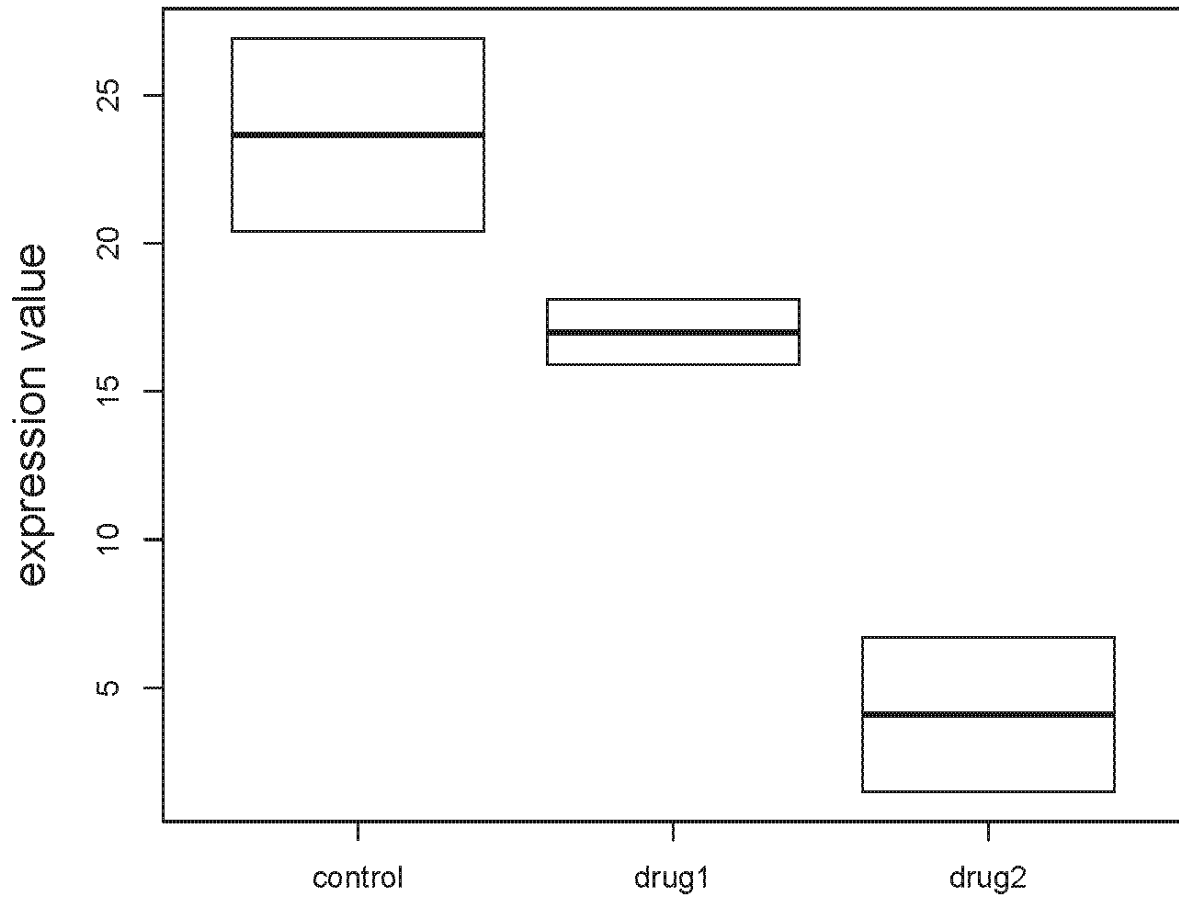


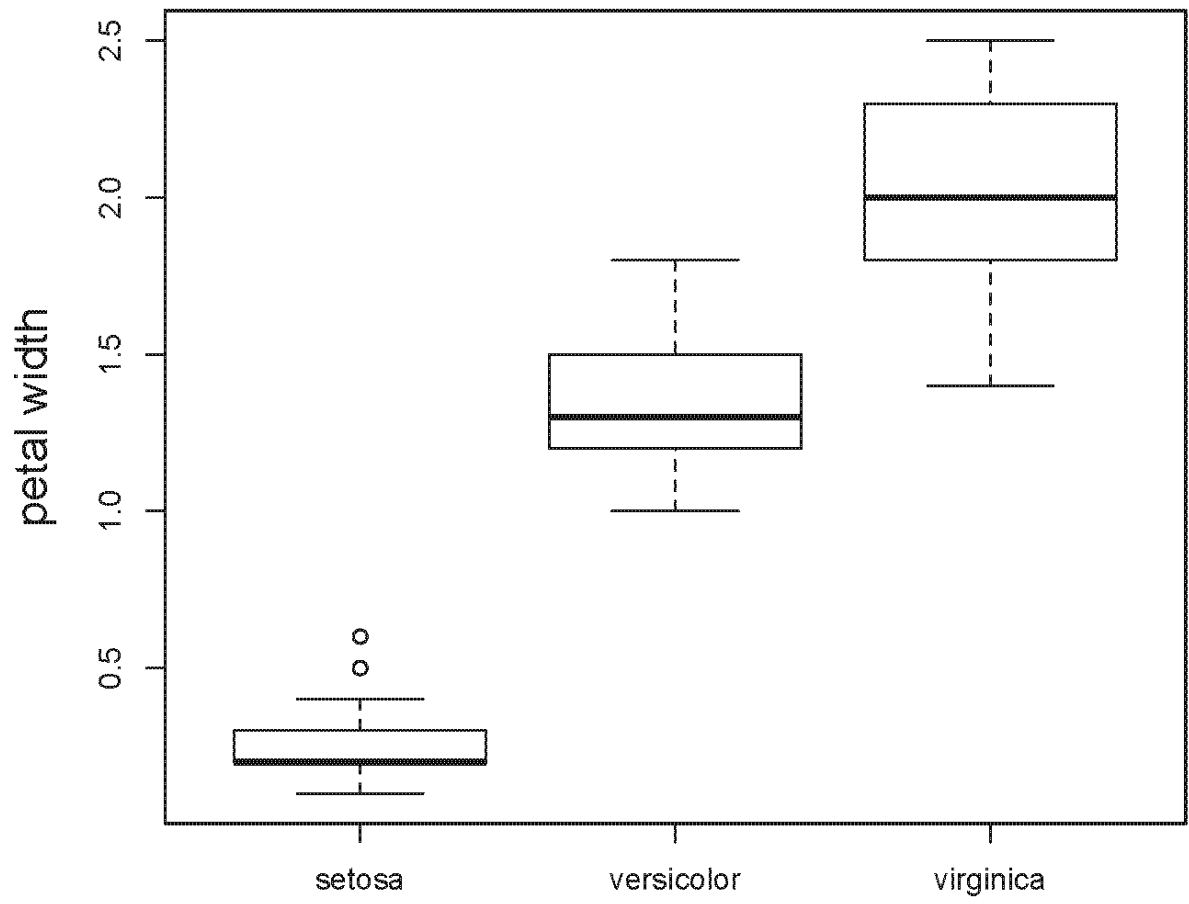


Death rates in VA



MYC Expression by Condition

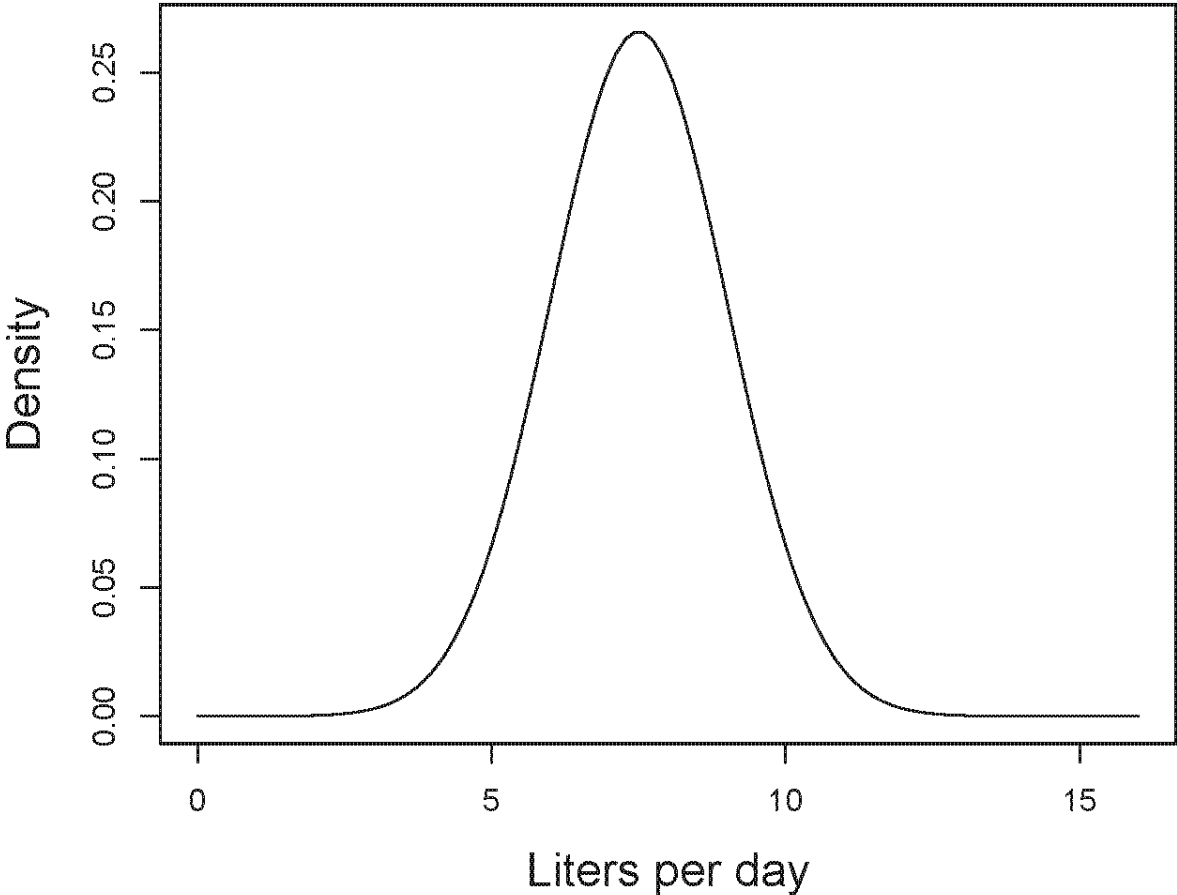


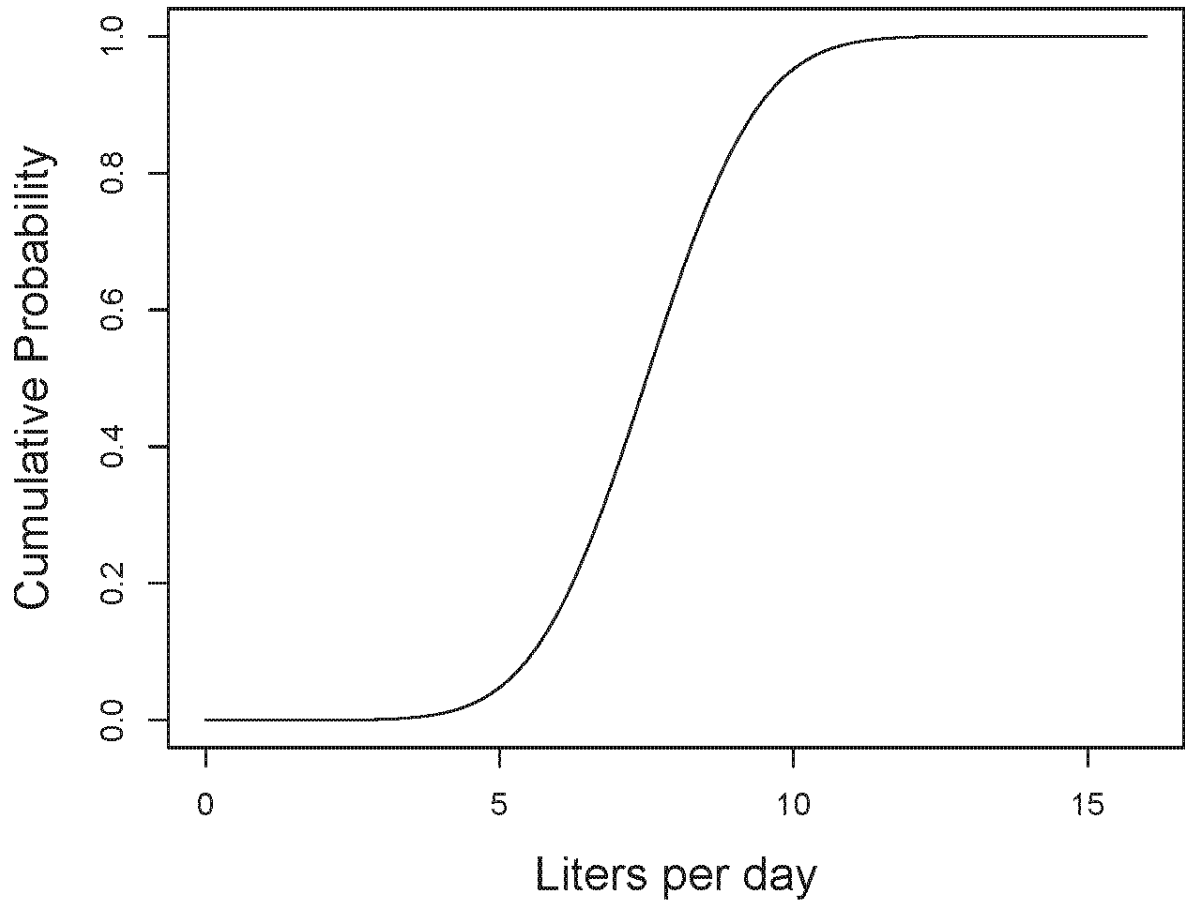


Help topics matching 'mean'		
Topic	Package	Description
DateTimeClasses	base	Date-Time Classes
Date	base	Date Class
colSums	base	Form Row and Column Sums and Means
difftime	base	Time Intervals
mean	base	Arithmetic Mean
sunspot	boot	Annual Mean Sunspot Numbers
runmean	caTools	Mean of a Moving Window
dates	chron	Generate Dates and Times Components from Input
CKME	clue	Cassini Data Partitions Obtained by K-Means
meanabsdev	cluster	Internal cluster functions
effectiveSize	coda	Effective sample size for estimating the mean
dispersionPlot	cummeRbund	Mean count vs dispersion plot
IDate	data.table	Integer based date class
fitDispersionFunction	DEXSeq	Fit the mean-variance function.
dglmStdResid	edgeR	Visualize the mean-variance relationship in DGE data using standardized residuals
loessByCol	edgeR	Locally Weighted Mean By Column
binMeanVar	edgeR	Explore the mean-variance relationship for DGE data
ghMoments	fBasics	Generalized Hyperbolic Distribution Moments
ghtMoments	fBasics	Generalized Hyperbolic Student-t Moments
hypMoments	fBasics	Hyperbolic Distribution Moments
nigMoments	fBasics	Moments for the Normal Inverse Gaussian
%in%-methods	flowCore	Filter-specific membership methods
kmeansFilter	flowCore	Class "kmeansFilter"

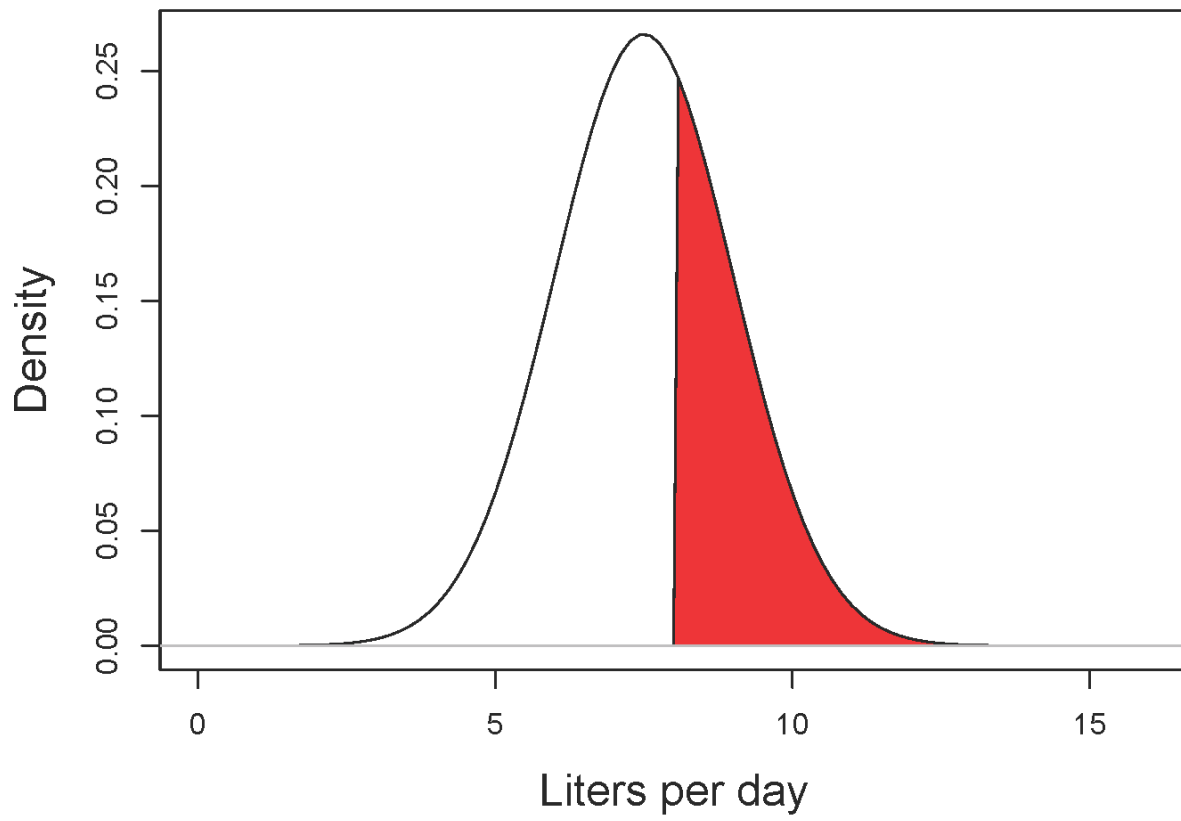
Chapter 2: Statistical Methods with R

Liters of water drank by school children < 12 years old

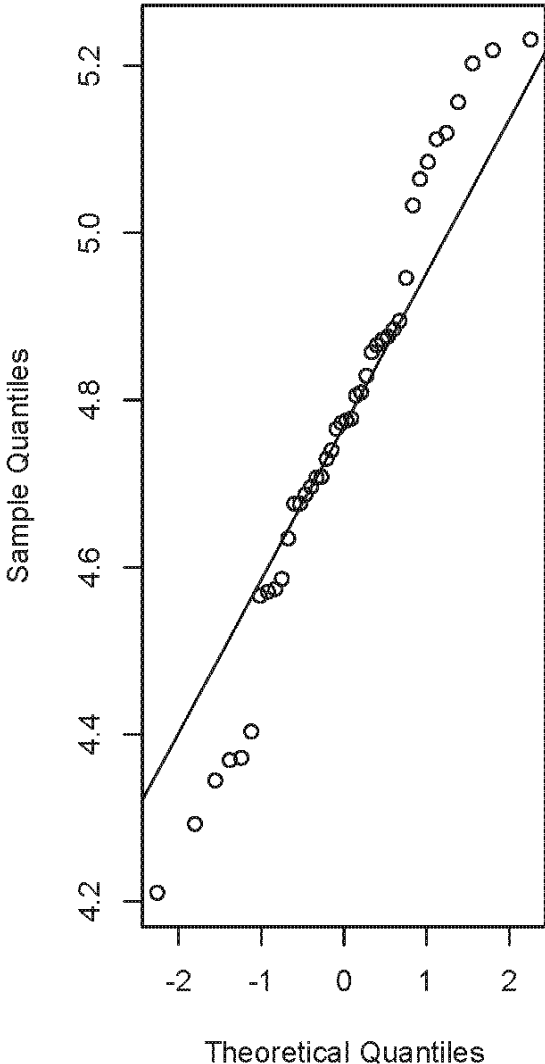




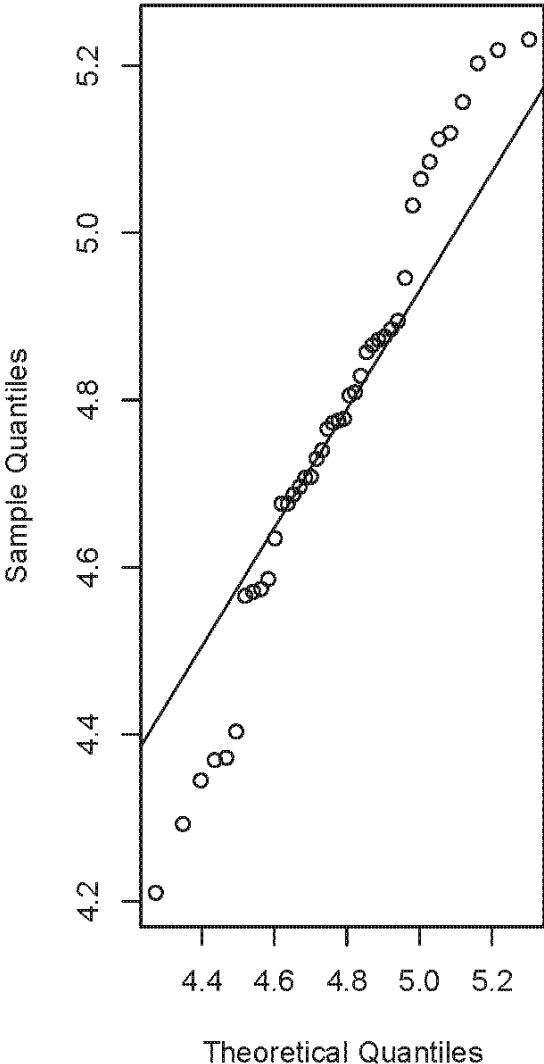
Cumulative probability of a child drinking > 8L/day: 0.37



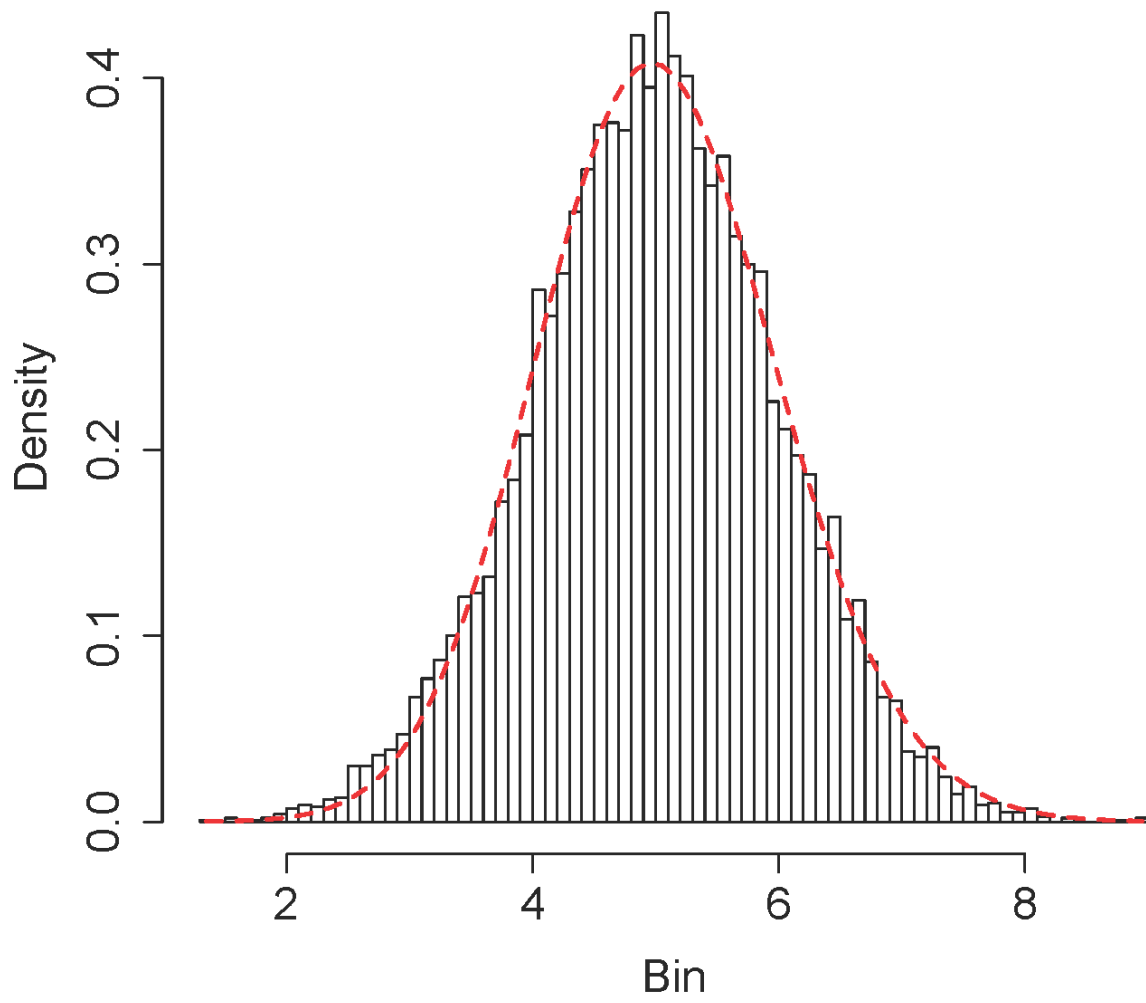
Normal Q-Q Plot



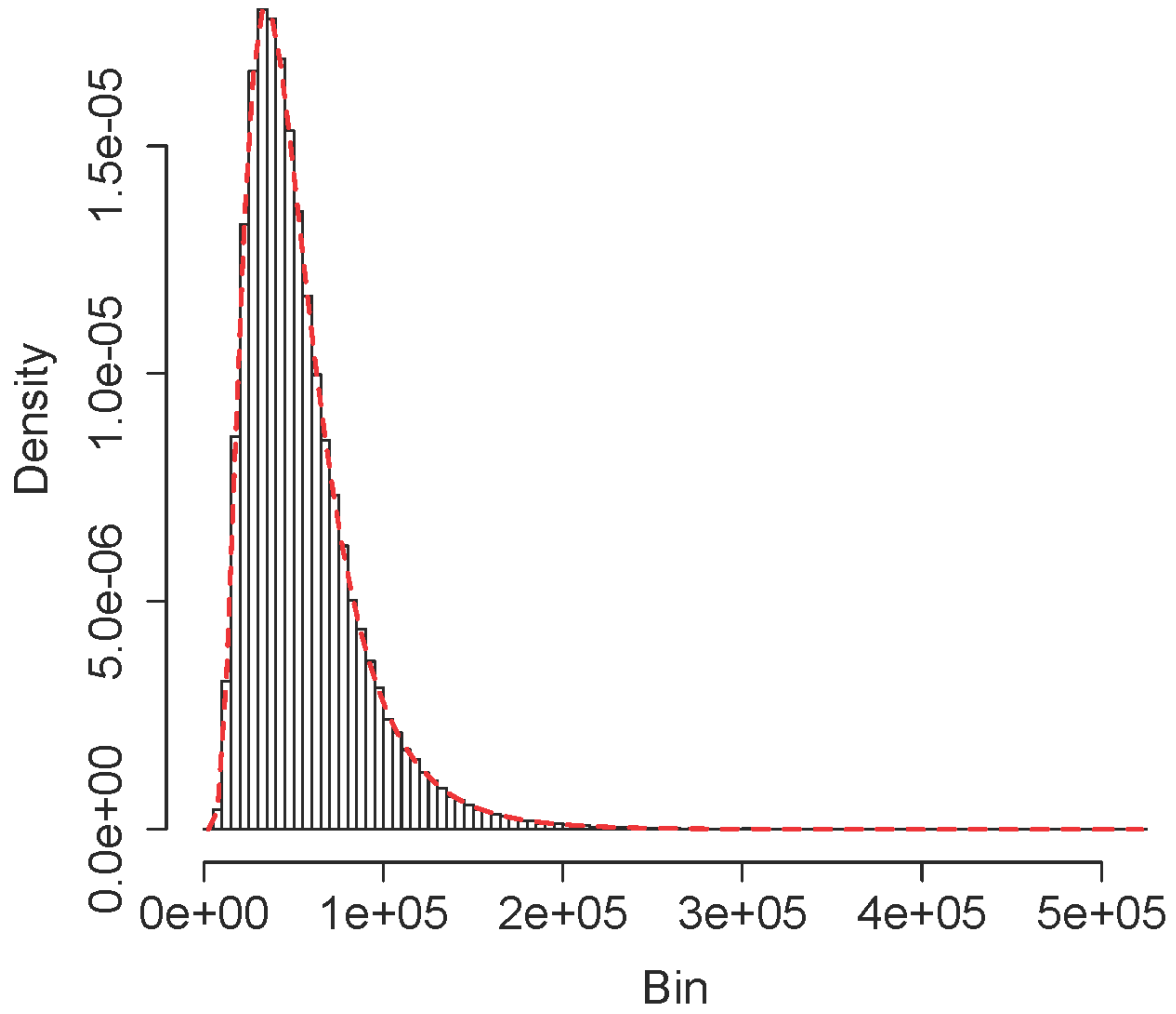
Gamma QQ-plot



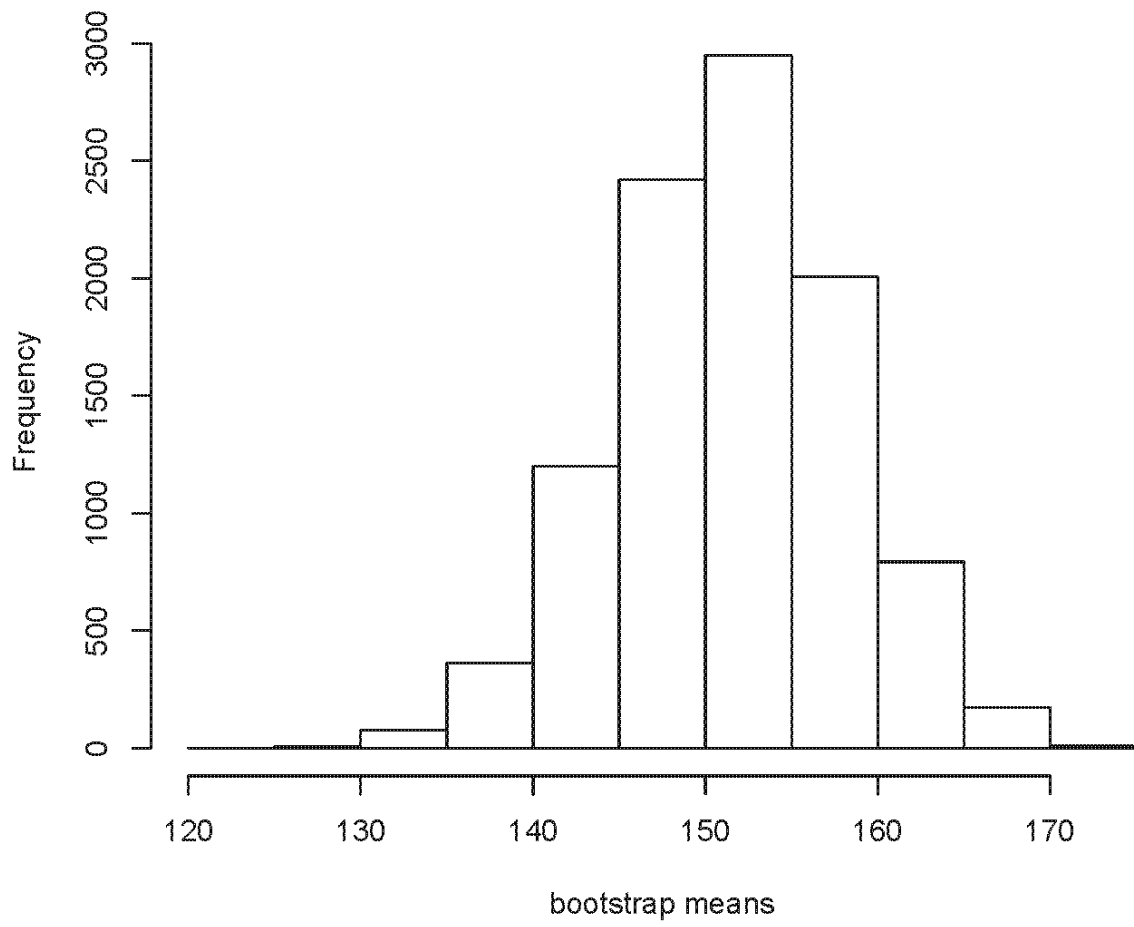
Skewed-normal distribution, AIC = -1056.764



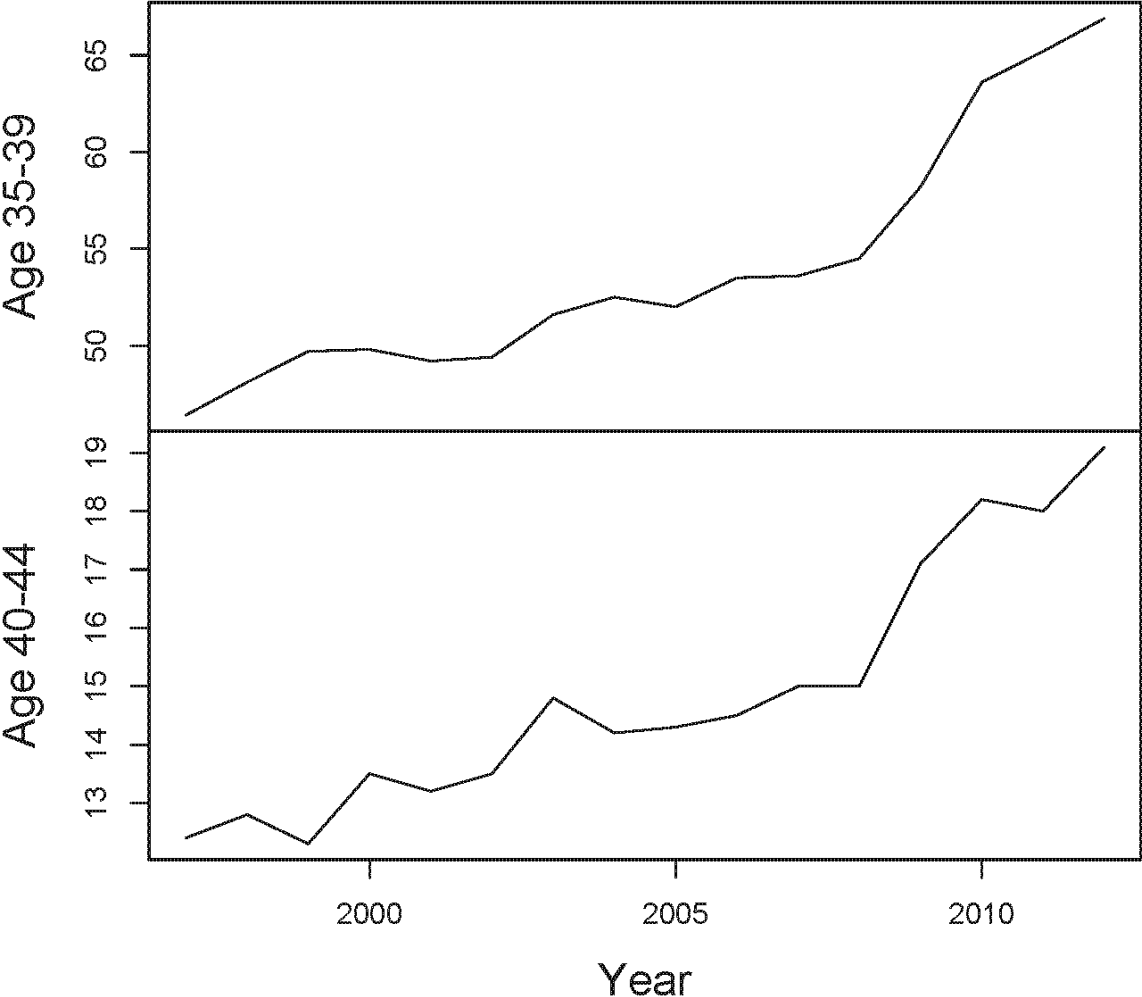
Generalized normal distribution, AIC = -6682.103



Histogram of f

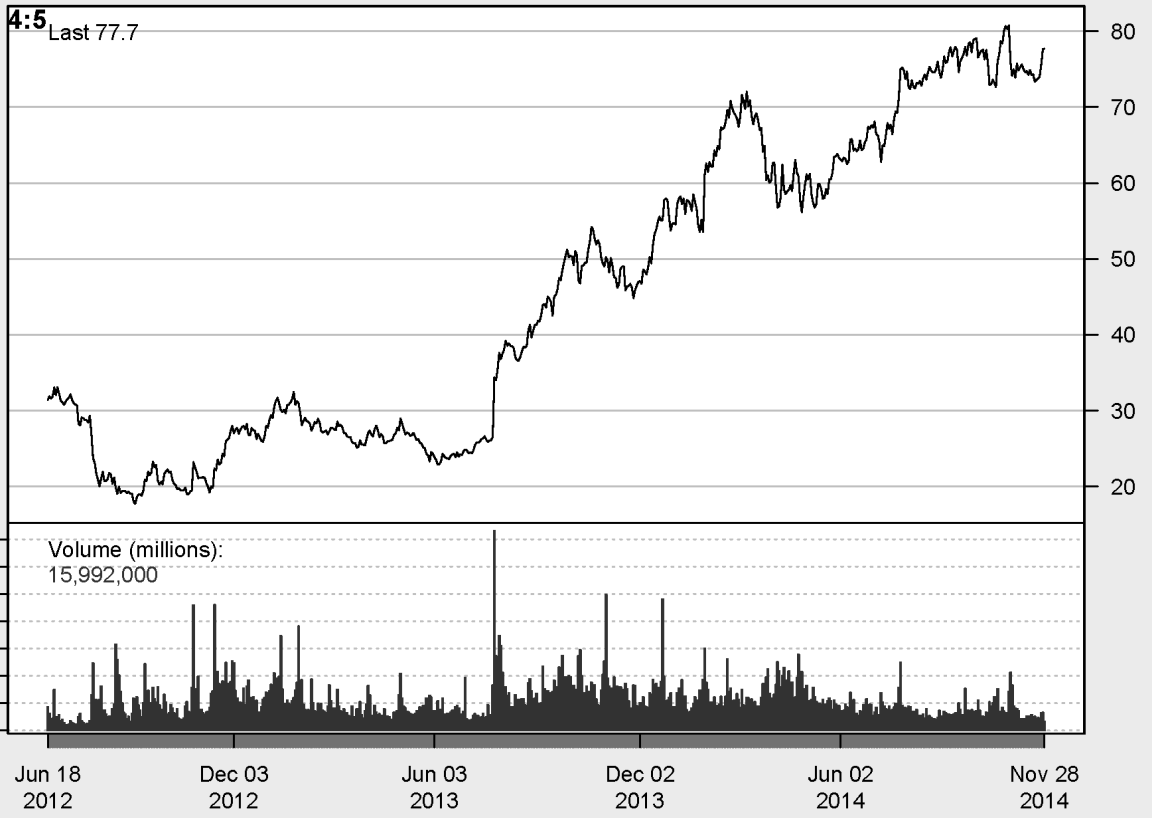


Fertility Rates for Females in NYC from 1997 to 2012

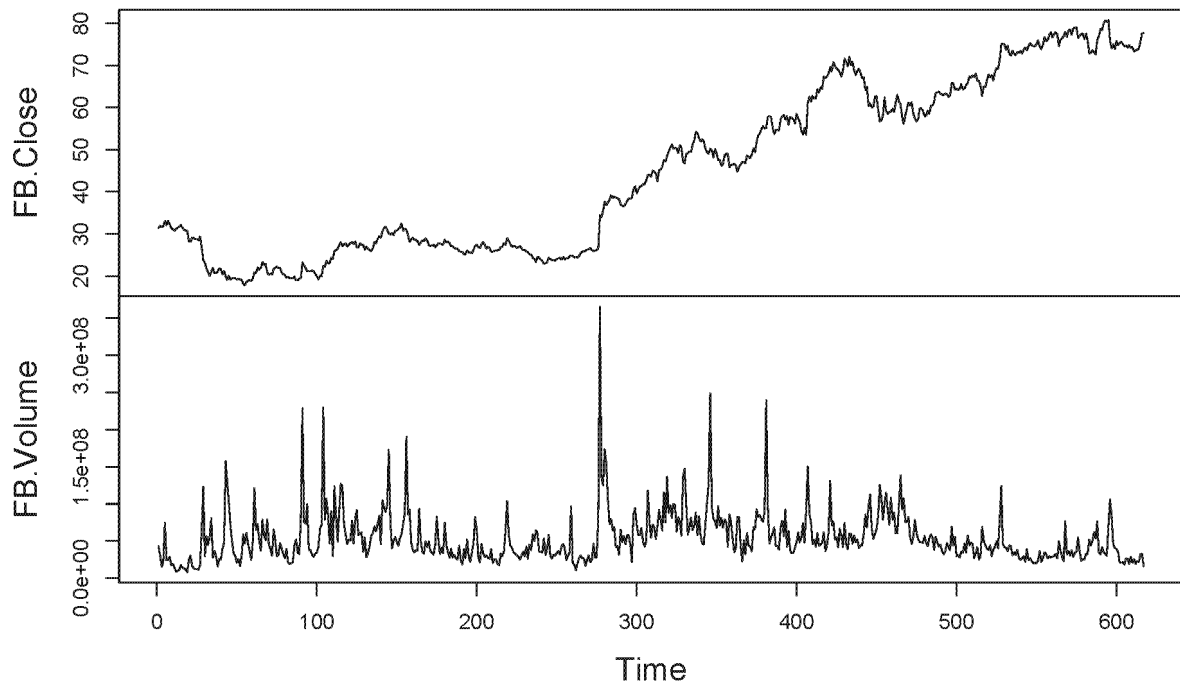


fbstock

[2012-06-18/2014-11-28]

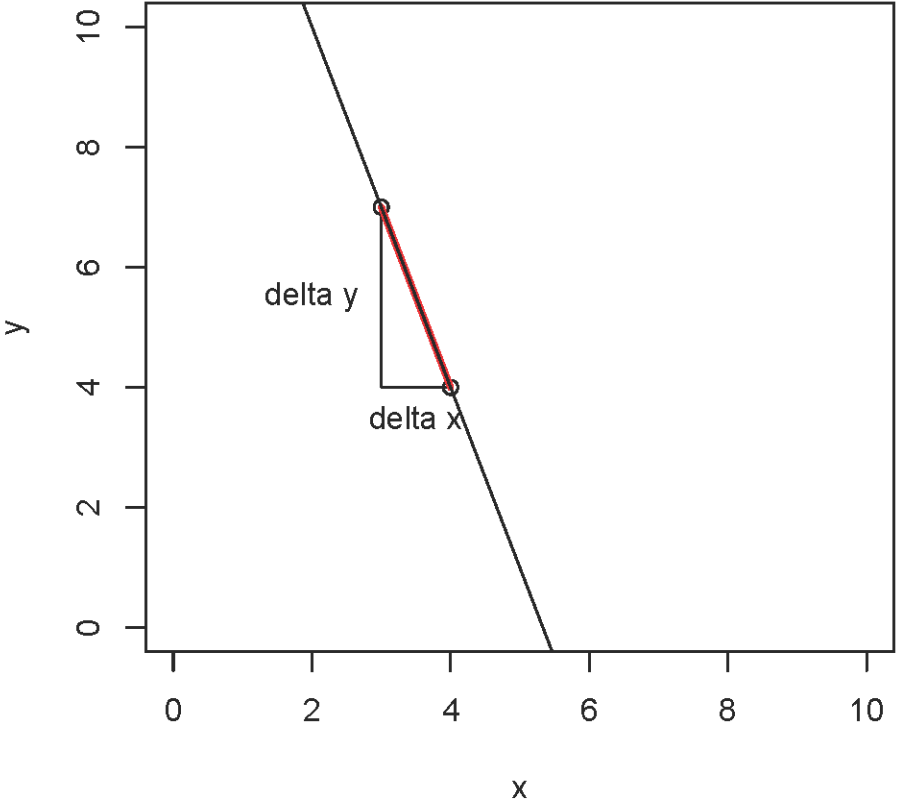


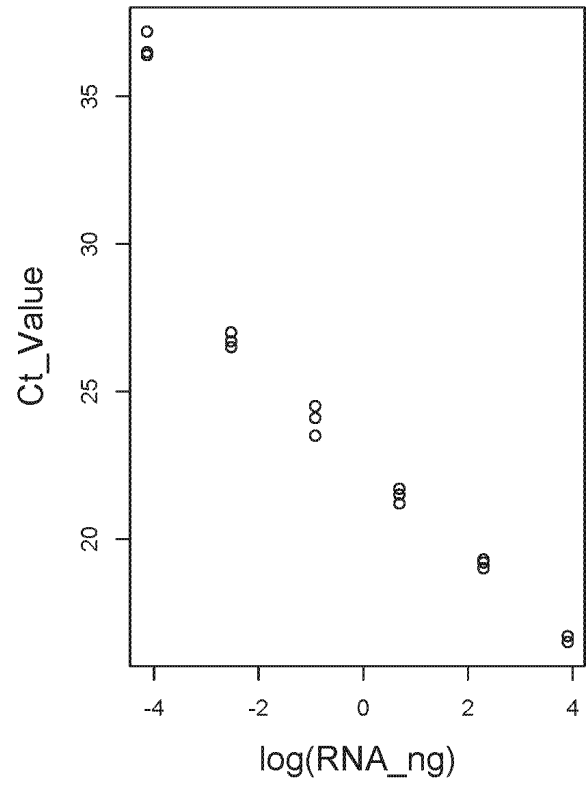
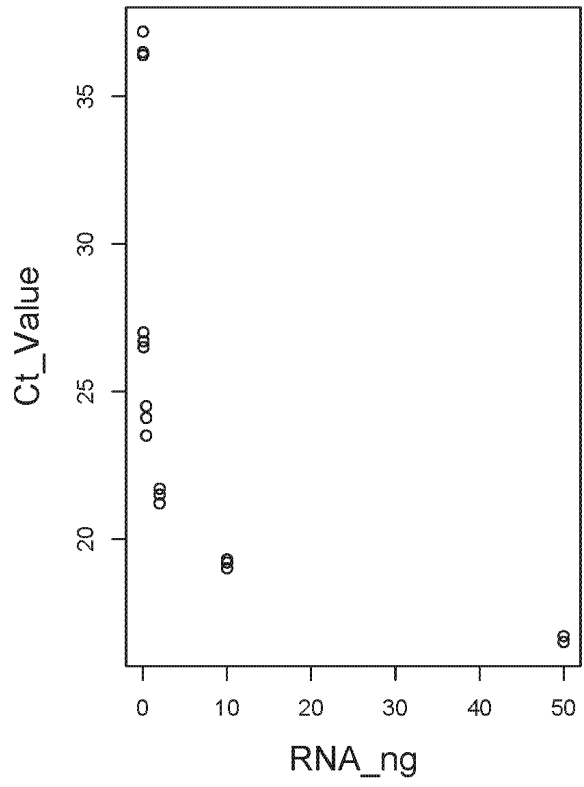
FACEBOOK Stock Information from 2012-06-18 to 2014-11-28

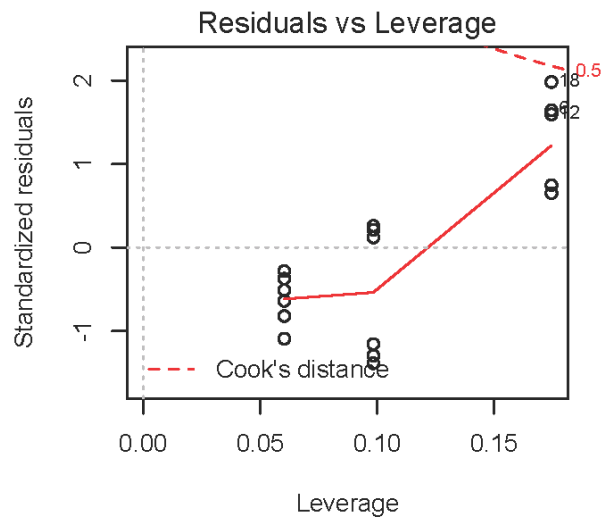
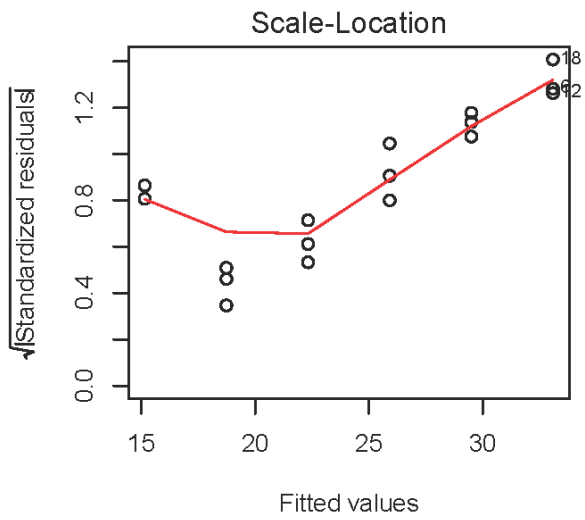
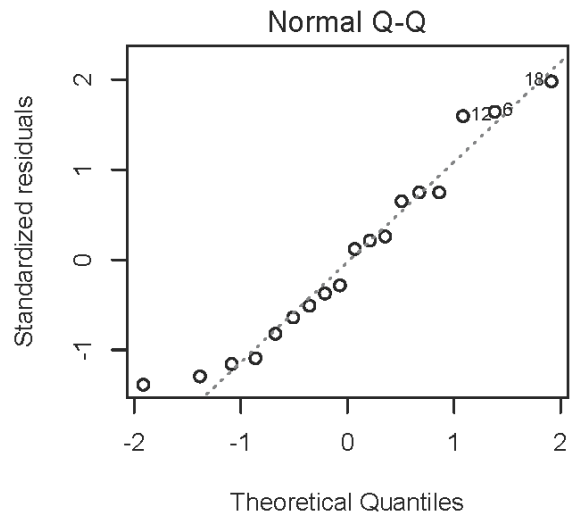
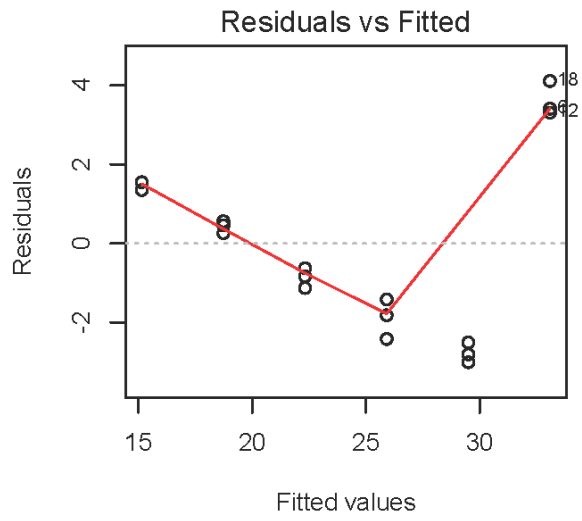


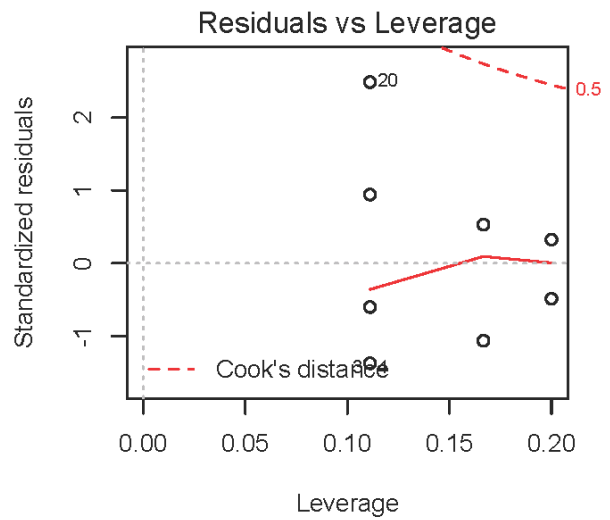
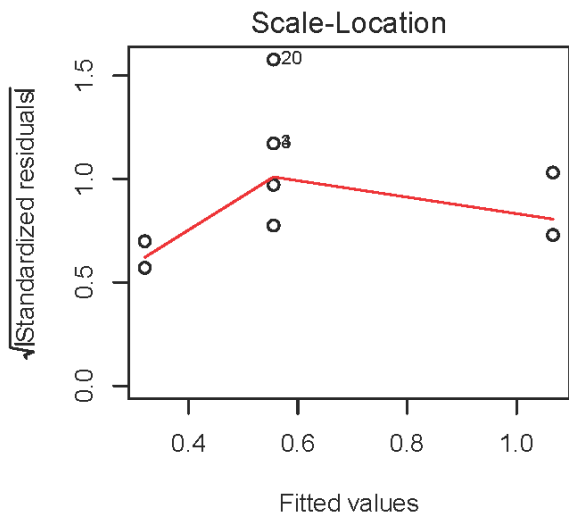
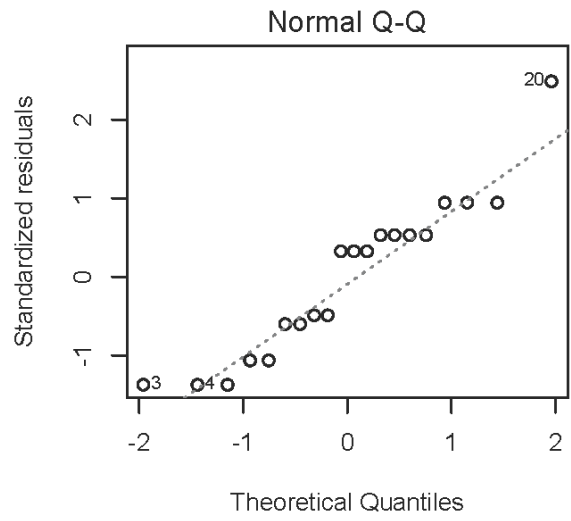
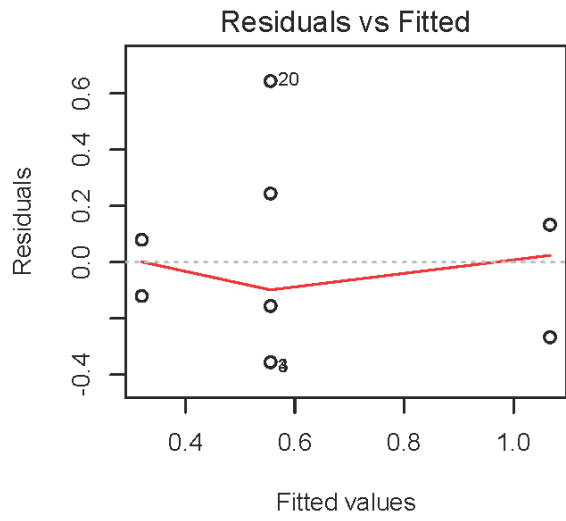
Chapter 3: Linear Models

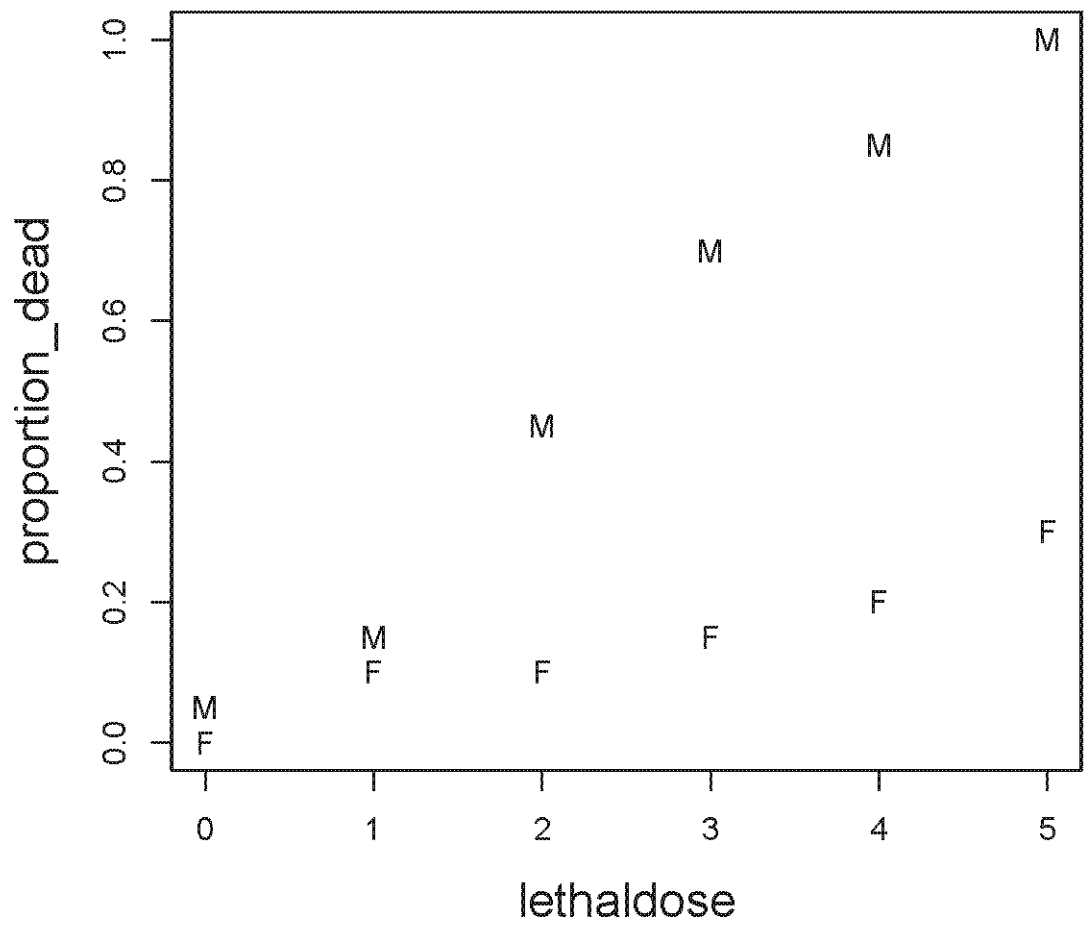
Slope from coordinates (3,7) and (10,4)

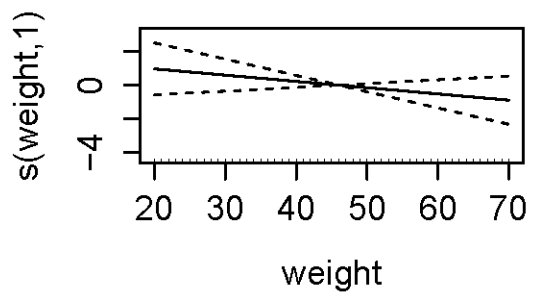
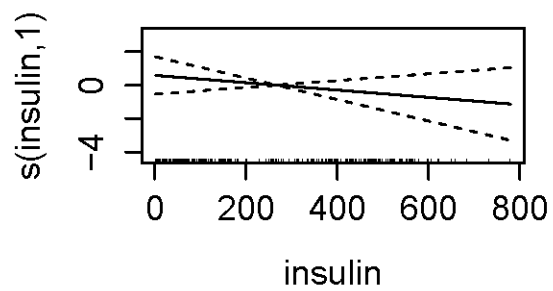
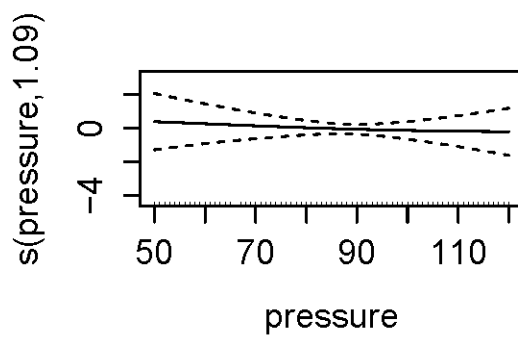
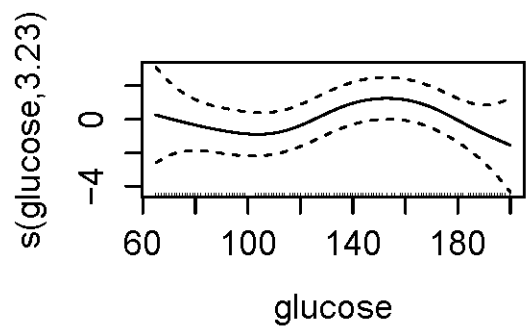


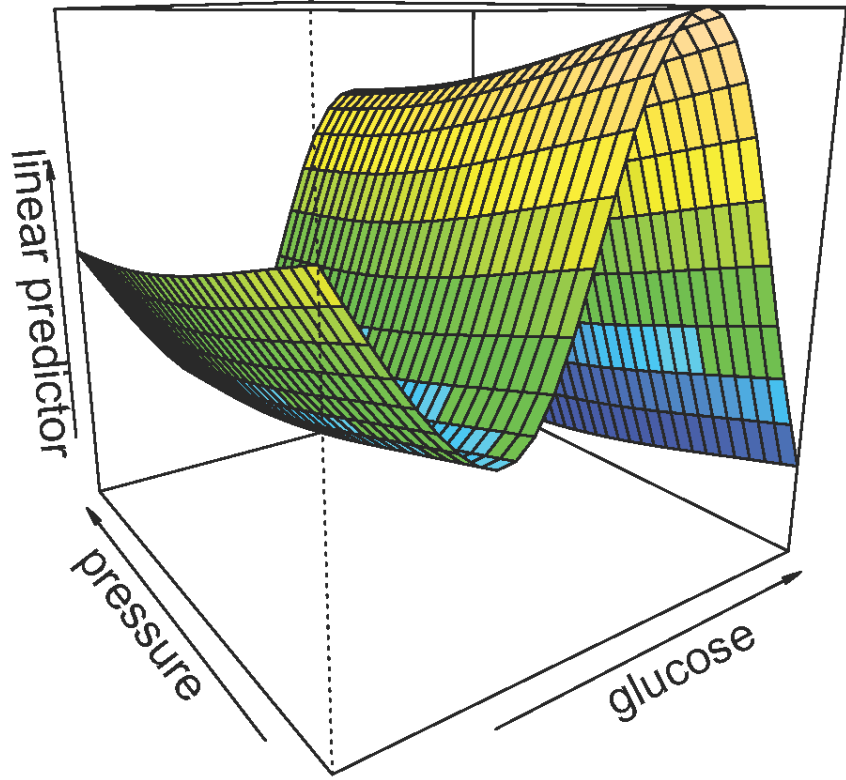


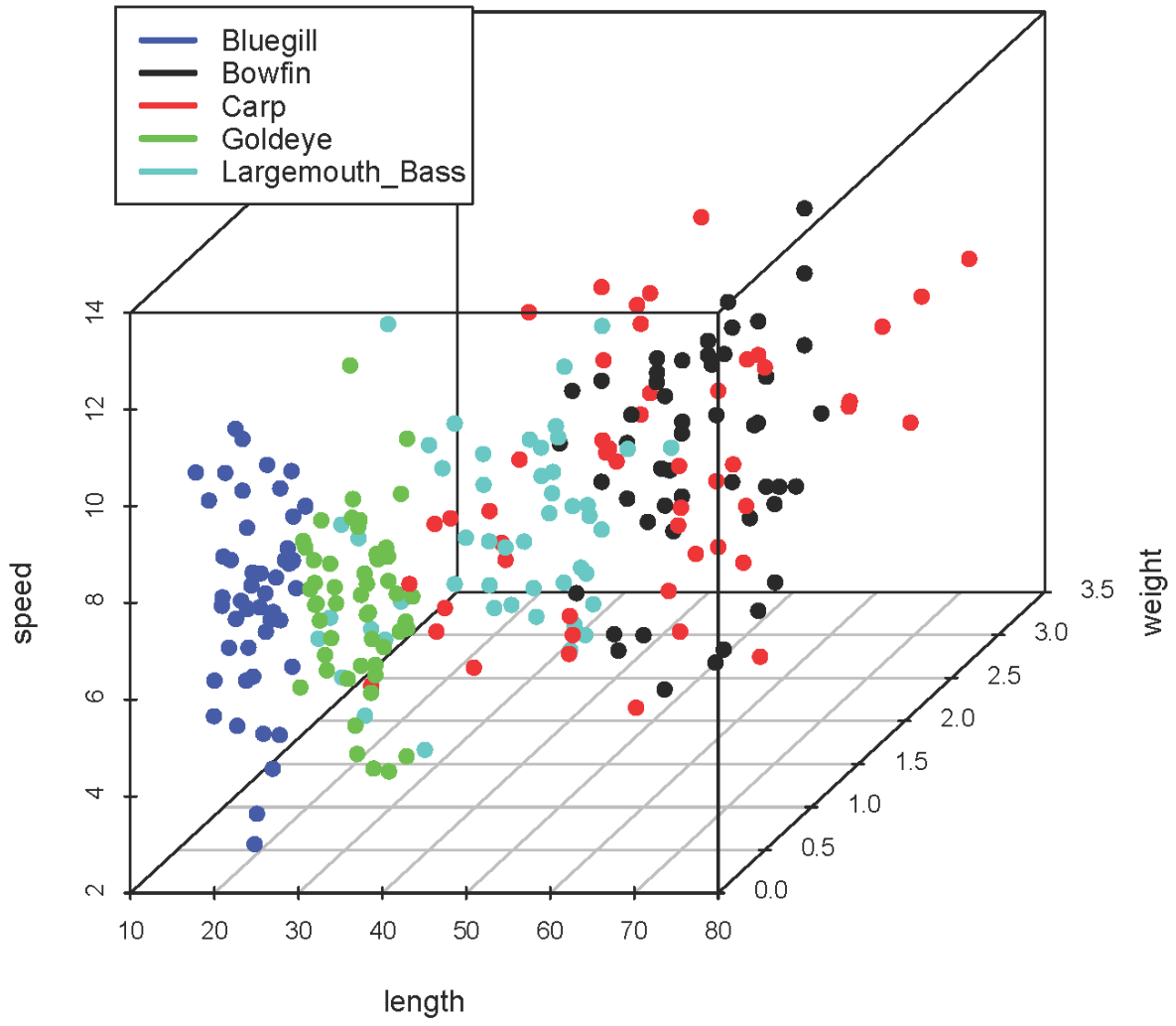


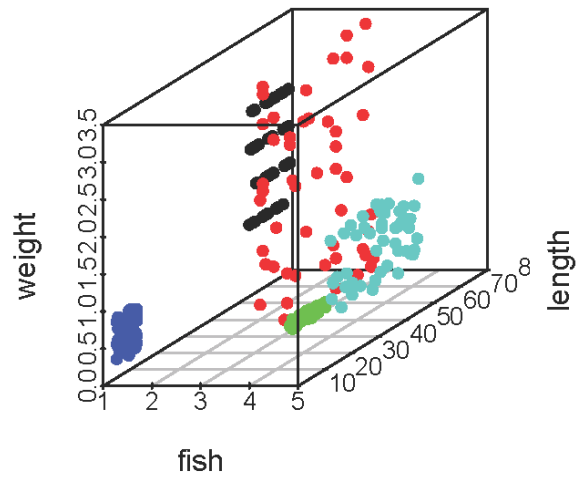
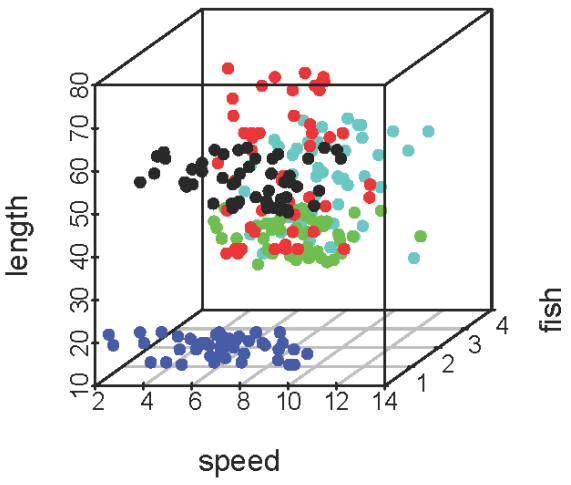
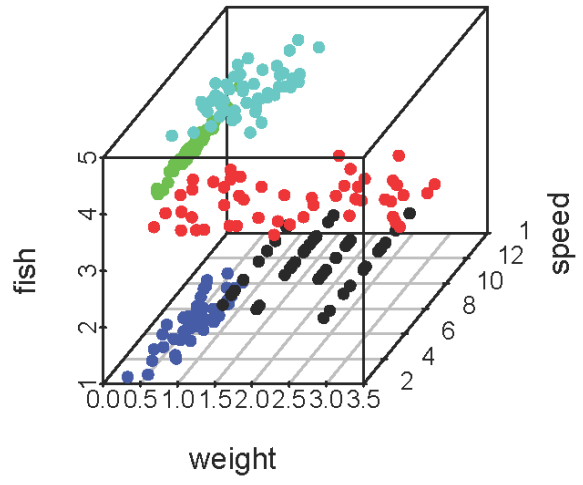
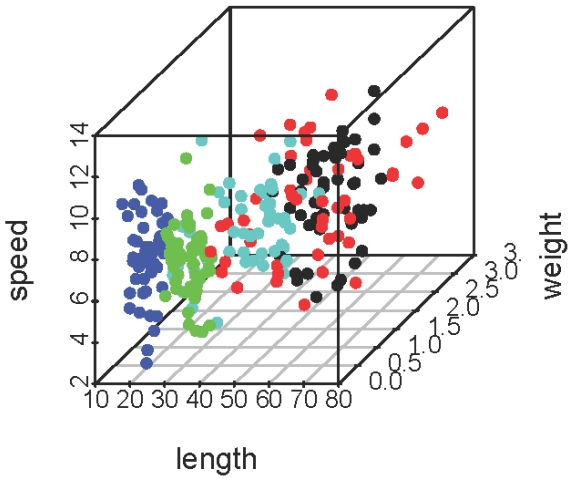




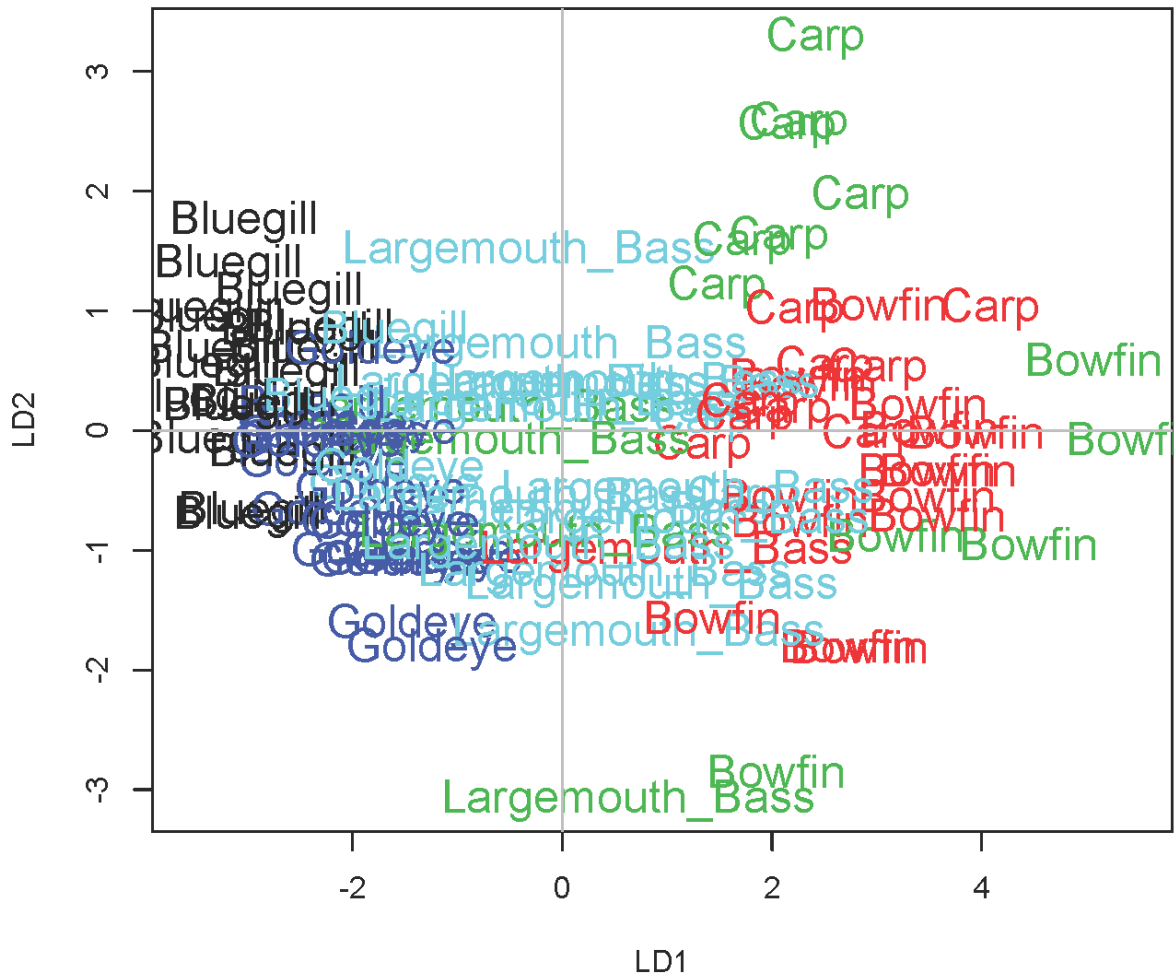


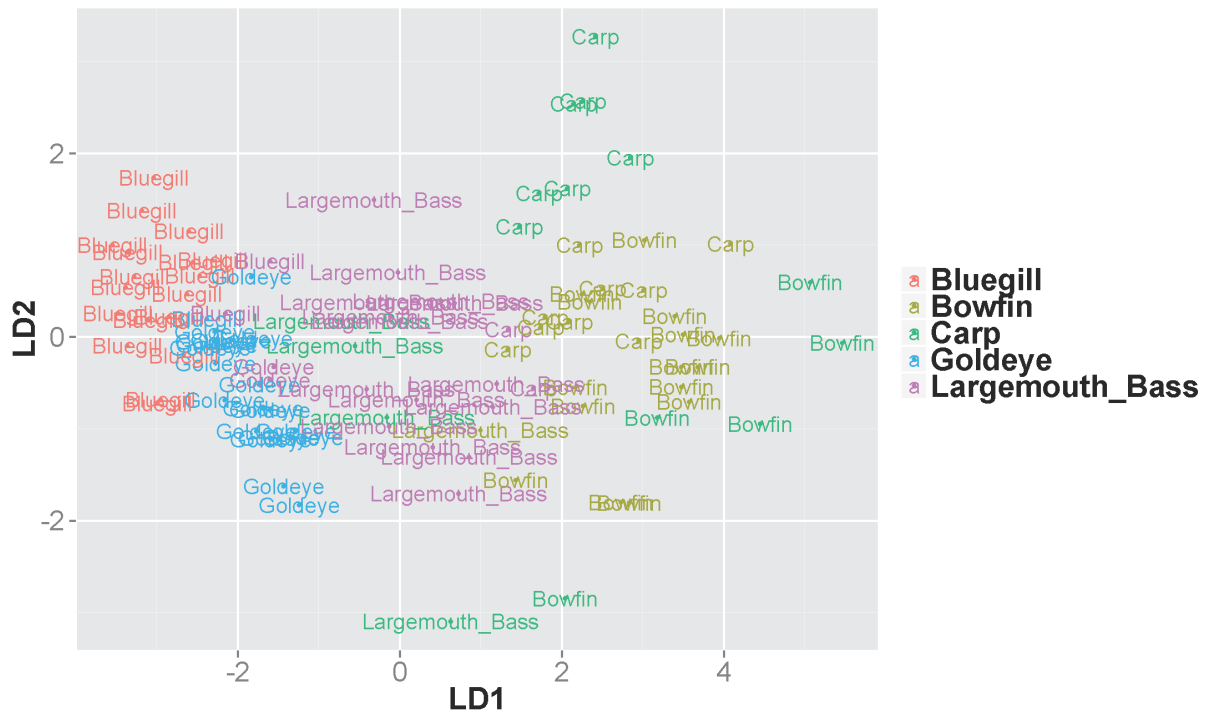


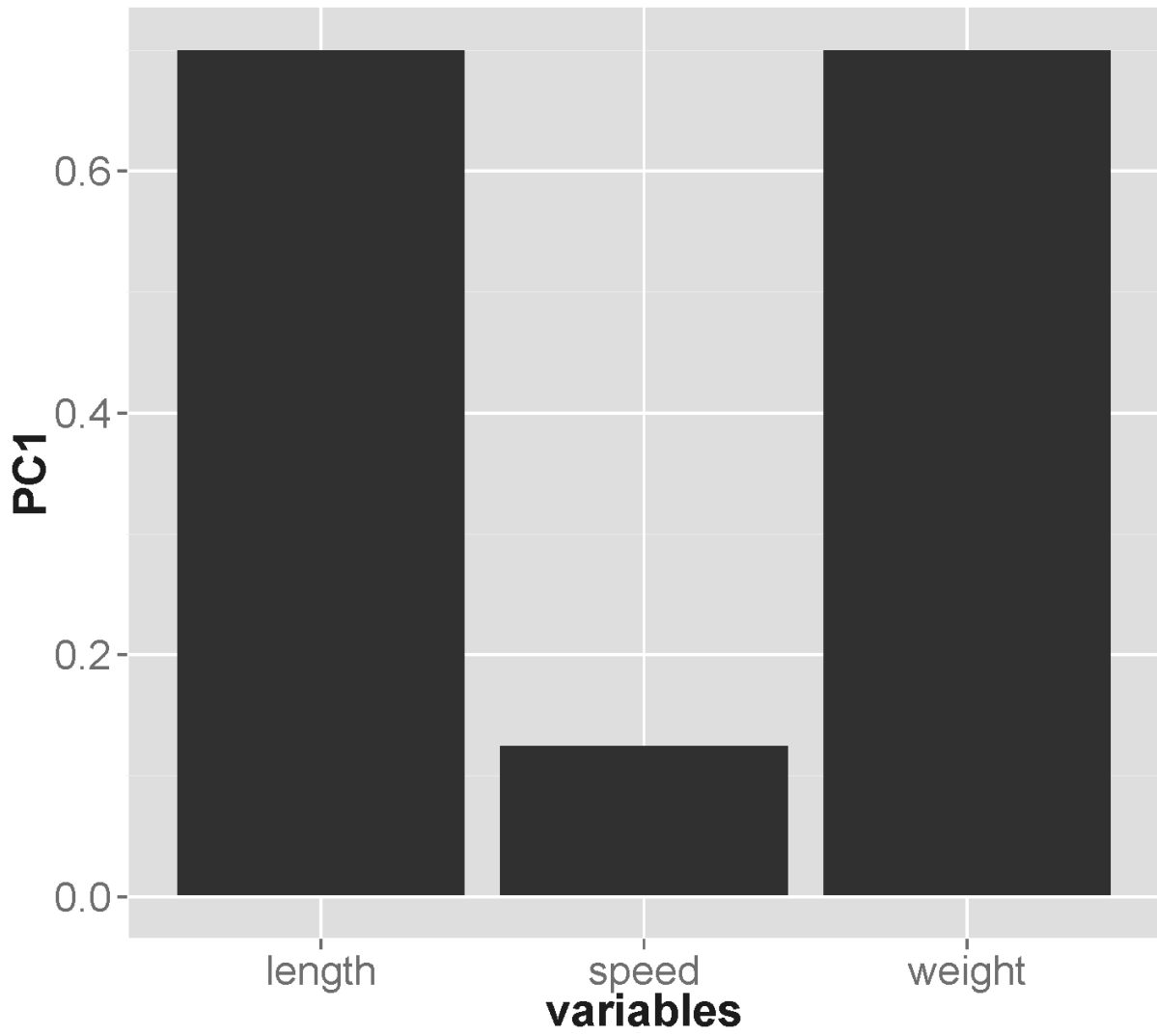


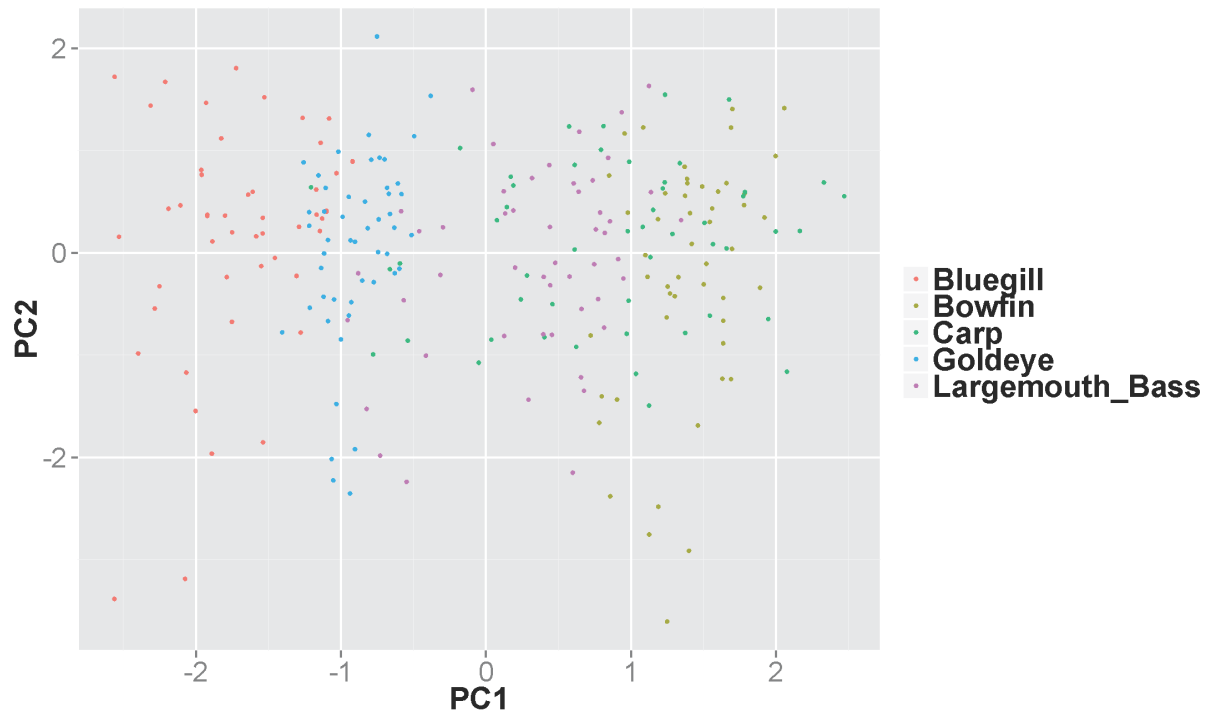


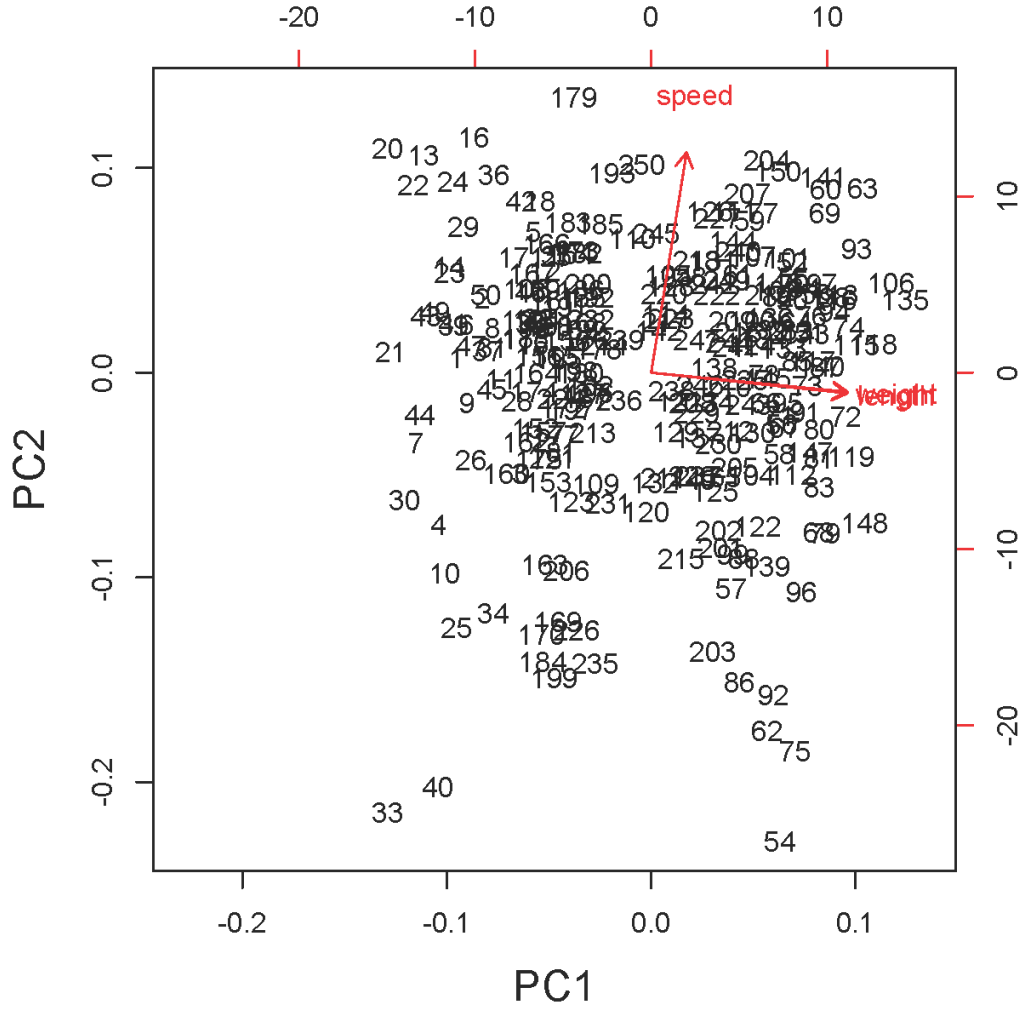
TrainingSetLDA Results(n=50)



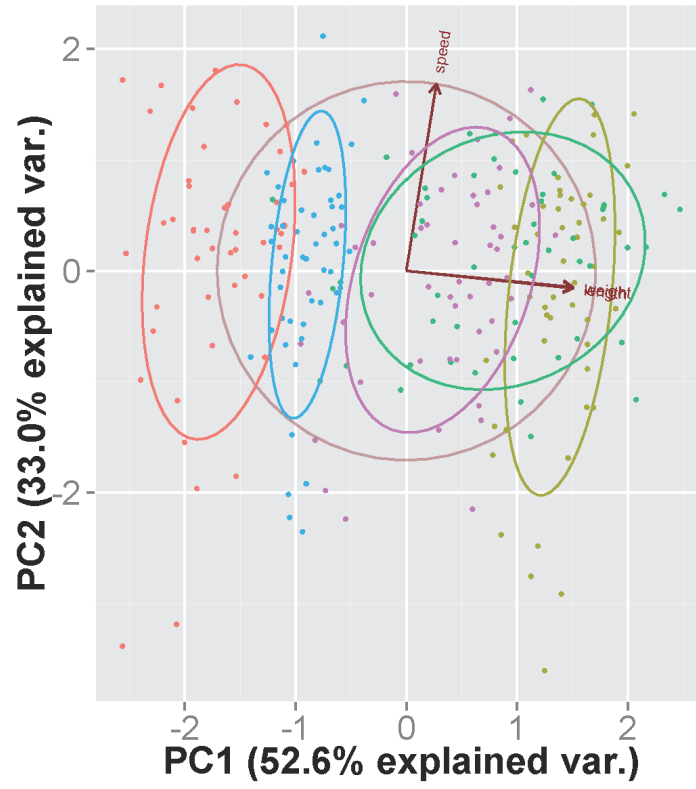


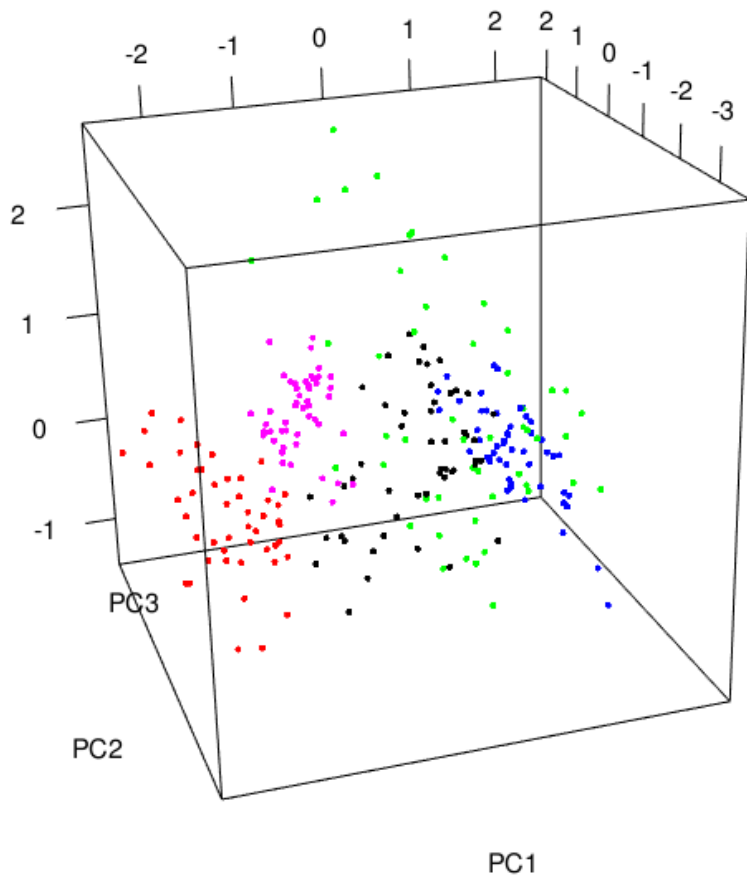


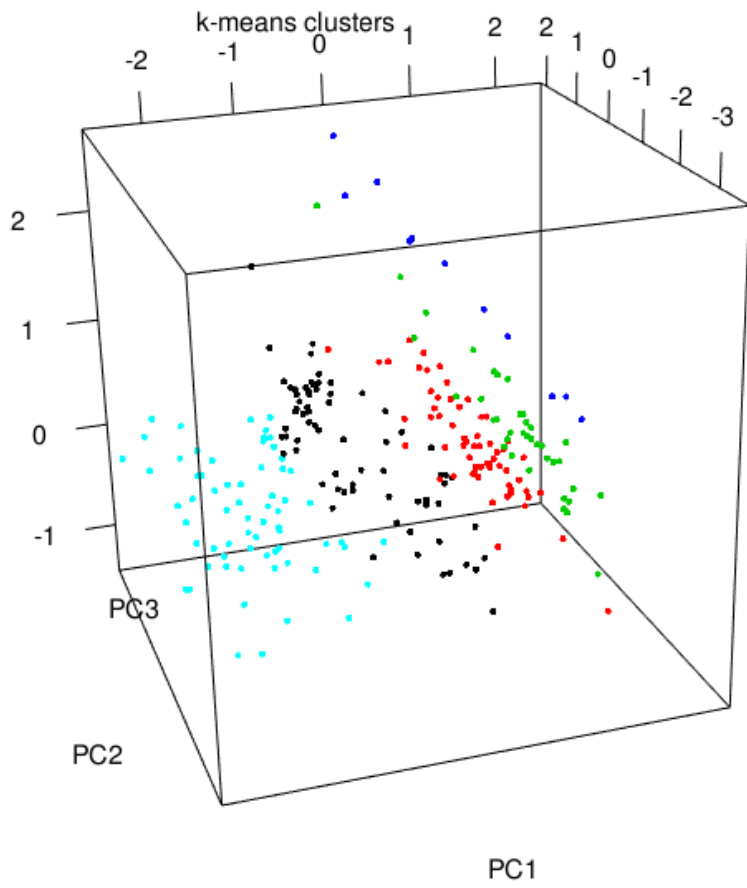




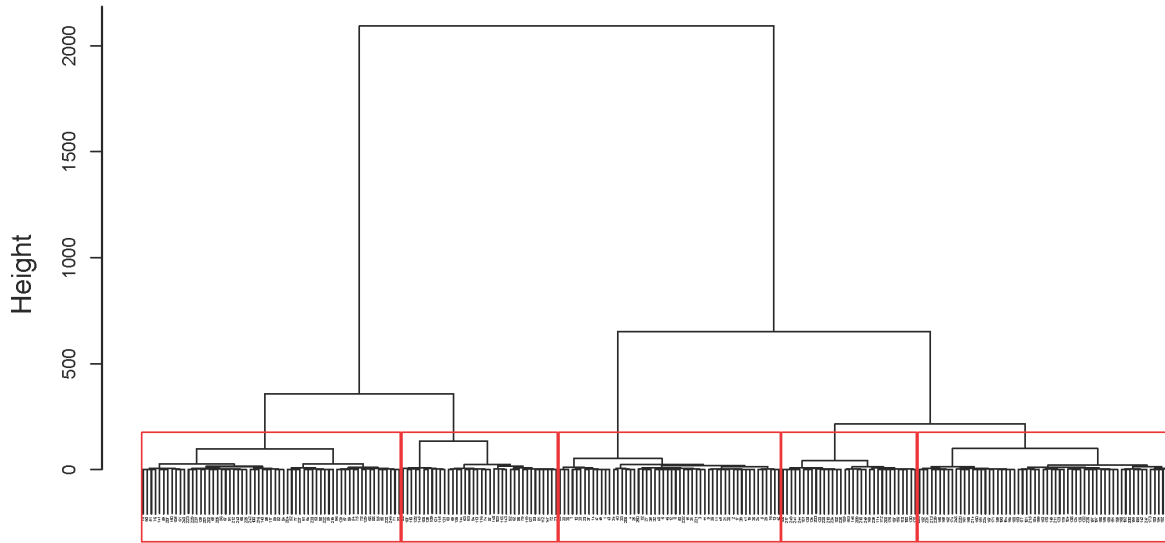
Bluegill Bowfin Carp Goldeye Largemouth_Bass



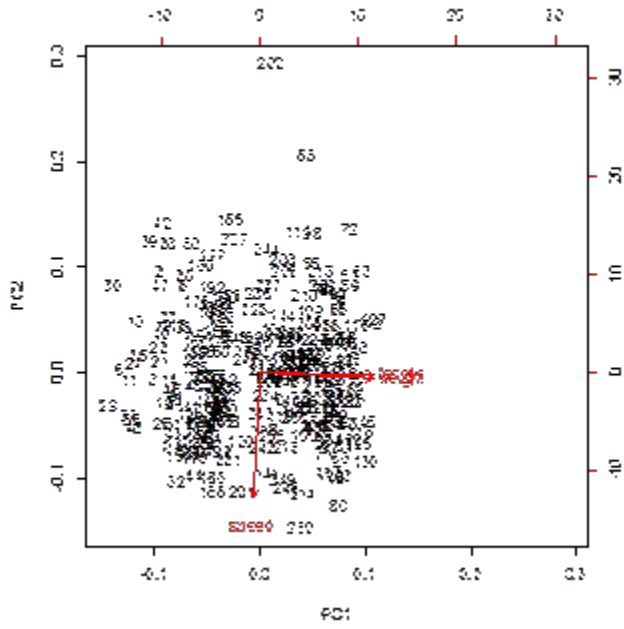




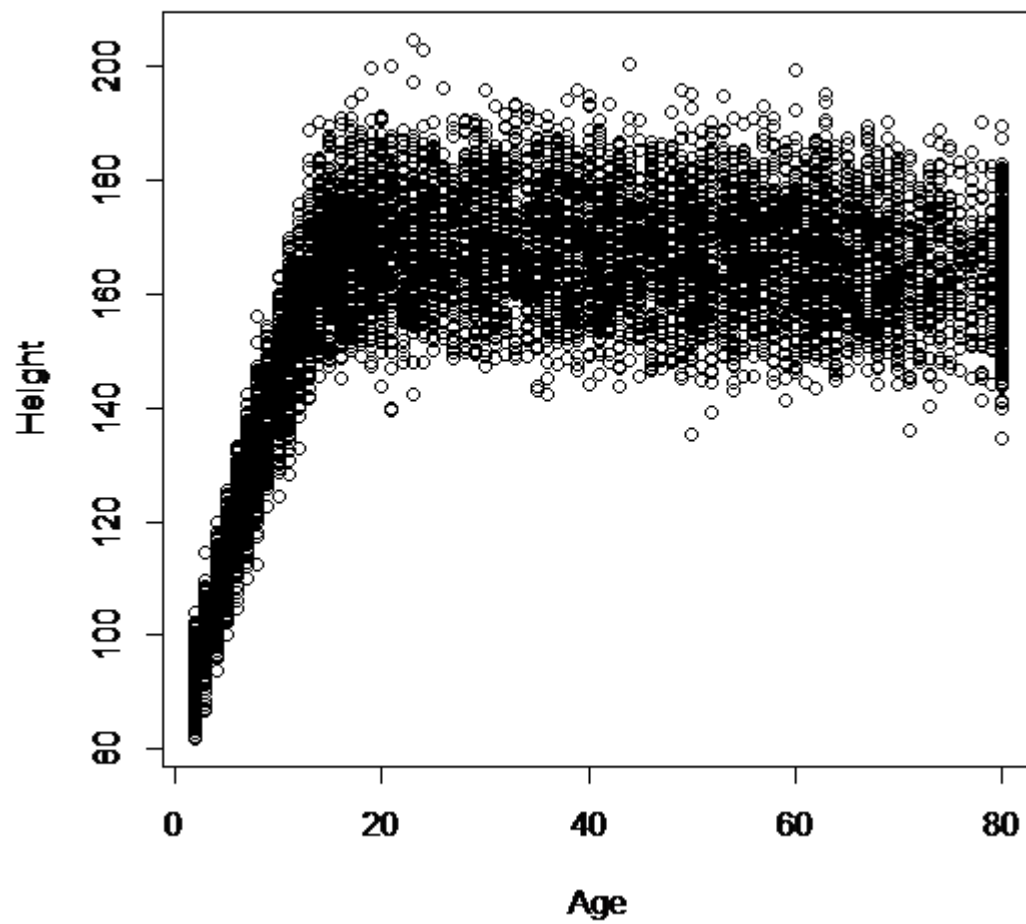
Cluster Dendrogram



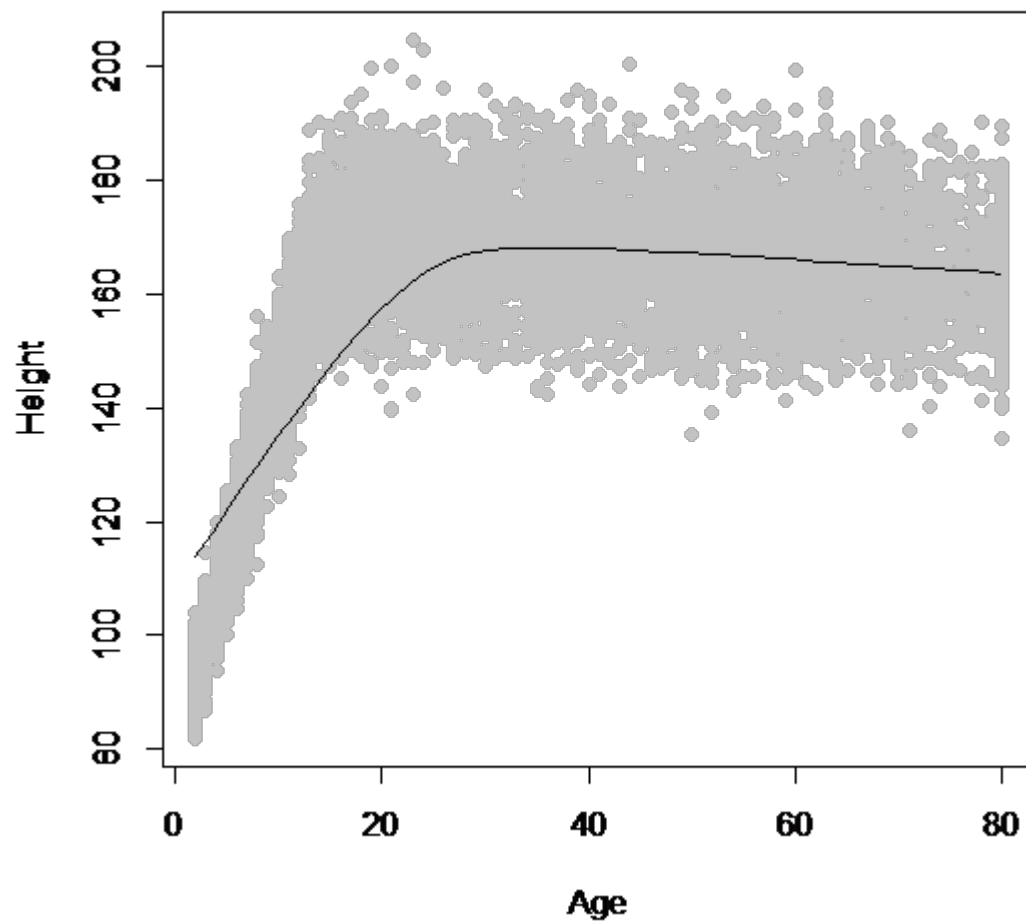
hclust (*, "ward")



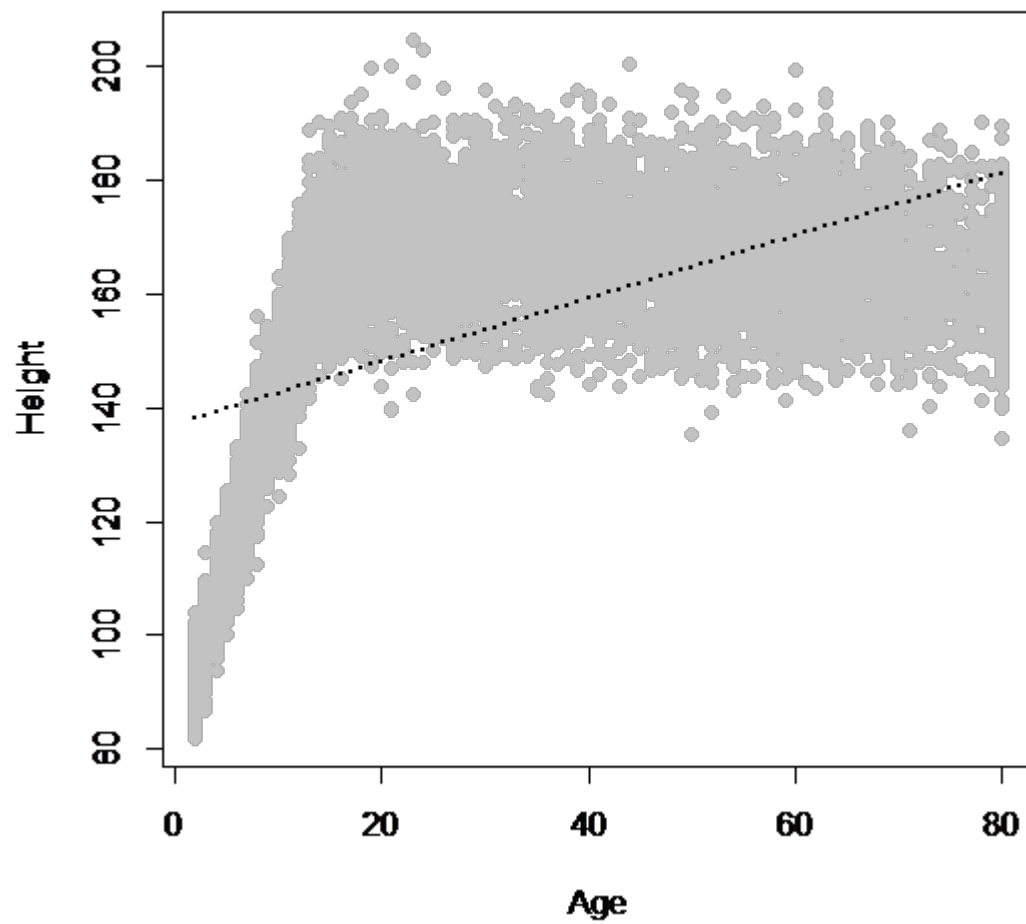
Height vs Age



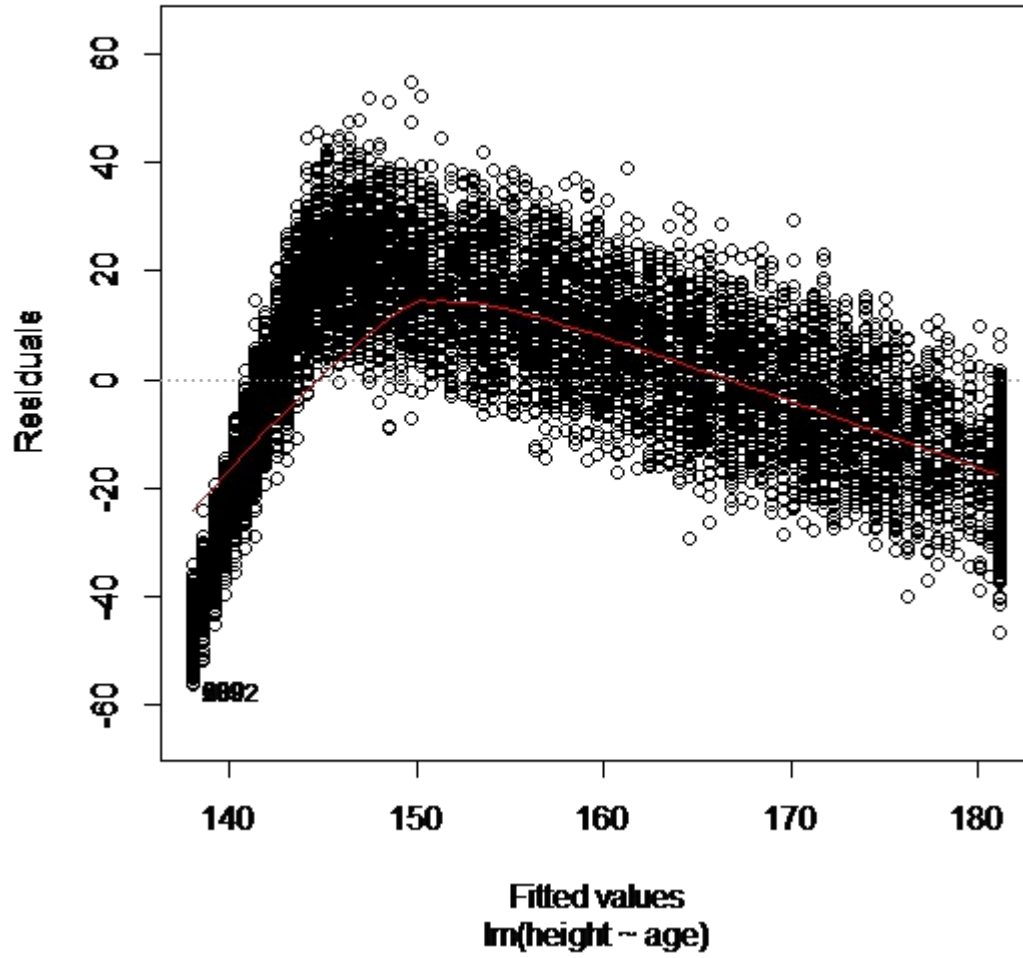
Height vs Age

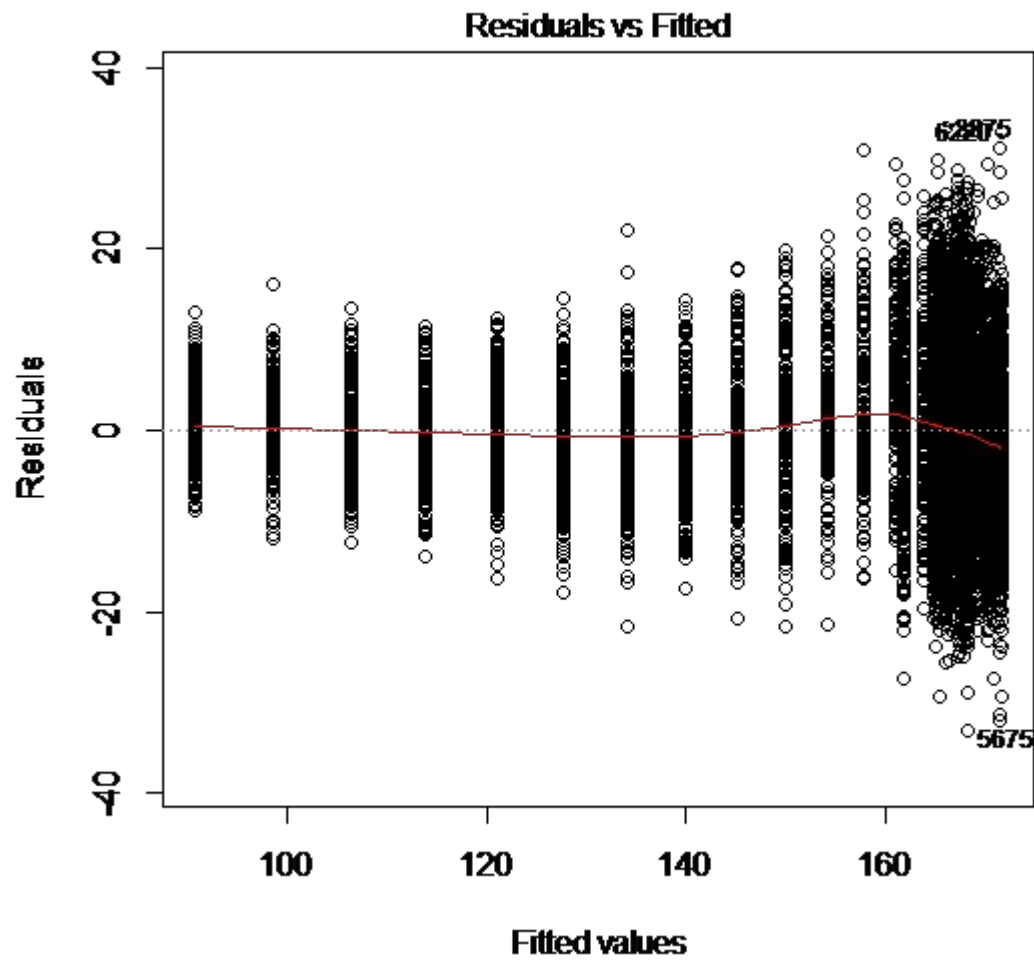


Height vs Age

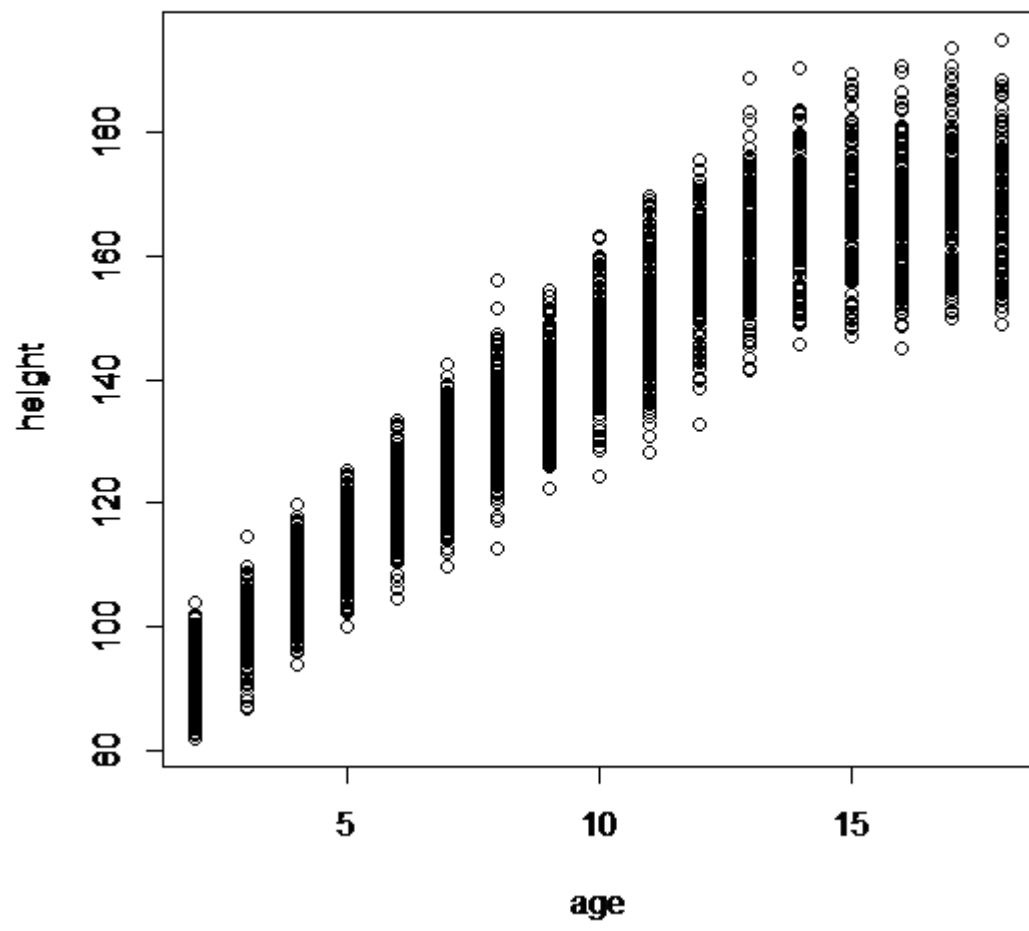


Residuals vs Fitted

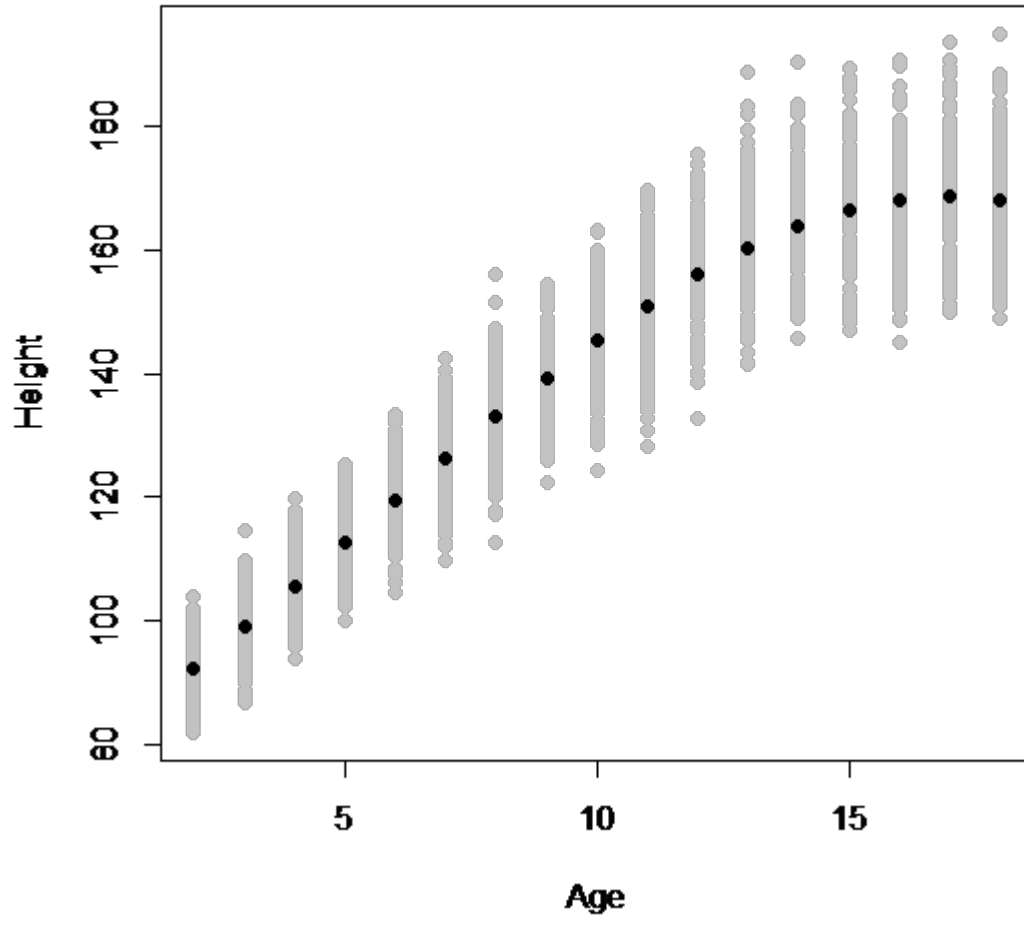




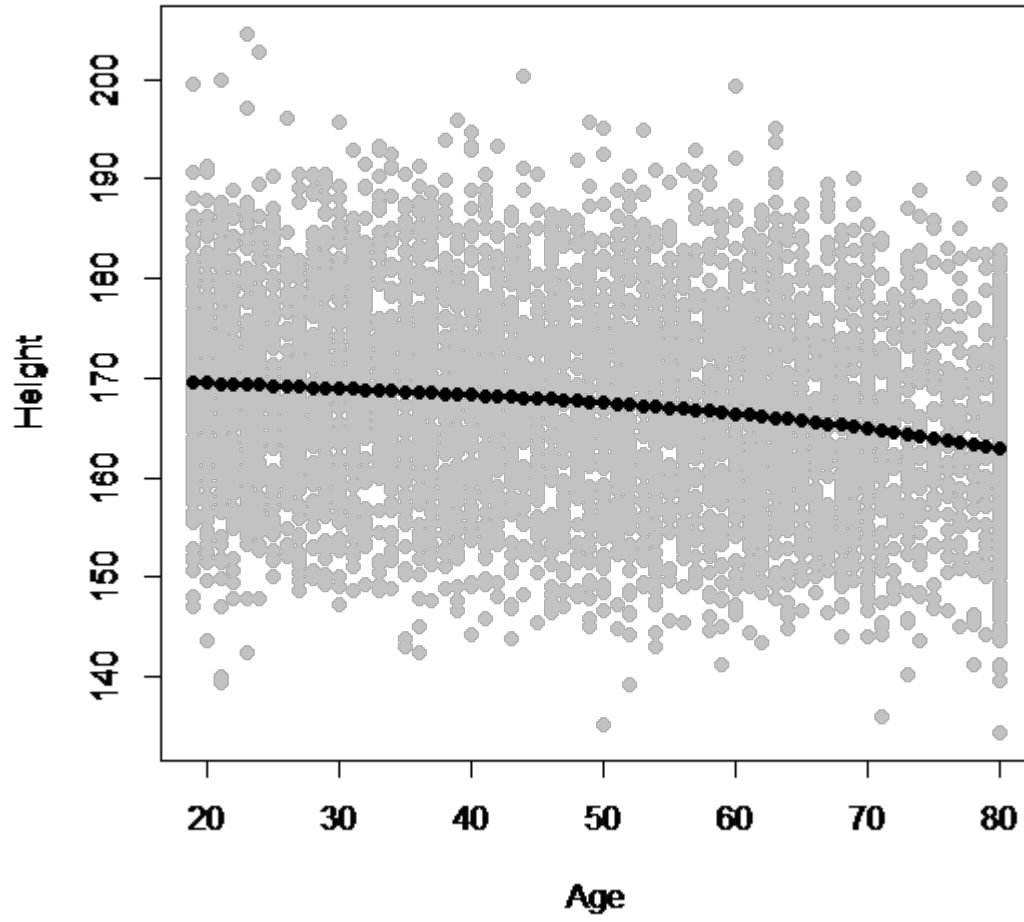
$\text{lm}(\text{height} \sim \text{age} + \text{age} + \text{K}(\text{age}^2) + \text{K}(\text{age}^3) + \text{K}(\text{age}^4) + \text{K}(\text{age}^5) + \text{K}(\text{age}^6 \dots$



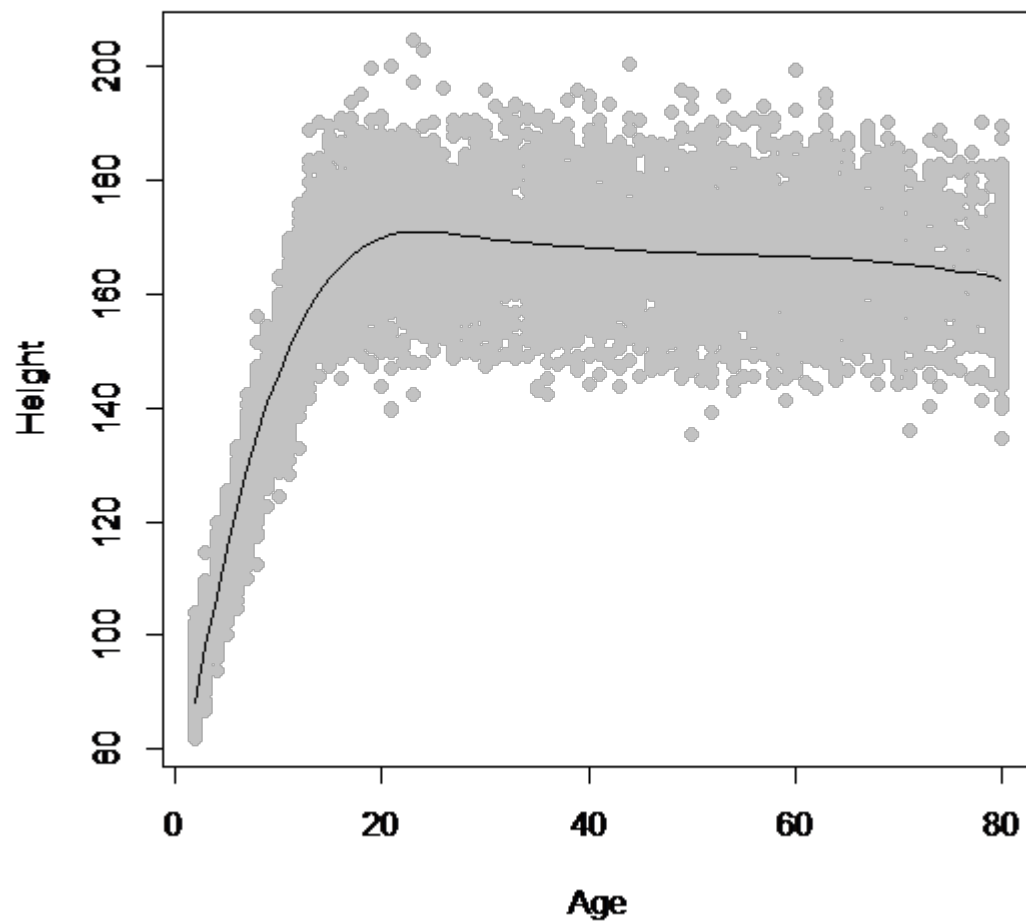
Height vs Age (in youths)



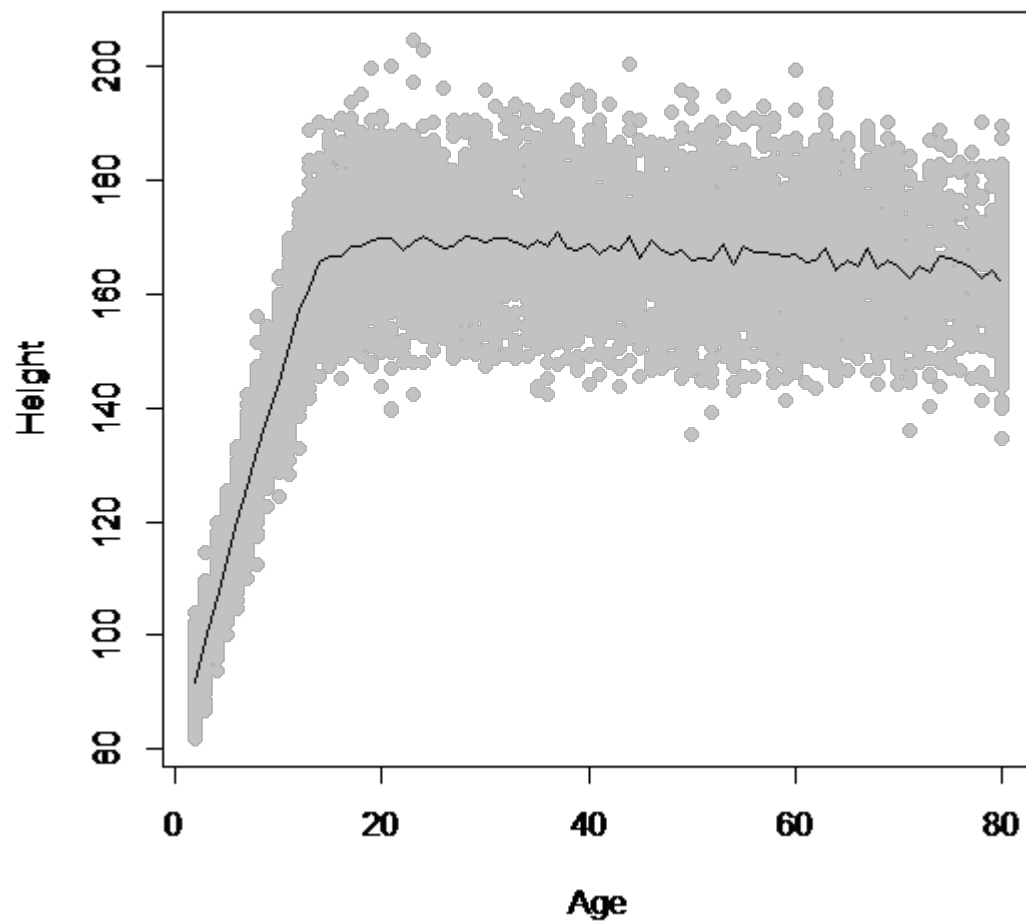
Height vs Age (in adults)



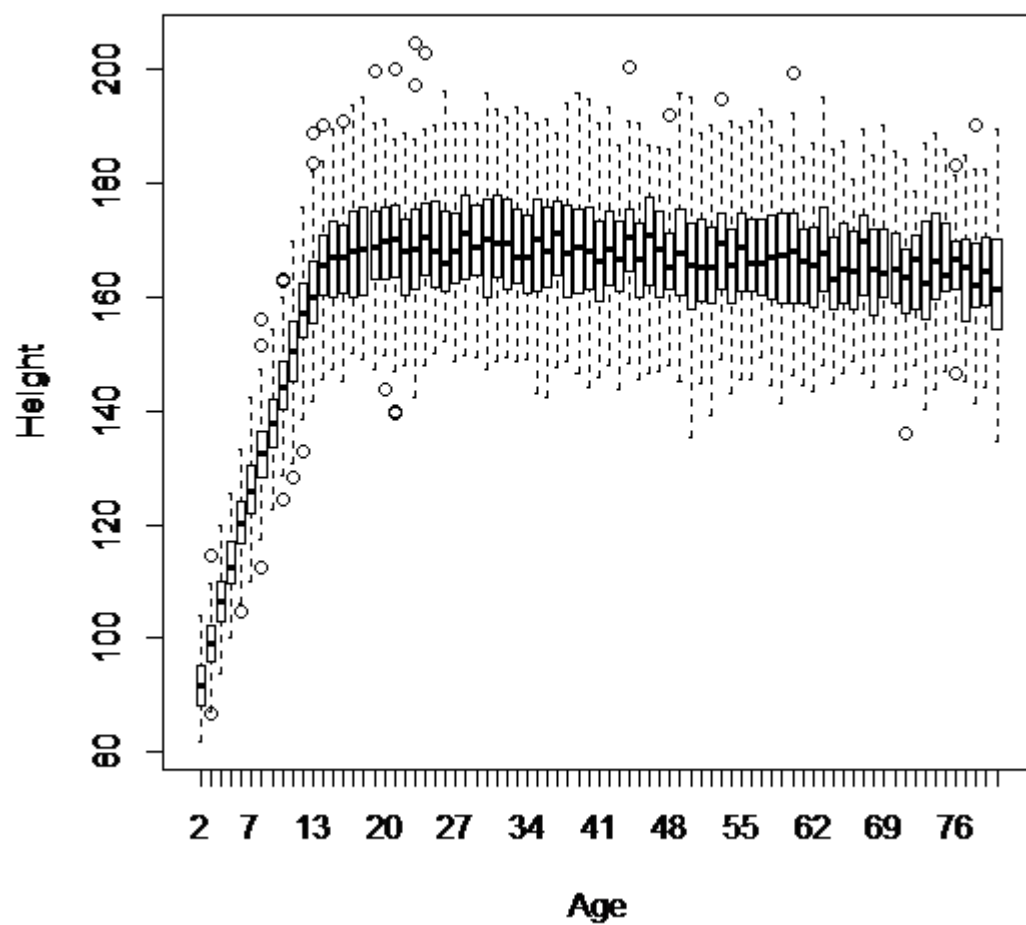
Height vs Age



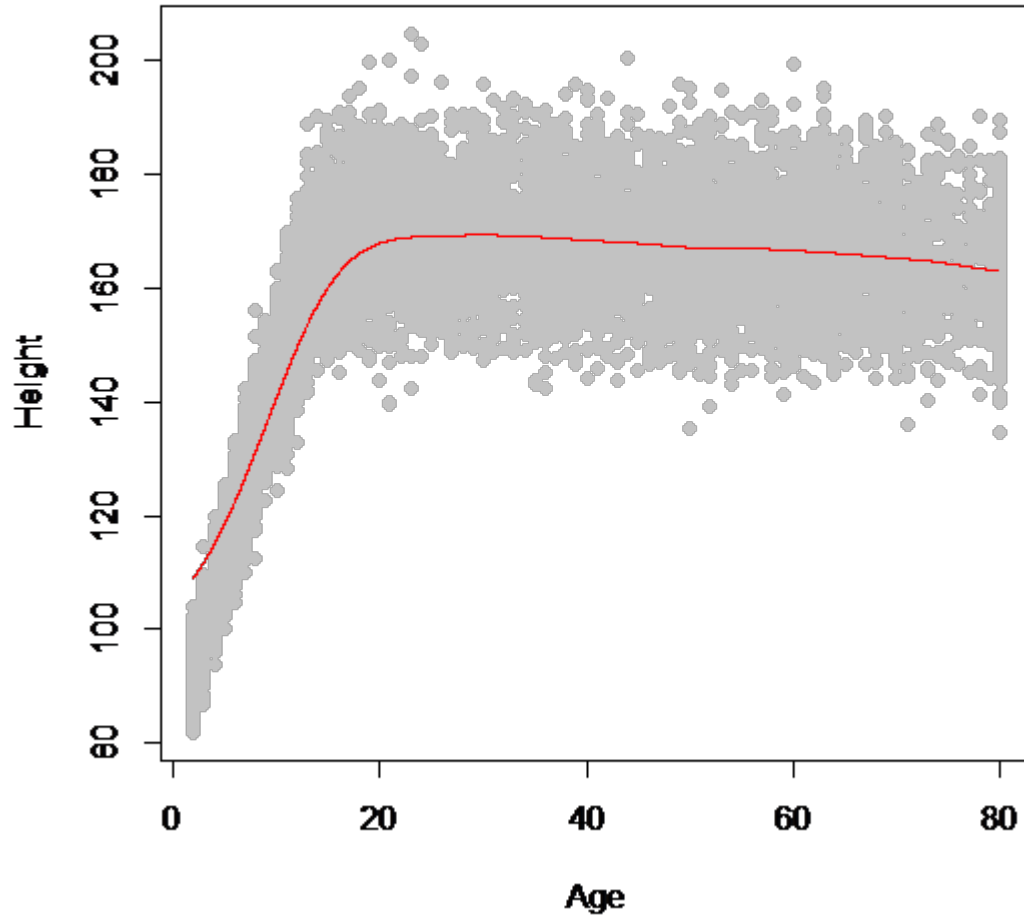
Height vs Age



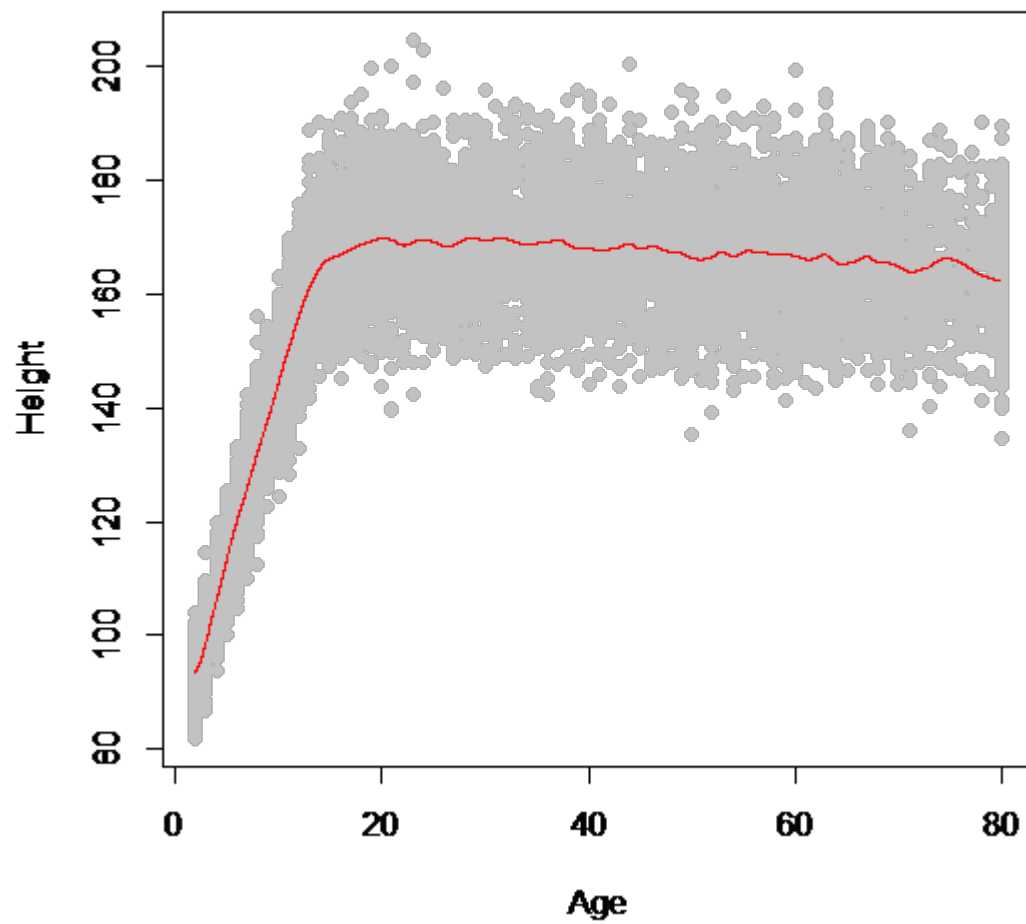
Height vs Age



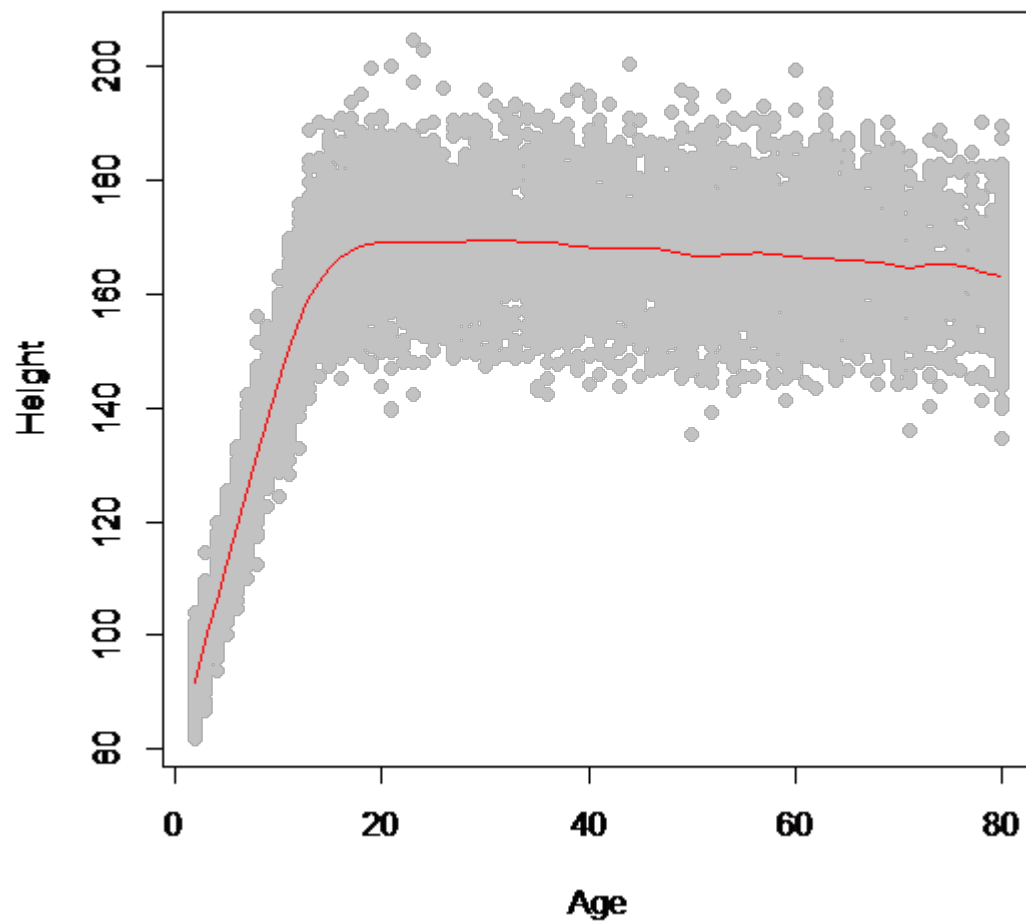
Height vs Age in Males

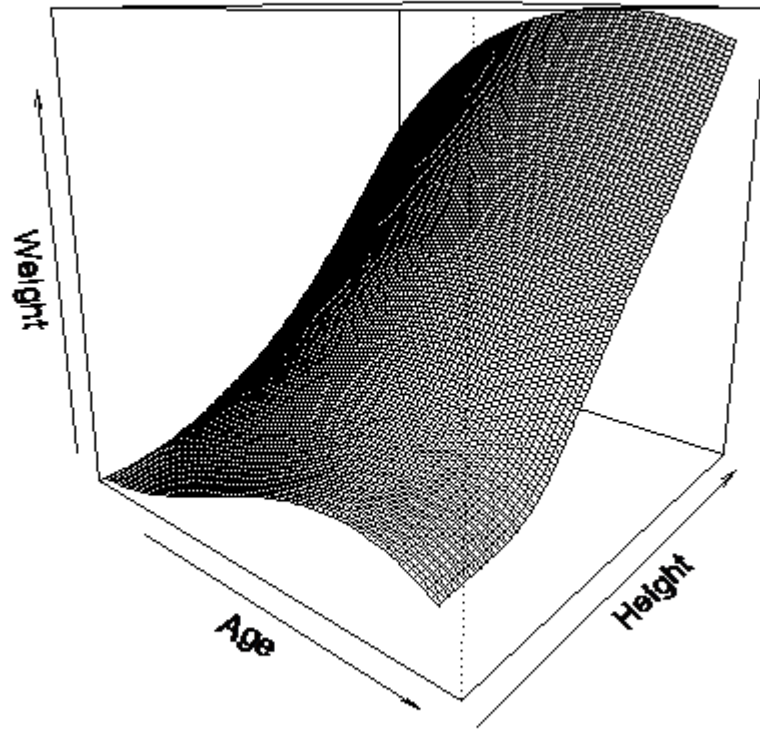


Height vs Age

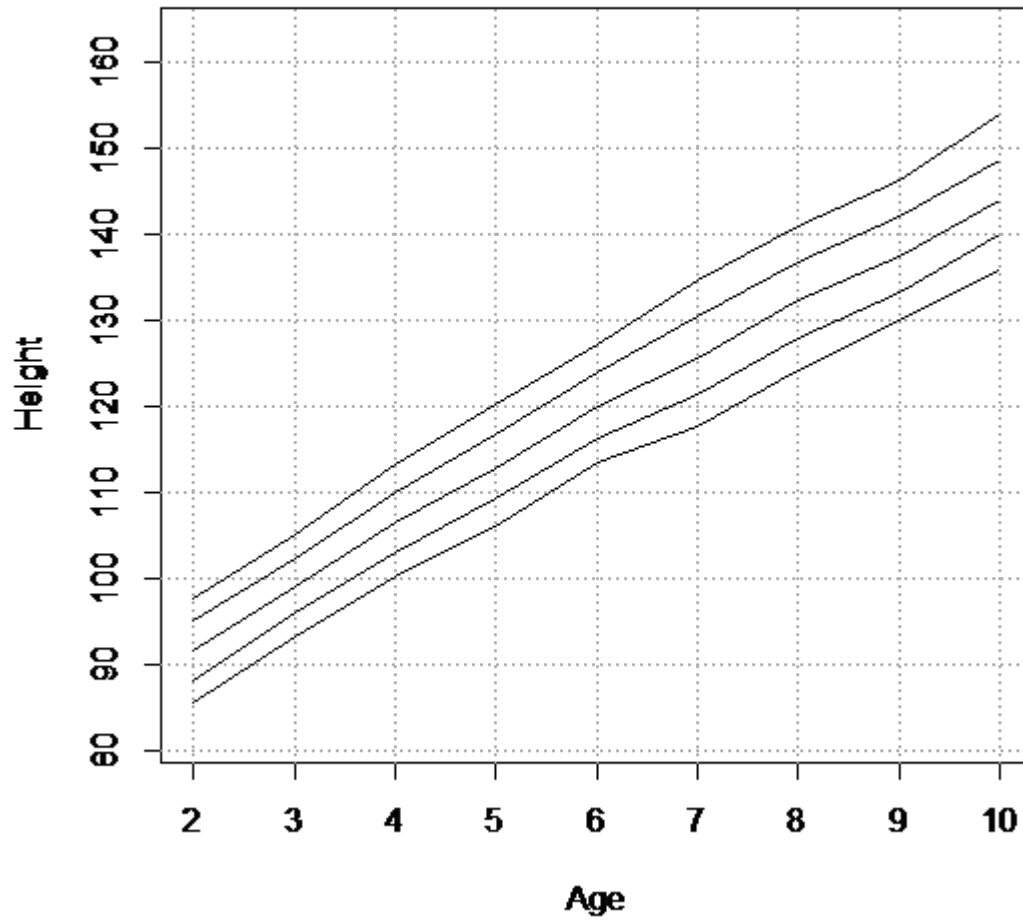


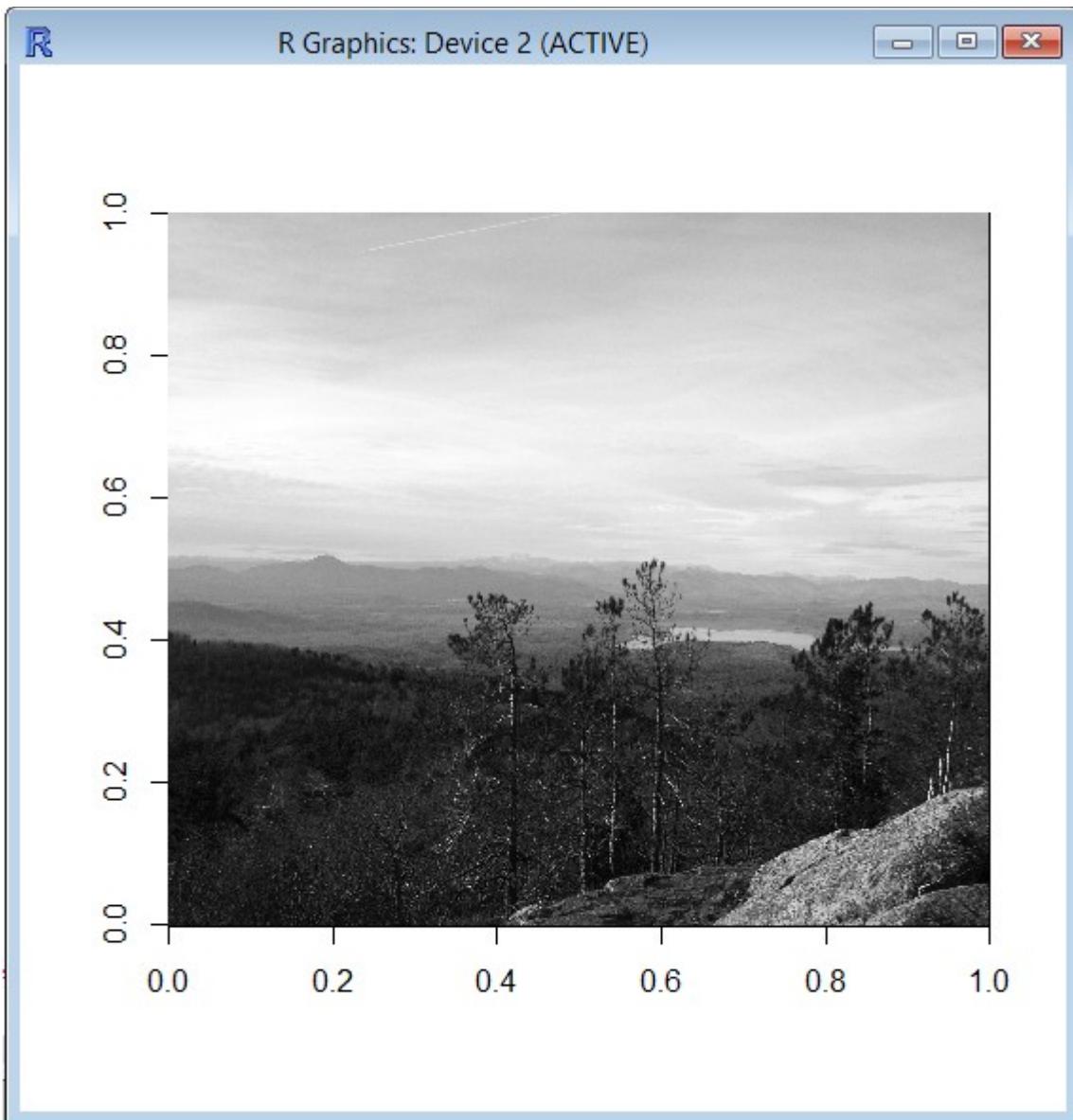
Height vs Age

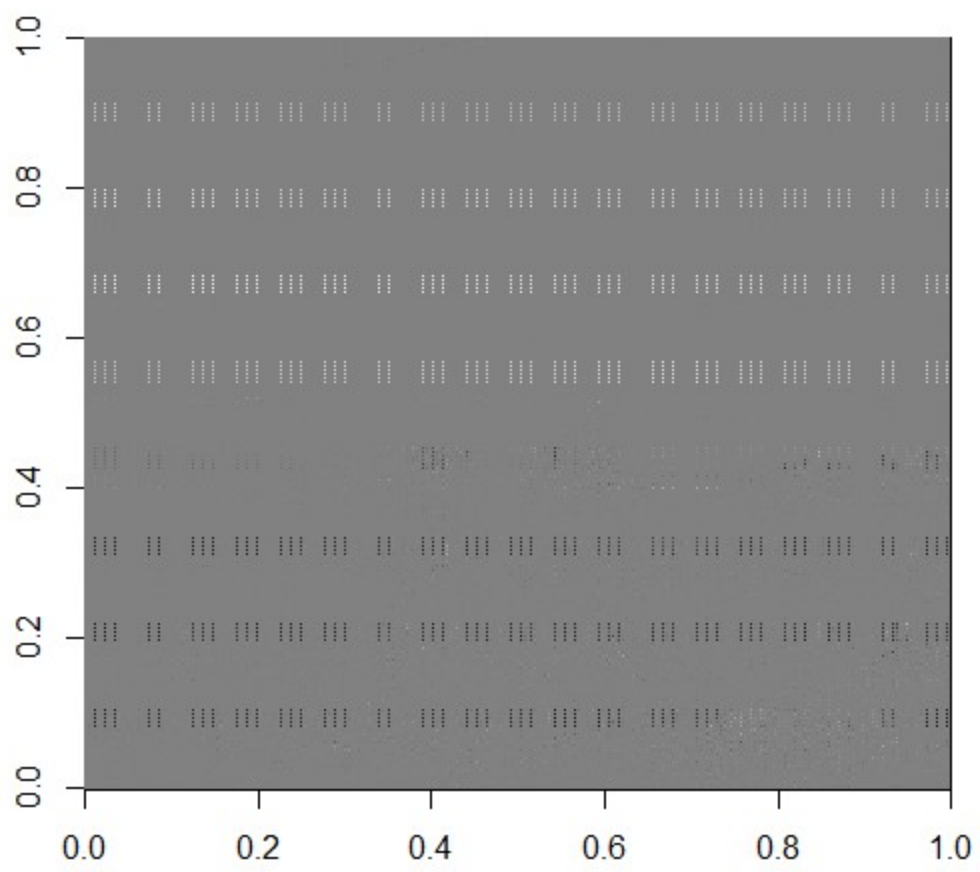


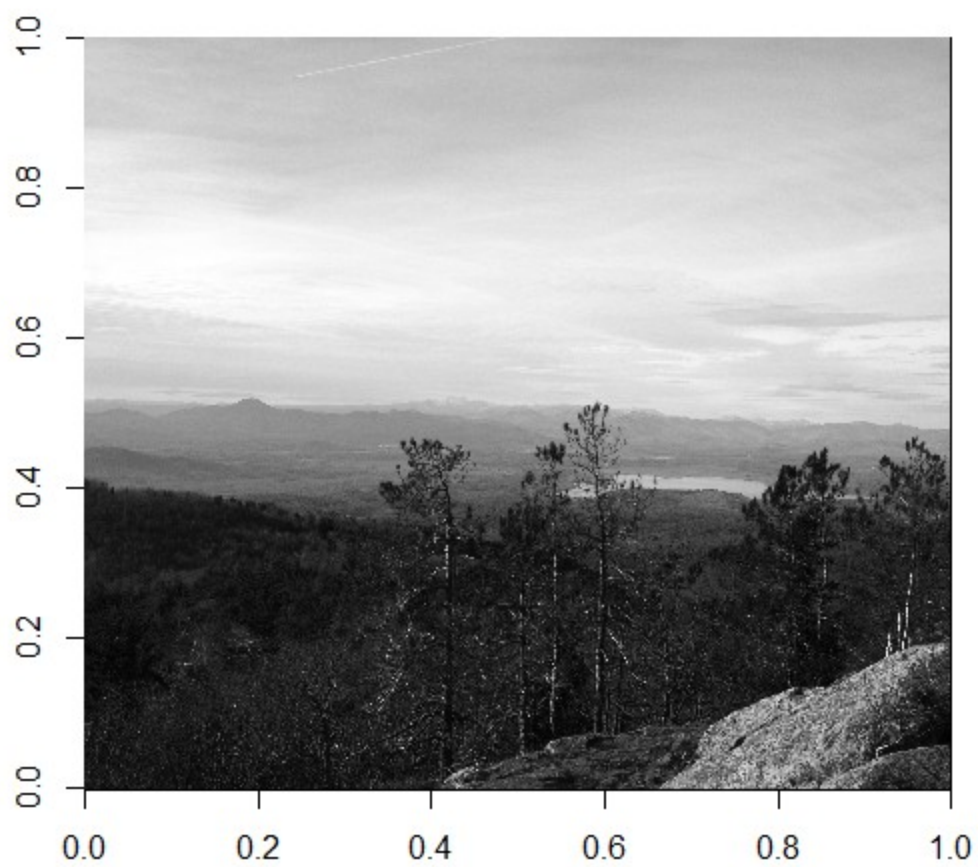


Quantiles of Age for Height

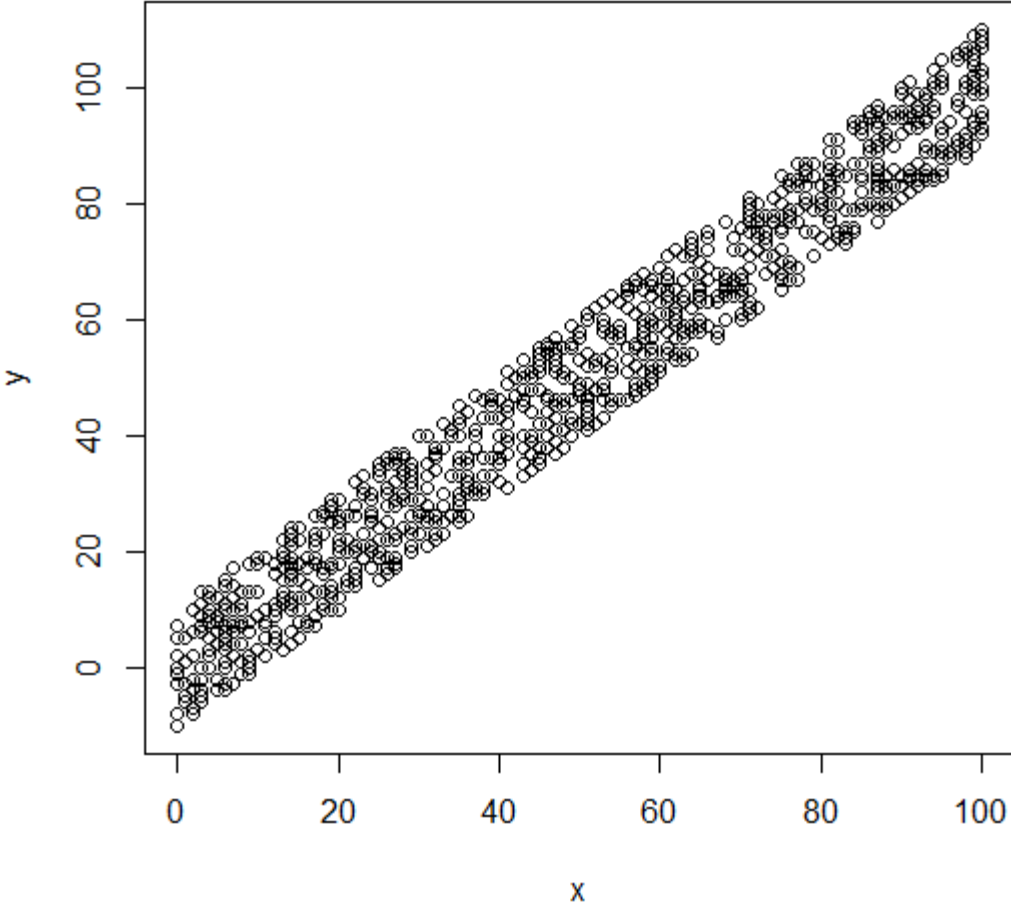


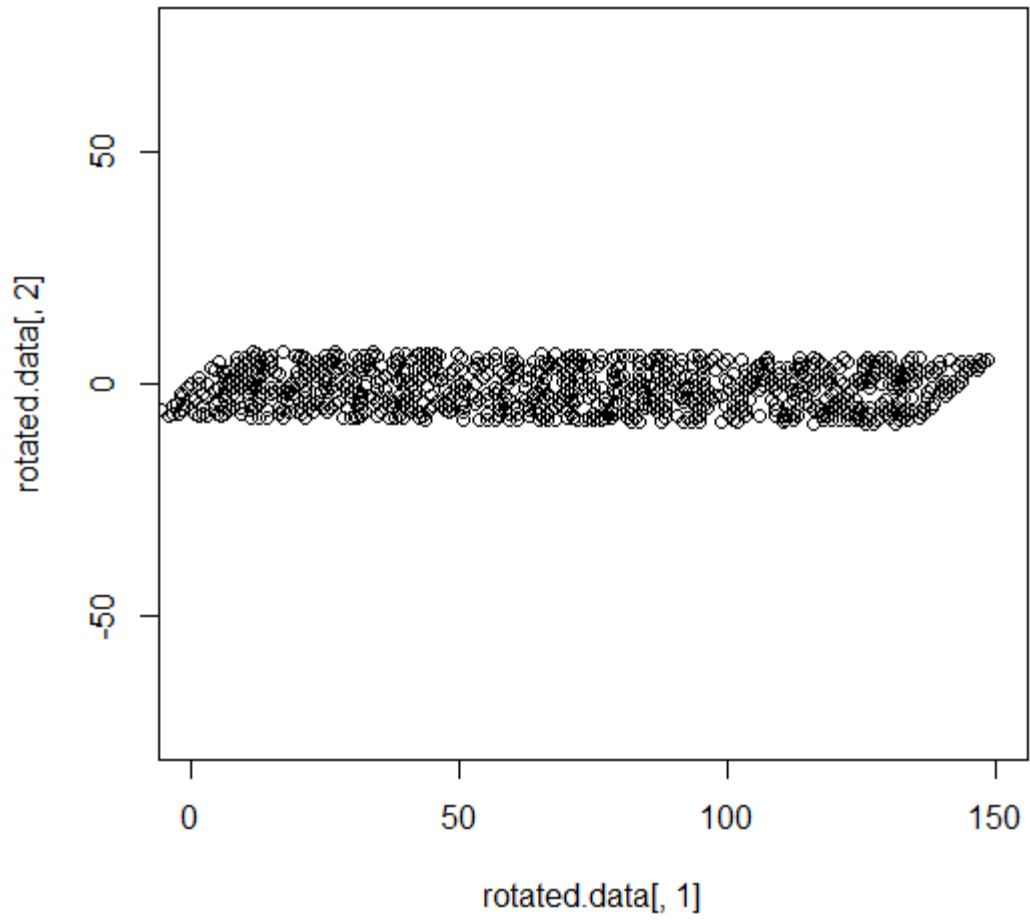


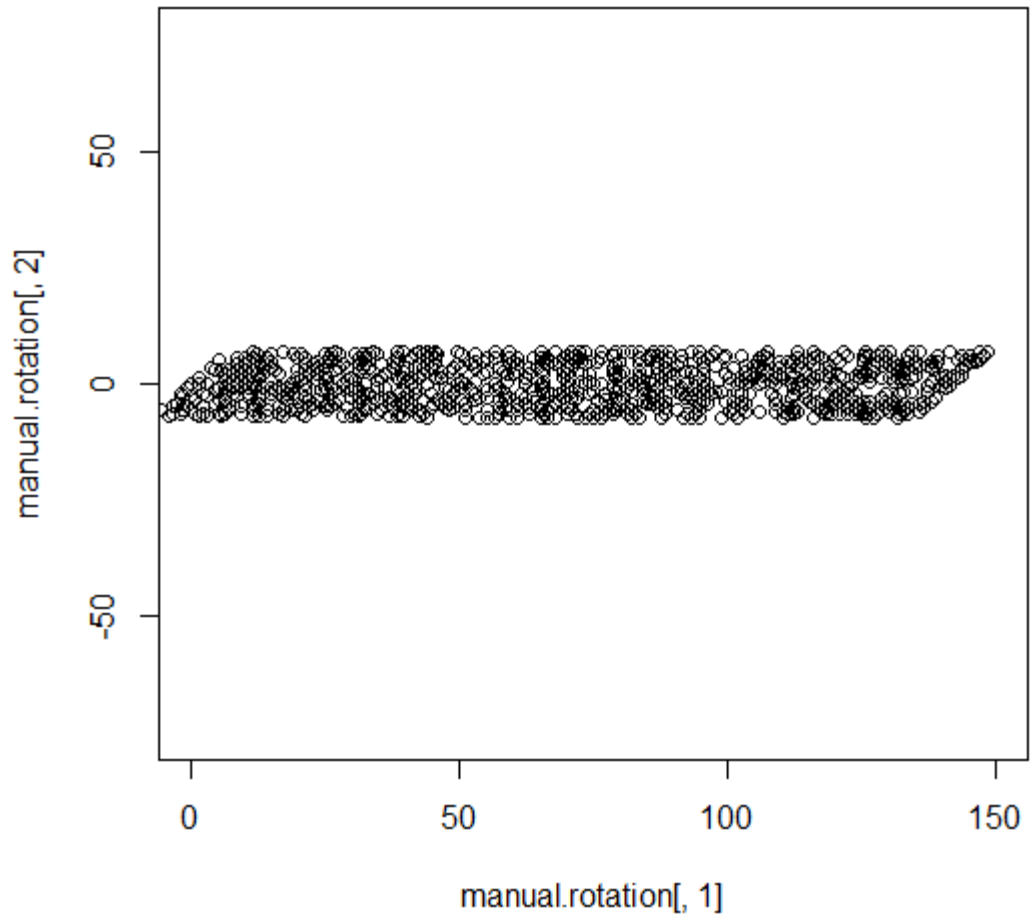




Chapter 6: Principal Component Analysis and the Common Factor Model

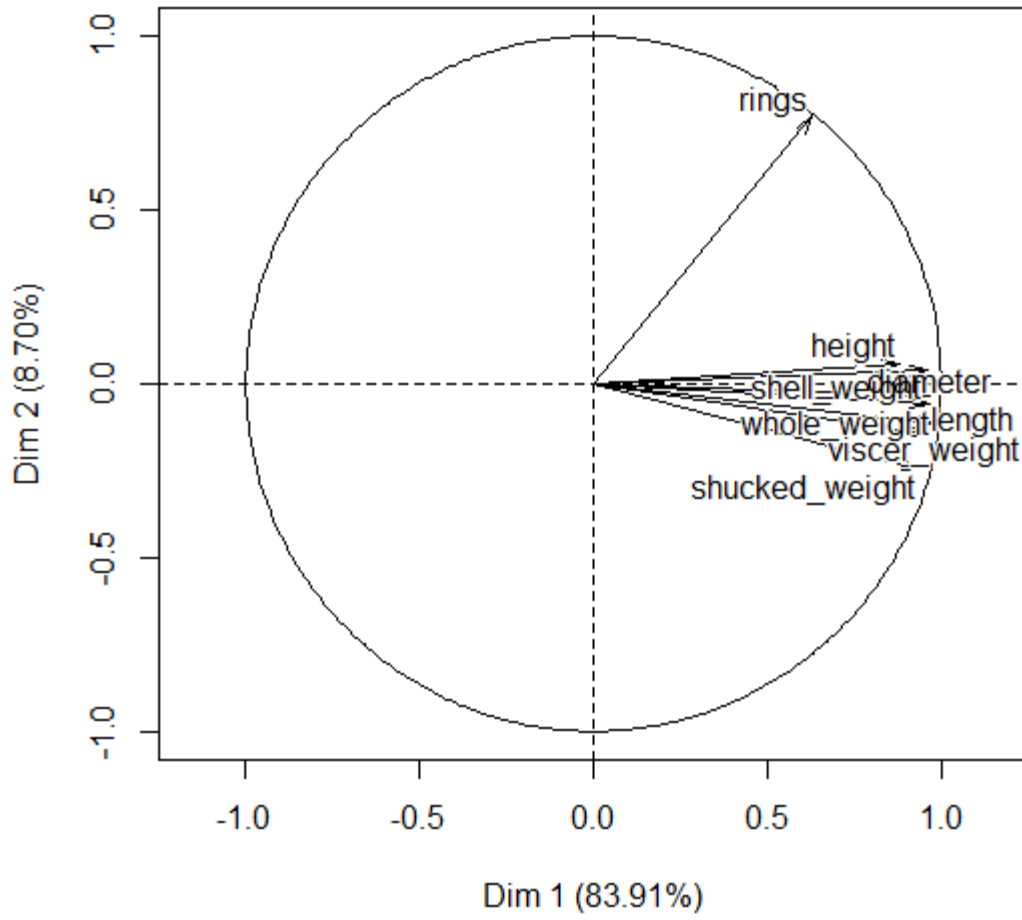




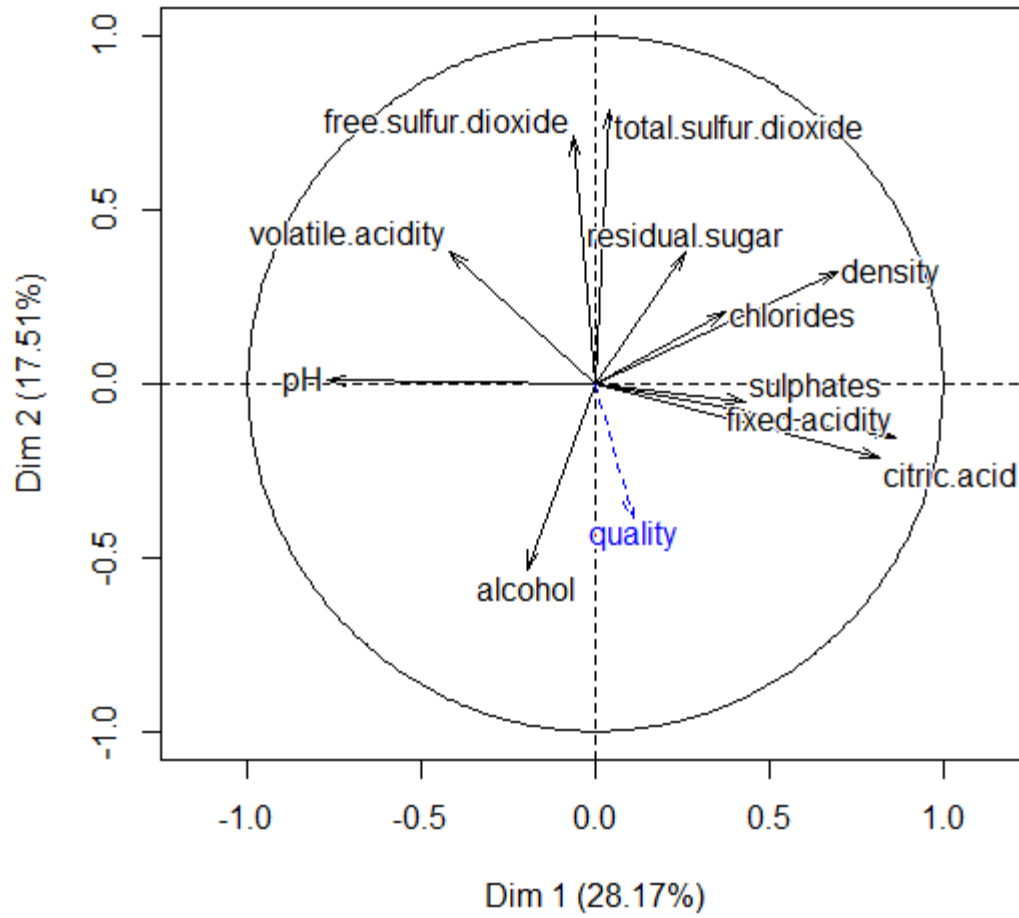


```
R Console
> summary(wine.prcomp)
Importance of components:
              PC1      PC2      PC3      PC4      PC5
Standard deviation 33.6721  7.61153  1.76105  1.34886  1.02291
Proportion of Variance 0.9466  0.04837  0.00259  0.00152  0.00087
Cumulative Proportion 0.9466  0.99495  0.99753  0.99905  0.99993
              PC6      PC7      PC8      PC9      PC10
Standard deviation  0.20346  0.15229  0.10652  0.10039  0.03814
Proportion of Variance 0.00003  0.00002  0.00001  0.00001  0.00000
Cumulative Proportion 0.99996  0.99998  0.99999  1.00000  1.00000
              PC11
Standard deviation   0.0007493
Proportion of Variance 0.0000000
Cumulative Proportion 1.0000000
> summary(wine.prcomp.scaled)
Importance of components:
              PC1      PC2      PC3      PC4      PC5      PC6
Standard deviation  1.7604  1.3878  1.2452  1.1015  0.97943  0.81216
Proportion of Variance 0.2817  0.1751  0.1410  0.1103  0.08721  0.05996
Cumulative Proportion 0.2817  0.4568  0.5978  0.7081  0.79528  0.85525
              PC7      PC8      PC9      PC10      PC11
Standard deviation  0.76406  0.65035  0.58706  0.42583  0.24405
Proportion of Variance 0.05307  0.03845  0.03133  0.01648  0.00541
Cumulative Proportion 0.90832  0.94677  0.97810  0.99459  1.00000
>
> |
```


Variables factor map (PCA)



Variables factor map (PCA)





R Console



```
> summary(wine.pca)
```

Call:

```
PCA(red.wine, quanti.sup = 12)
```

Eigenvalues

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6	Dim.7	Dim.8	Dim.9	Dim.10	Dim.11
Variance	3.099	1.926	1.551	1.213	0.959	0.660	0.584	0.423	0.345	0.181	0.060
% of var.	28.174	17.508	14.096	11.029	8.721	5.996	5.307	3.845	3.133	1.648	0.541
Cumulative % of var.	28.174	45.682	59.778	70.807	79.528	85.525	90.832	94.677	97.810	99.459	100.000

Individuals (the 10 first)

	Dist	Dim.1	ctr	cos2	Dim.2	ctr	cos2	Dim.3	ctr	cos2
1	2.645	-1.620	0.053	0.375	0.451	0.007	0.029	-1.774	0.127	0.450
2	2.824	-0.799	0.013	0.080	1.857	0.112	0.432	-0.912	0.034	0.104
3	1.936	-0.748	0.011	0.149	0.882	0.025	0.208	-1.171	0.055	0.366
4	3.045	2.358	0.112	0.600	-0.270	0.002	0.008	0.243	0.002	0.006
5	2.645	-1.620	0.053	0.375	0.451	0.007	0.029	-1.774	0.127	0.450
6	2.540	-1.584	0.051	0.389	0.569	0.011	0.050	-1.538	0.095	0.367
7	2.115	-1.101	0.024	0.271	0.608	0.012	0.083	-1.076	0.047	0.259
8	2.726	-2.249	0.102	0.681	-0.417	0.006	0.023	-0.987	0.039	0.131
9	2.093	-1.087	0.024	0.270	-0.309	0.003	0.022	-1.518	0.093	0.526
10	3.302	0.655	0.009	0.039	1.665	0.090	0.254	1.209	0.059	0.134

Variables (the 10 first)

	Dim.1	ctr	cos2	Dim.2	ctr	cos2	Dim.3	ctr	cos2
fixed.acidity	0.861	23.943	0.742	-0.153	1.221	0.024	-0.154	1.520	0.024
volatile.acidity	-0.420	5.692	0.176	0.382	7.559	0.146	-0.560	20.247	0.314
citric.acid	0.816	21.495	0.666	-0.211	2.304	0.044	0.297	5.676	0.088
residual.sugar	0.257	2.135	0.066	0.378	7.403	0.143	0.126	1.026	0.016
chlorides	0.374	4.505	0.140	0.205	2.192	0.042	-0.115	0.858	0.013
free.sulfur.dioxide	-0.064	0.131	0.004	0.713	26.375	0.508	0.534	18.386	0.285
total.sulfur.dioxide	0.042	0.056	0.002	0.790	32.432	0.625	0.401	10.395	0.161
density	0.696	15.630	0.484	0.324	5.456	0.105	-0.422	11.483	0.178
pH	-0.772	19.230	0.596	0.009	0.005	0.000	0.072	0.333	0.005
sulphates	0.428	5.901	0.183	-0.052	0.141	0.003	0.348	7.828	0.121

Supplementary continuous variable

	Dim.1	cos2	Dim.2	cos2	Dim.3	cos2
quality	0.110	0.012	-0.387	0.150	0.399	0.159

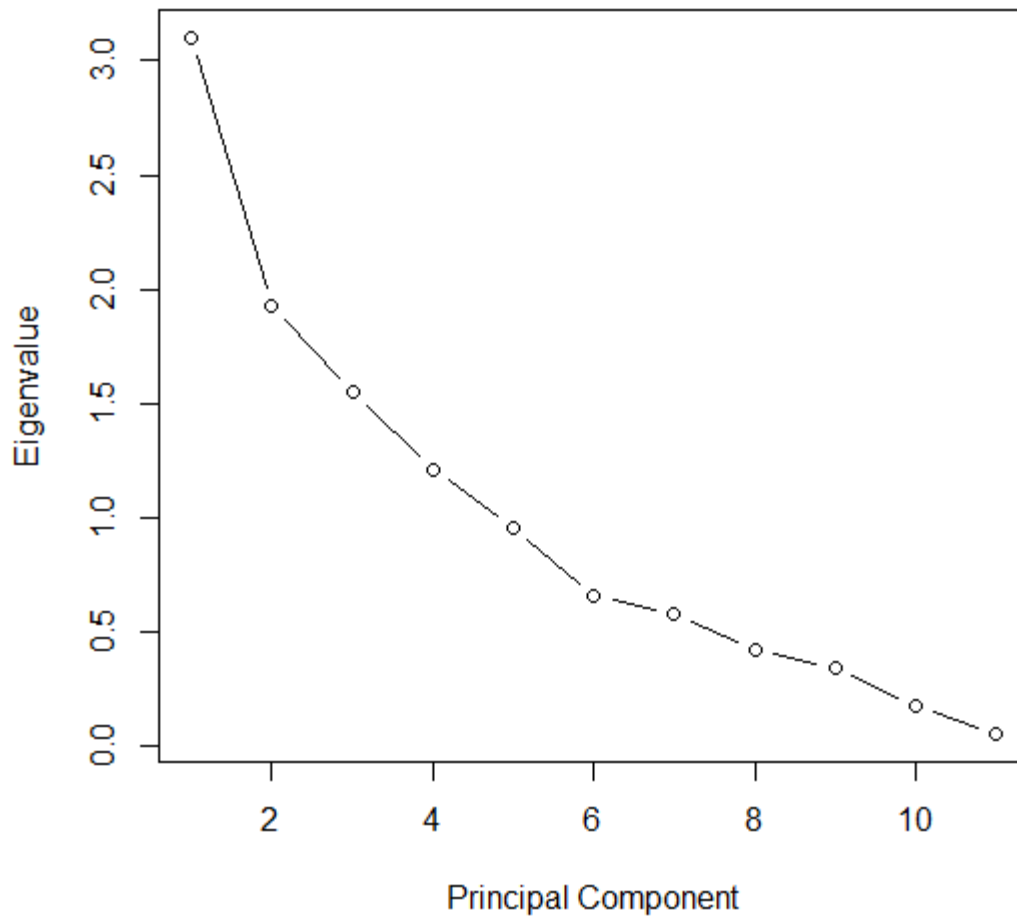
```
> |
```

```
<
```

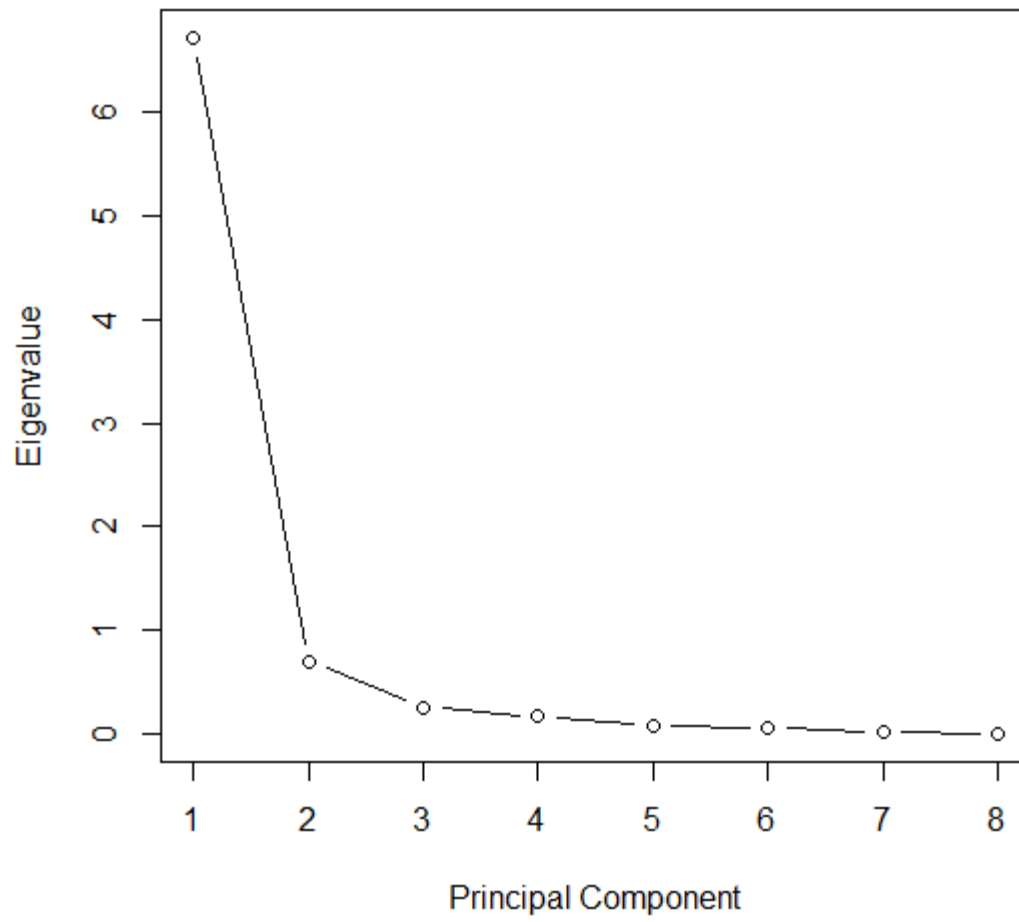
```
>
```



Eigenvalues of Principal Components



Eigenvalues of Principal Components



```

R Console
> summary(wine.pca, ncp = 4, )

Call:
PCA(red.wine, quanti.sup = 12)

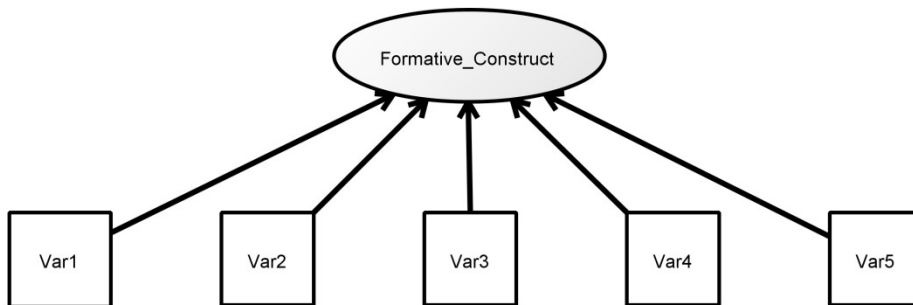
Eigenvalues
      Dim.1 Dim.2 Dim.3 Dim.4 Dim.5 Dim.6 Dim.7 Dim.8 Dim.9 Dim.10 Dim.11
Variance  3.099  1.926  1.551  1.213  0.959  0.660  0.584  0.423  0.345  0.181  0.060
% of var. 28.174 17.508 14.096 11.029  8.721  5.996  5.307  3.845  3.133  1.648  0.541
Cumulative % of var. 28.174 45.682 59.778 70.807 79.528 85.525 90.832 94.677 97.810 99.459 100.000

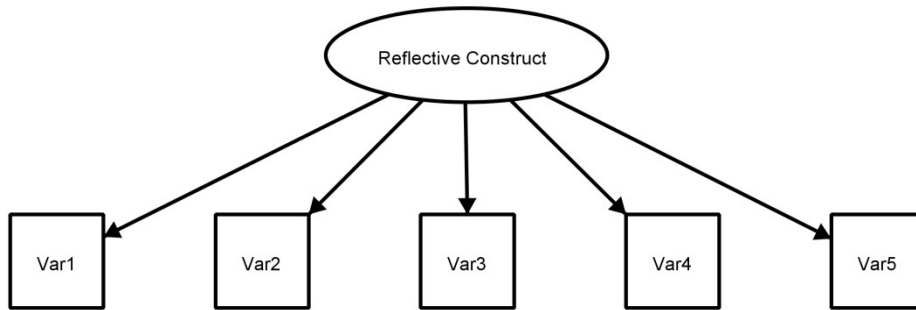
Individuals (the 10 first)
      Dist Dim.1 ctr cos2 Dim.2 ctr cos2 Dim.3 ctr cos2 Dim.4 ctr cos2
1 | 2.645 | -1.620 0.053 0.375 | 0.451 0.007 0.029 | -1.774 0.127 0.450 | 0.044 0.000 0.000 |
2 | 2.824 | -0.799 0.013 0.080 | 1.857 0.112 0.432 | -0.912 0.034 0.104 | 0.548 0.015 0.038 |
3 | 1.936 | -0.748 0.011 0.149 | 0.882 0.025 0.208 | -1.171 0.055 0.366 | 0.411 0.009 0.045 |
4 | 3.045 | 2.358 0.112 0.600 | -0.270 0.002 0.008 | 0.243 0.002 0.006 | -0.928 0.044 0.093 |
5 | 2.645 | -1.620 0.053 0.375 | 0.451 0.007 0.029 | -1.774 0.127 0.450 | 0.044 0.000 0.000 |
6 | 2.540 | -1.584 0.051 0.389 | 0.569 0.011 0.050 | -1.538 0.095 0.367 | 0.024 0.000 0.000 |
7 | 2.115 | -1.101 0.024 0.271 | 0.608 0.012 0.083 | -1.076 0.047 0.259 | -0.344 0.006 0.026 |
8 | 2.726 | -2.249 0.102 0.681 | -0.417 0.006 0.023 | -0.987 0.039 0.131 | -0.001 0.000 0.000 |
9 | 2.093 | -1.087 0.024 0.270 | -0.309 0.003 0.022 | -1.518 0.093 0.526 | 0.003 0.000 0.000 |
10 | 3.302 | 0.655 0.009 0.039 | 1.665 0.090 0.254 | 1.209 0.059 0.134 | -0.825 0.035 0.062 |

Variables (the 10 first)
      Dim.1 ctr cos2 Dim.2 ctr cos2 Dim.3 ctr cos2 Dim.4 ctr cos2
fixed.acidity | 0.861 23.943 0.742 | -0.153 1.221 0.024 | -0.154 1.520 0.024 | -0.253 5.272 0.064 |
volatile.acidity | -0.420 5.692 0.176 | 0.382 7.559 0.146 | -0.560 20.247 0.314 | 0.087 0.623 0.008 |
citric.acid | 0.816 21.495 0.666 | -0.211 2.304 0.044 | 0.297 5.676 0.088 | -0.087 0.631 0.008 |
residual.sugar | 0.257 2.135 0.066 | 0.378 7.403 0.143 | 0.126 1.026 0.016 | -0.411 13.897 0.169 |
chlorides | 0.374 4.505 0.140 | 0.205 2.192 0.042 | -0.115 0.858 0.013 | 0.734 44.382 0.538 |
free.sulfur.dioxide | -0.064 0.131 0.004 | 0.713 26.375 0.508 | 0.534 18.386 0.285 | -0.048 0.190 0.002 |
total.sulfur.dioxide | 0.042 0.056 0.002 | 0.790 32.432 0.625 | 0.401 10.395 0.161 | -0.038 0.120 0.001 |
density | 0.696 15.630 0.484 | 0.324 5.456 0.105 | -0.422 11.483 0.178 | -0.192 3.045 0.037 |
pH | -0.772 19.230 0.596 | 0.009 0.005 0.000 | 0.072 0.333 0.005 | -0.004 0.001 0.000 |
sulphates | 0.428 5.901 0.183 | -0.052 0.141 0.003 | 0.348 7.828 0.121 | 0.607 30.346 0.368 |

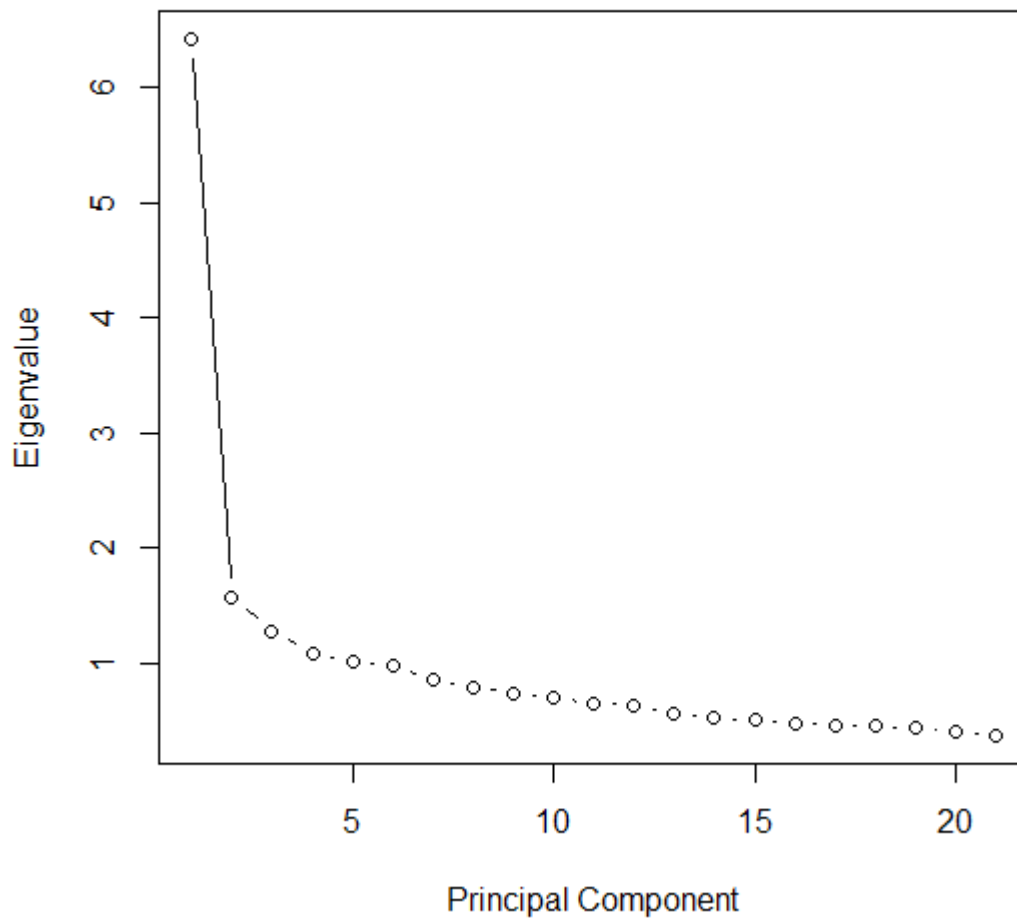
Supplementary continuous variable
      Dim.1 cos2 Dim.2 cos2 Dim.3 cos2 Dim.4 cos2
quality | 0.110 0.012 | -0.387 0.150 | 0.399 0.159 | -0.044 0.002 |

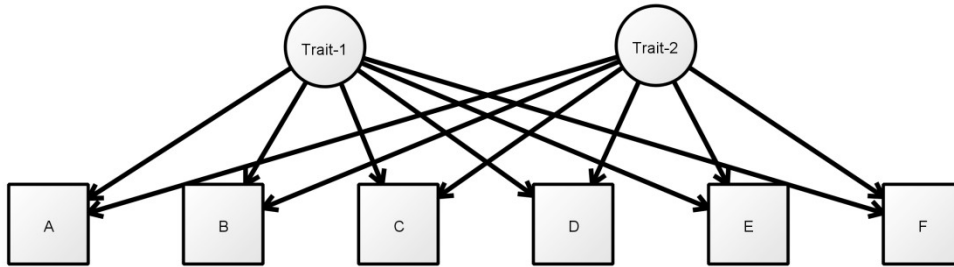
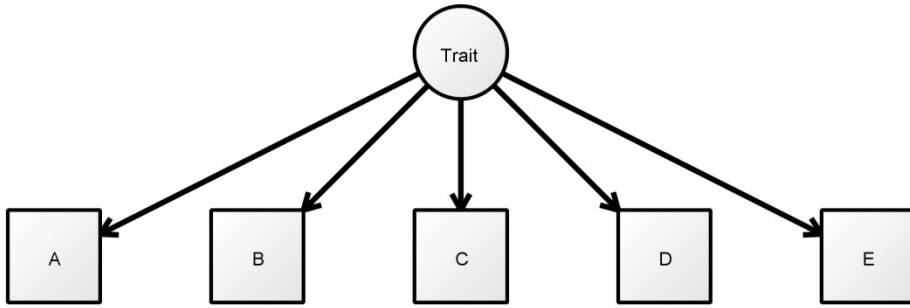
```



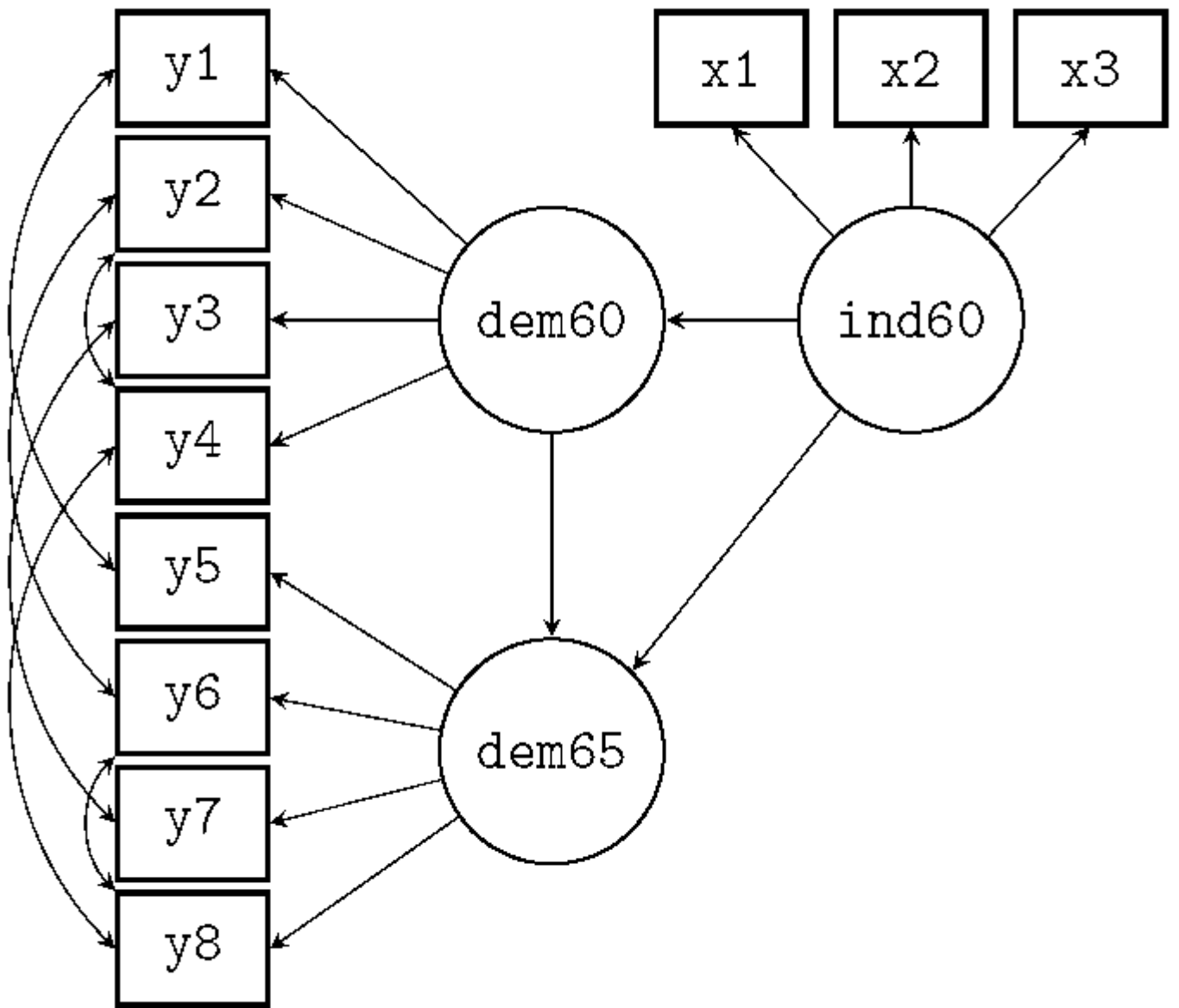


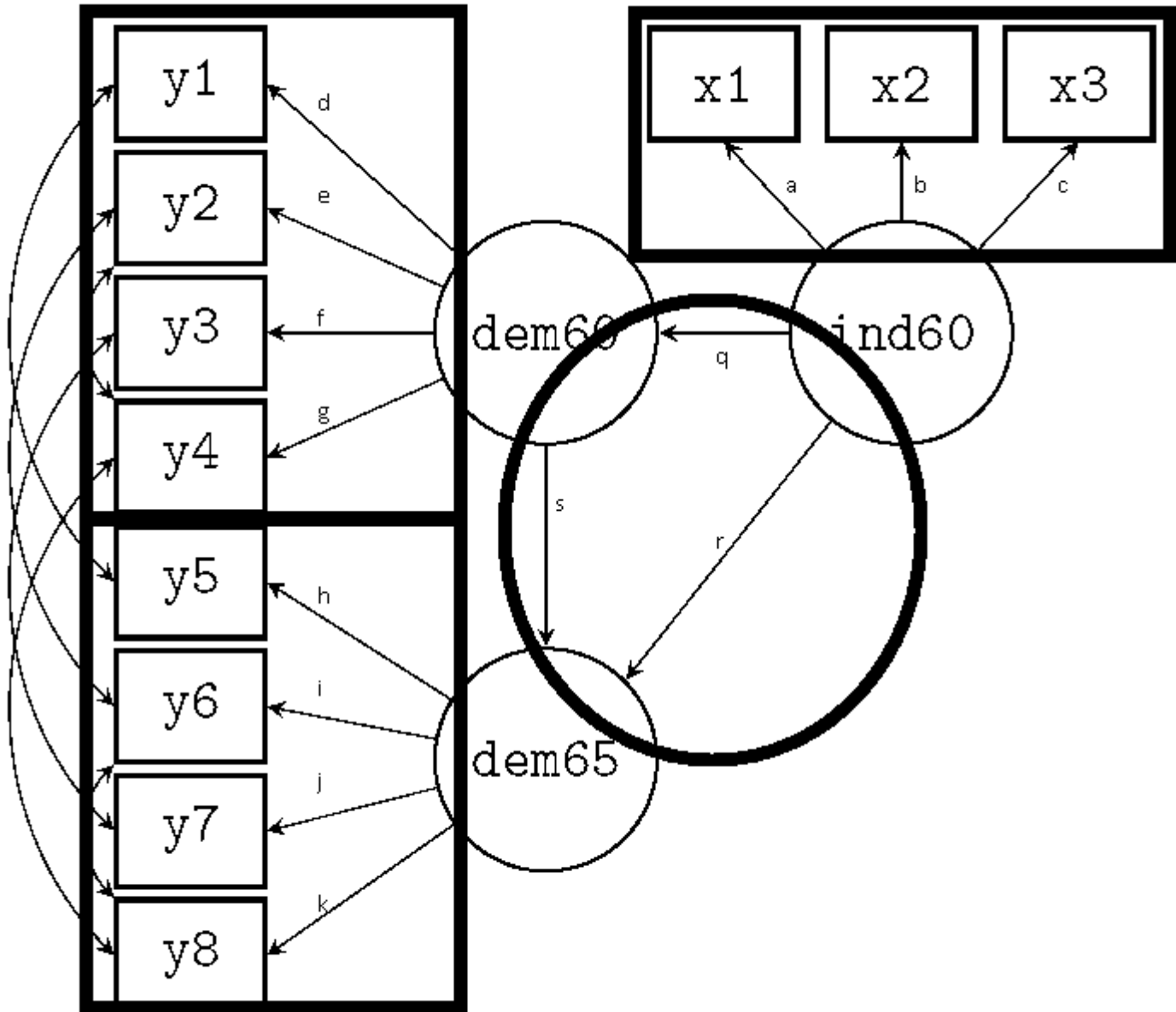
Eigenvalues of Principal Components





Chapter 7: Structural Equation Modeling and Confirmatory Factor Analysis





```

R Console
> summary(mxRun(factorModel.2))
Running Political Democracy Model
data:
$`Political Democracy Model.data`
$`Political Democracy Model.data`$cov
      y1      y2      y3      y4      y5      y6      y7$
y1 6.8785658 6.2513659 5.8388354 6.088643 5.063811 5.7458262 5.8119432$
y2 6.2513659 15.5798162 5.8386258 9.508553 5.603072 9.3862893 7.5354623$
y3 5.8388354 5.8386258 10.7642473 6.687937 4.939027 4.7273963 7.0064347$
y4 6.0886427 9.5085534 6.6879370 11.218932 5.702059 7.4421821 7.4879988$
y5 5.0638108 5.6030723 4.9390271 5.702059 6.825690 4.9768045 5.8213662$
y6 5.7458262 9.3862893 4.7273963 7.442182 4.976805 11.3753251 6.7481190$
y7 5.8119432 7.5354623 7.0064347 7.487999 5.821366 6.7481190 10.7993719$
y8 5.6710576 7.7582132 5.6391391 8.012646 5.339386 8.2468297 7.5924331$
x1 0.7343930 0.6194955 0.7868787 1.150461 1.081626 0.8527889 0.9368373$
x2 1.2733939 1.4912729 1.5519351 2.240957 2.063733 1.8054863 1.9955952$
x3 0.9114547 1.1698131 1.0390796 1.838023 1.583465 1.5720662 1.6259535$

The final iterate satisfies the optimality conditions to the accuracy require$

free parameters:
      name matrix row col Estimate Std.Error $
1 Political Democracy Model.A[10,12] A 10 12 2.18036325 0.13994831 $
2 Political Democracy Model.A[11,12] A 11 12 1.81850749 0.15321581 $
3 Political Democracy Model.A[13,12] A 13 12 1.48299937 0.40003769 $
4 Political Democracy Model.A[14,12] A 14 12 0.57233772 0.23532664 $
5 Political Democracy Model.A[2,13] A 2 13 1.25674159 0.18676663 $
6 Political Democracy Model.A[3,13] A 3 13 1.05771456 0.14934347 $
7 Political Democracy Model.A[4,13] A 4 13 1.26478130 0.15176605 $
8 Political Democracy Model.A[14,13] A 14 13 0.83734401 0.09946018 $
9 Political Democracy Model.A[6,14] A 6 14 1.18569339 0.17251176 $
10 Political Democracy Model.A[7,14] A 7 14 1.27950966 0.16146294 $

```

```
Open script
R Console
> summary(fit.hs.lavaan)
lavaan (0.5-15) converged normally after 22 iterations

Number of observations              301

Estimator                          ML
Minimum Function Test Statistic    85.306
Degrees of freedom                  24
P-value (Chi-square)               0.000

Parameter estimates:

Information                          Expected
Standard Errors                      Standard

Estimate  Std.err  Z-value  P(>|z|)
Latent variables:
visual =~
  x1      0.900   0.081   11.127   0.000
  x2      0.498   0.077    6.429   0.000
  x3      0.656   0.074    8.817   0.000
textual =~
  x4      0.990   0.057   17.474   0.000
  x5      1.102   0.063   17.576   0.000
  x6      0.917   0.054   17.082   0.000
speed =~
  x7      0.619   0.070    8.903   0.000
  x8      0.731   0.066   11.090   0.000
  x9      0.670   0.065   10.305   0.000

Covariances:
visual ~~
  textual  0.459   0.064    7.189   0.000
```

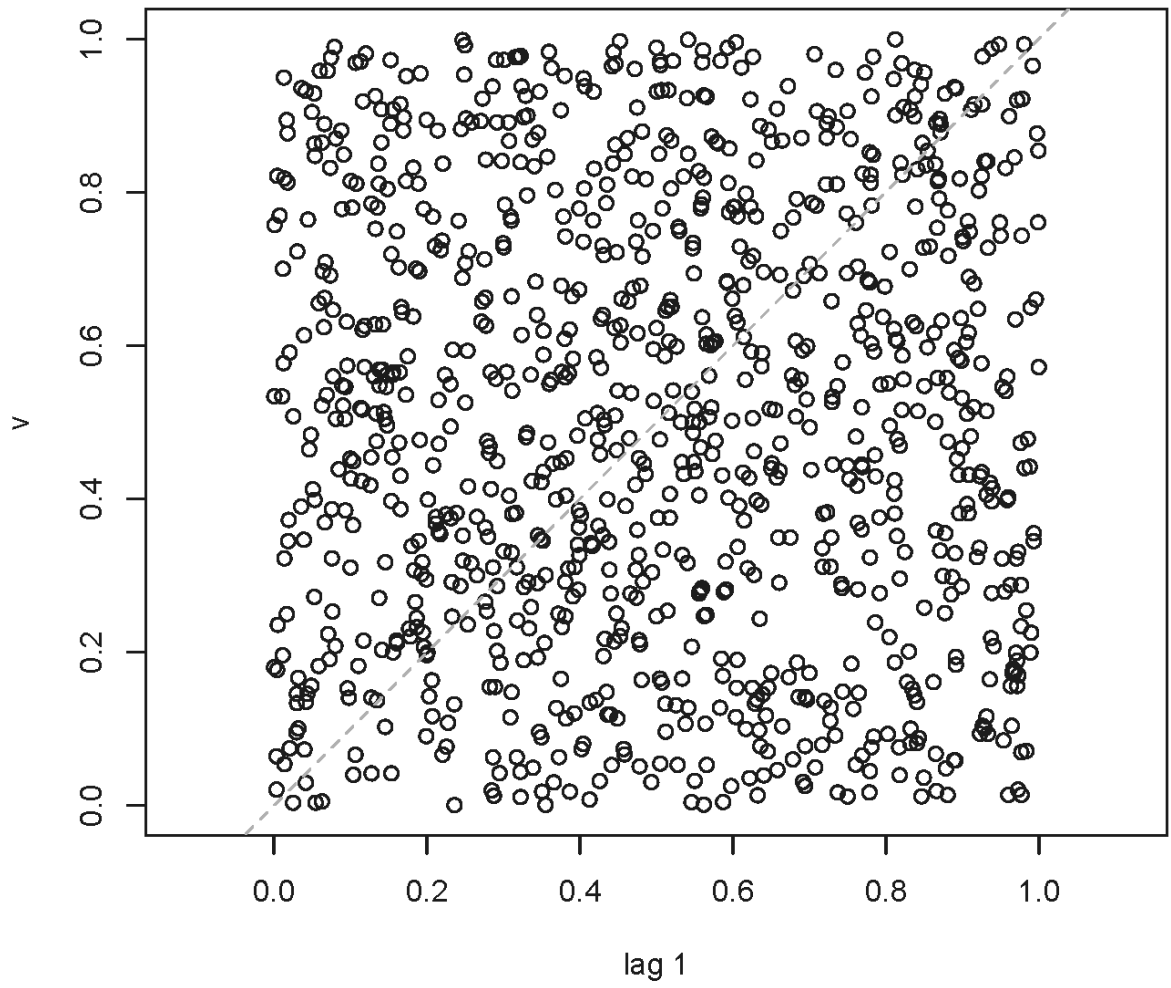
```

R Console
> summary(fit.hs.open.mx)
data:
$`Holzinger Swineford.data`
$`Holzinger Swineford.data`$cov
      x1      x2      x3      x4      x5      x6      $
x1 1.36289774 0.40872923 0.58183232 0.5065178 0.4420843 0.4563242 0.085047$
x2 0.40872923 1.38638981 0.45256748 0.2096198 0.2117947 0.2483697 -0.097073$
x3 0.58183232 0.45256748 1.27911441 0.2088635 0.1126706 0.2449227 0.088636$
x4 0.50651778 0.20961978 0.20886351 1.3551667 1.1014120 0.8985008 0.220475$
x5 0.44208426 0.21179471 0.11267061 1.1014120 1.6653184 1.0179058 0.143476$
x6 0.45632416 0.24836972 0.24492268 0.8985008 1.0179058 1.2003462 0.144558$
x7 0.08504799 -0.09707352 0.08863631 0.2204751 0.1434762 0.1445587 1.187083$
x8 0.26471714 0.11002492 0.21303038 0.1260120 0.1812072 0.1659824 0.537029$
x9 0.45986634 0.24482282 0.37509875 0.2441739 0.2962255 0.2367836 0.374541$

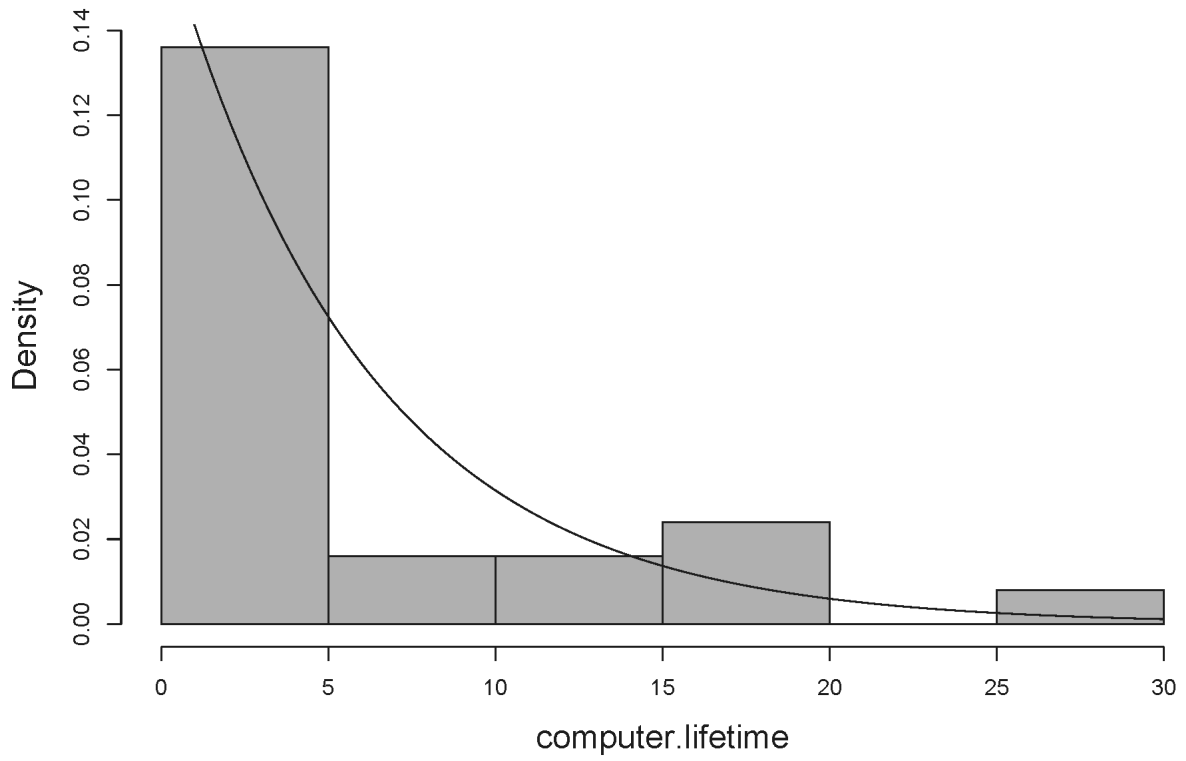
free parameters:
      name matrix      row      col Estimate Std.Error $
1 Holzinger Swineford.A[1,10] A      x1 visual 0.9011177 0.08351138 $
2 Holzinger Swineford.A[2,10] A      x2 visual 0.4987688 0.08105689 $
3 Holzinger Swineford.A[3,10] A      x3 visual 0.6572487 0.07783756 $
4 Holzinger Swineford.A[4,11] A      x4 textual 0.9913408 0.05687873 $
5 Holzinger Swineford.A[5,11] A      x5 textual 1.1034381 0.06279691 $
6 Holzinger Swineford.A[6,11] A      x6 textual 0.9181265 0.05393494 $
7 Holzinger Swineford.A[7,12] A      x7 speed 0.6205055 0.07455608 $
8 Holzinger Swineford.A[8,12] A      x8 speed 0.7321655 0.07574673 $
9 Holzinger Swineford.A[9,12] A      x9 speed 0.6710954 0.07778219 $
10 Holzinger Swineford.S[1,1] S      x1 x1 0.5508846 0.11964676 $
11 Holzinger Swineford.S[2,2] S      x2 x2 1.1376195 0.10478550 $
12 Holzinger Swineford.S[3,3] S      x3 x3 0.8471385 0.09555064 $
13 Holzinger Swineford.S[4,4] S      x4 x4 0.3724102 0.04820283 $
14 Holzinger Swineford.S[5,5] S      x5 x5 0.4477426 0.05822352 $
15 Holzinger Swineford.S[6,6] S      x6 x6 0.3573899 0.04365787 $

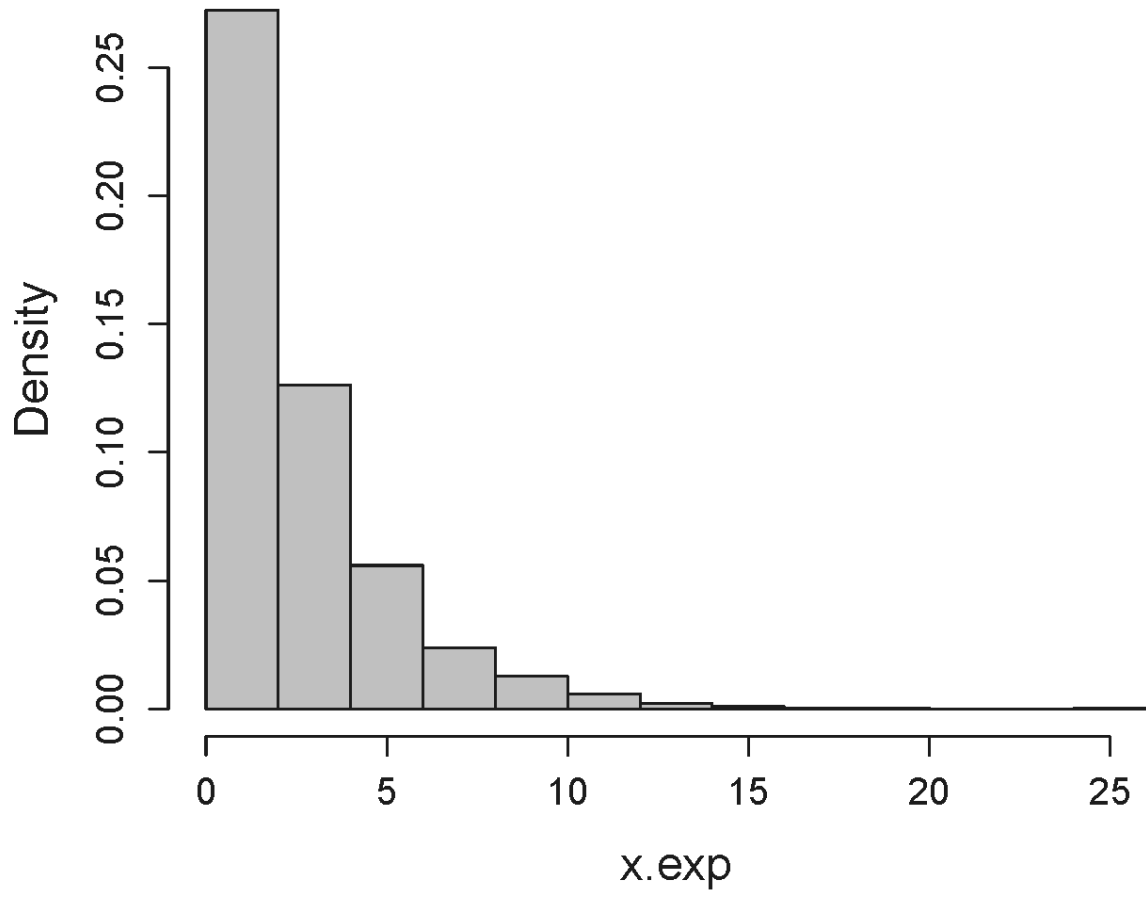
```

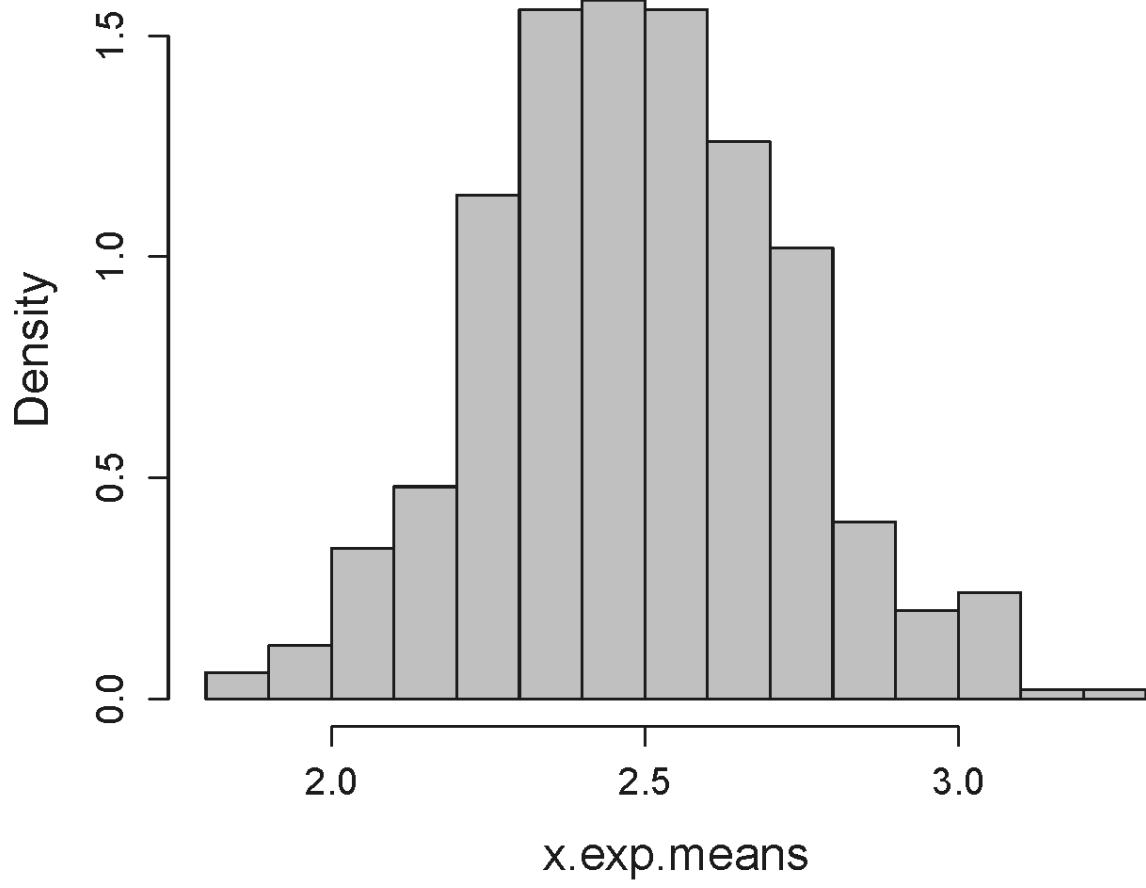
Chapter 8: Simulations



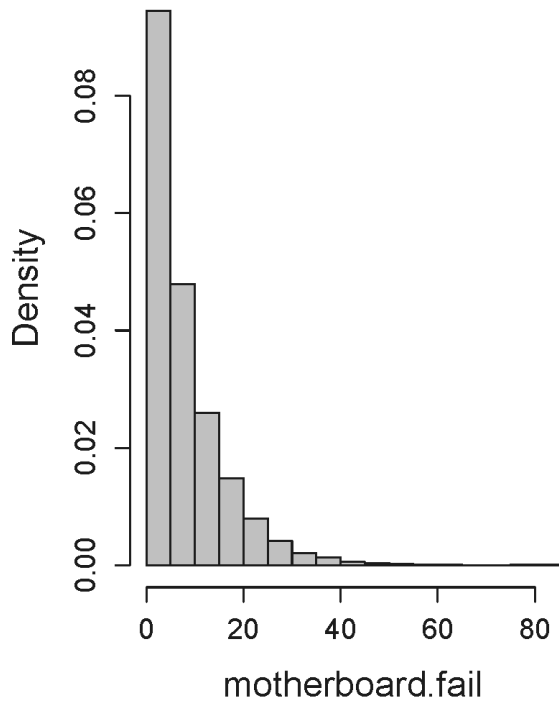
Exponential curve for computers with a mean time to failure of 6 years



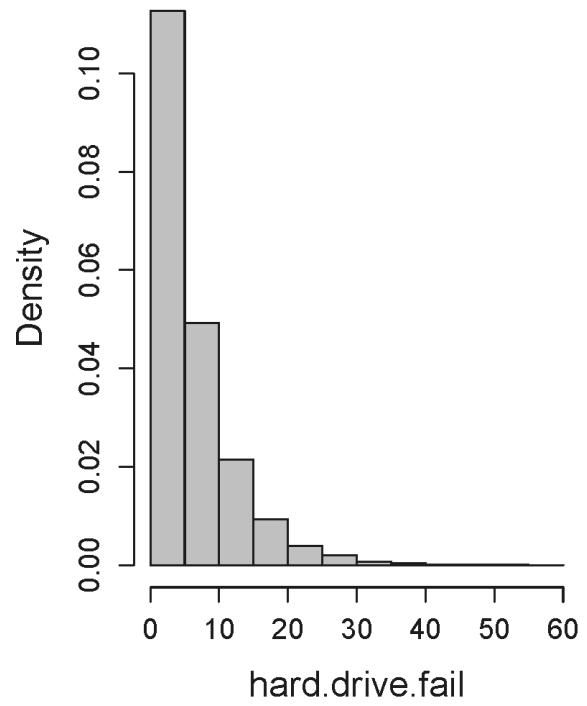


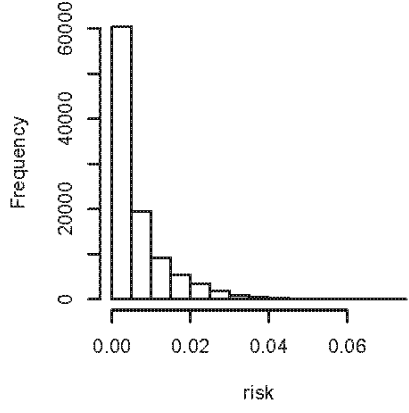
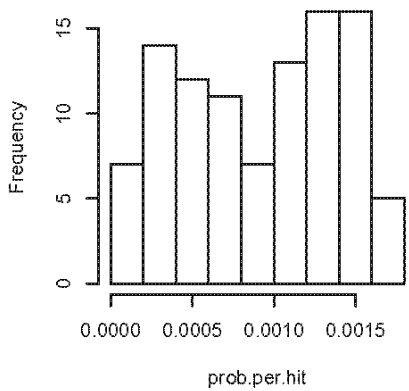
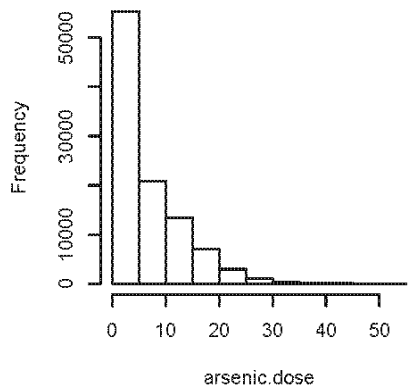
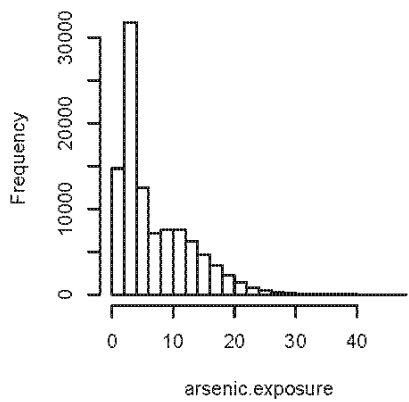
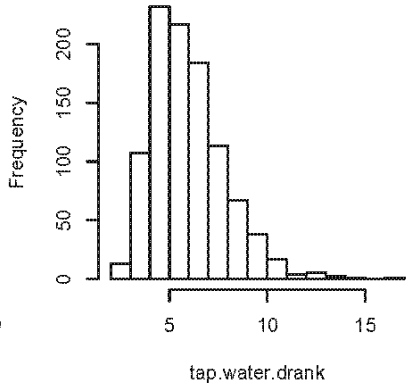
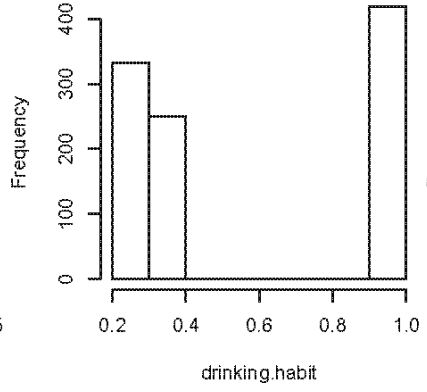
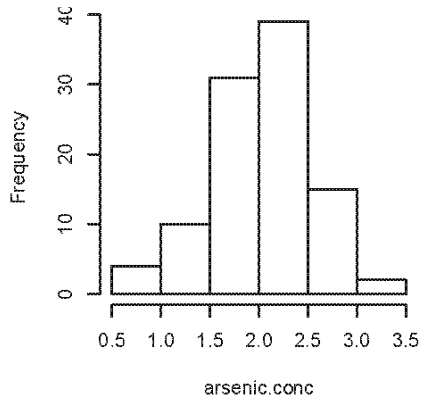


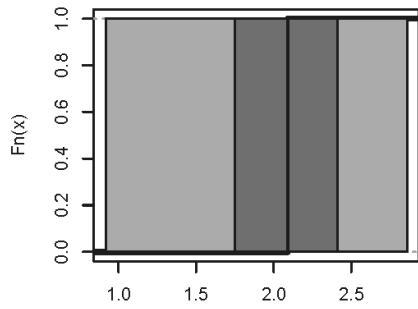
Simulated motherboard time to failure



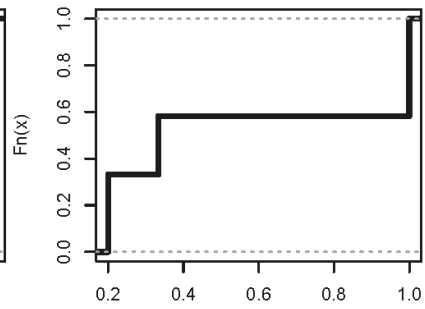
Simulated hard drive time to failure



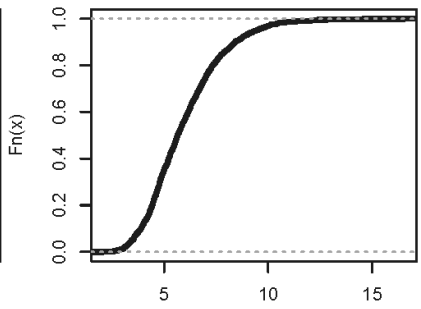




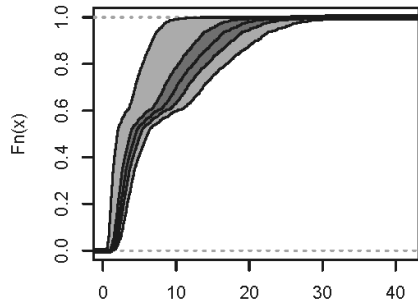
arsenic.conc



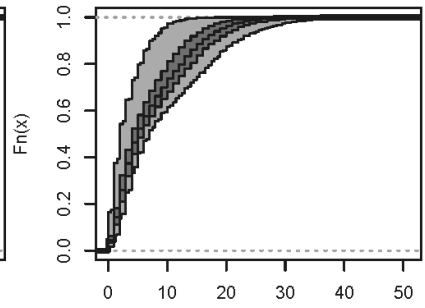
drinking.habit



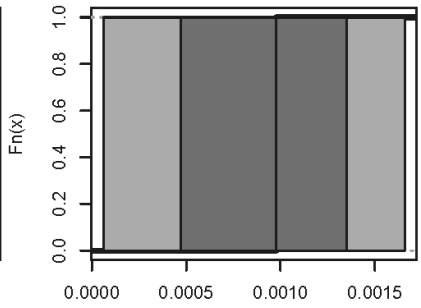
tap.water.drunk



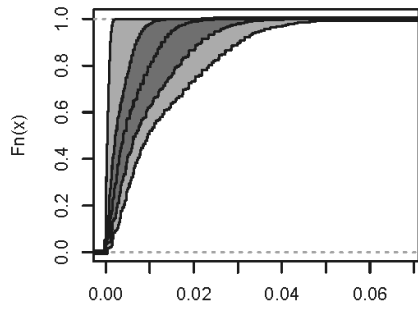
arsenic.exposure



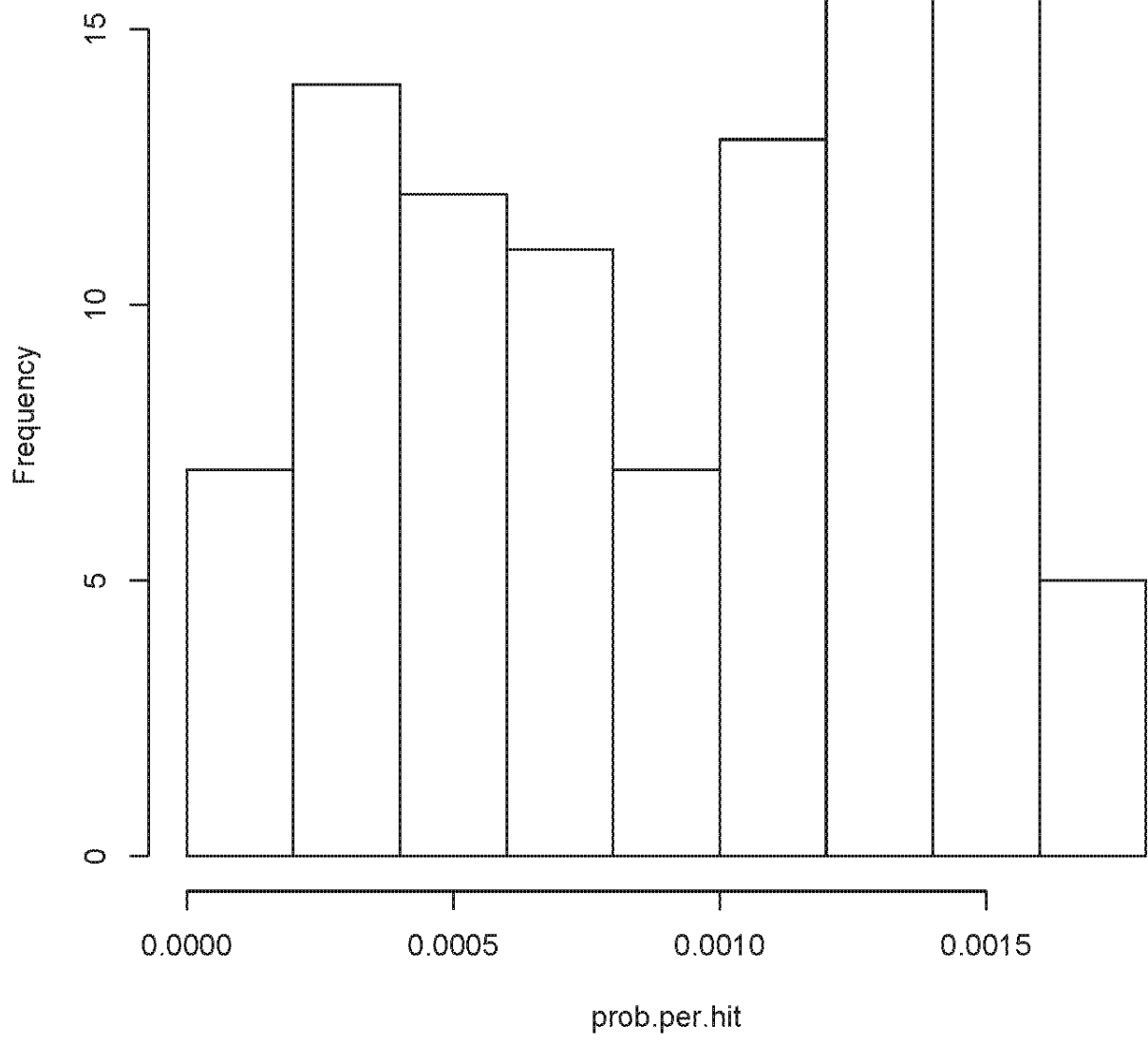
arsenic.dose

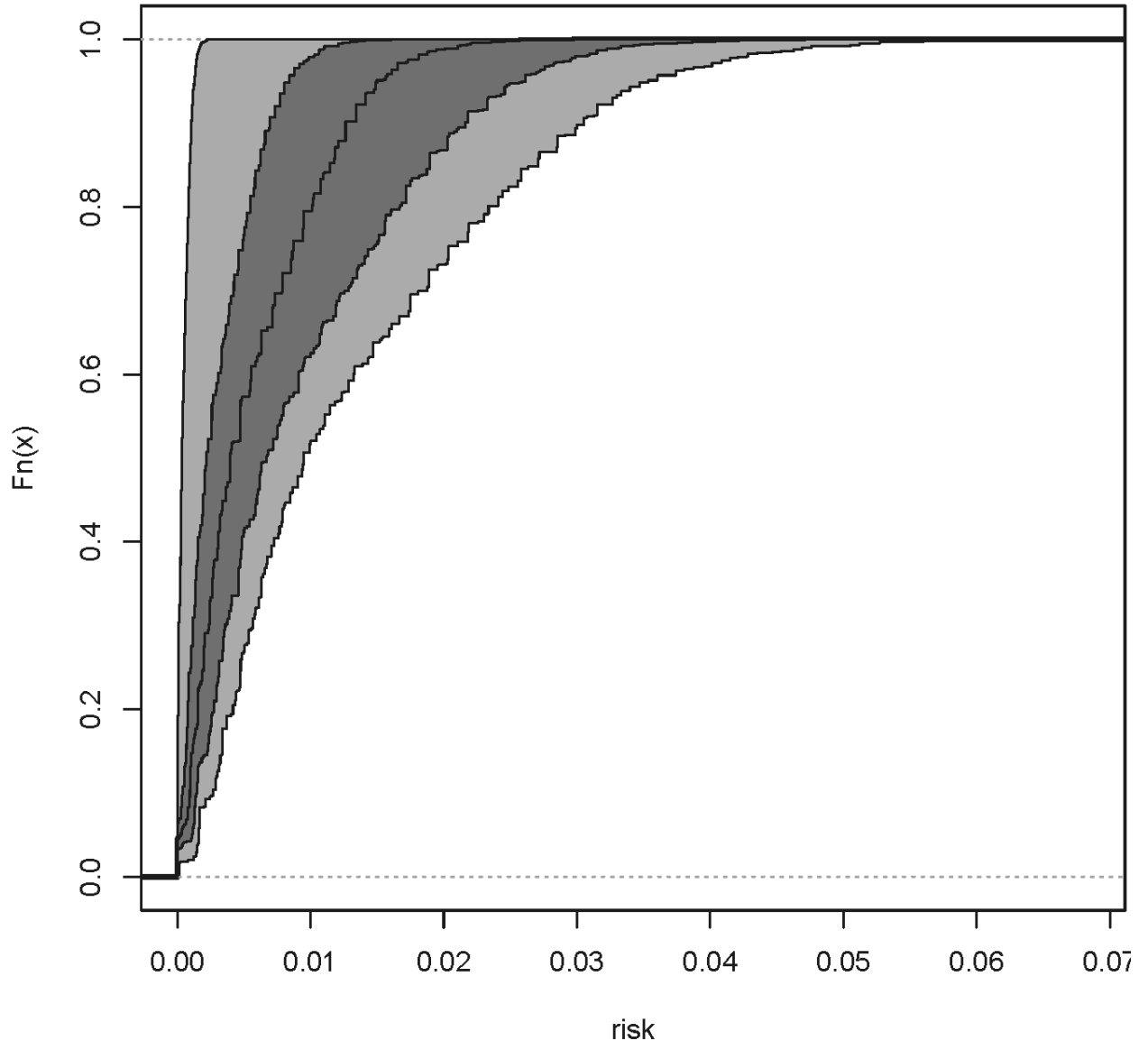


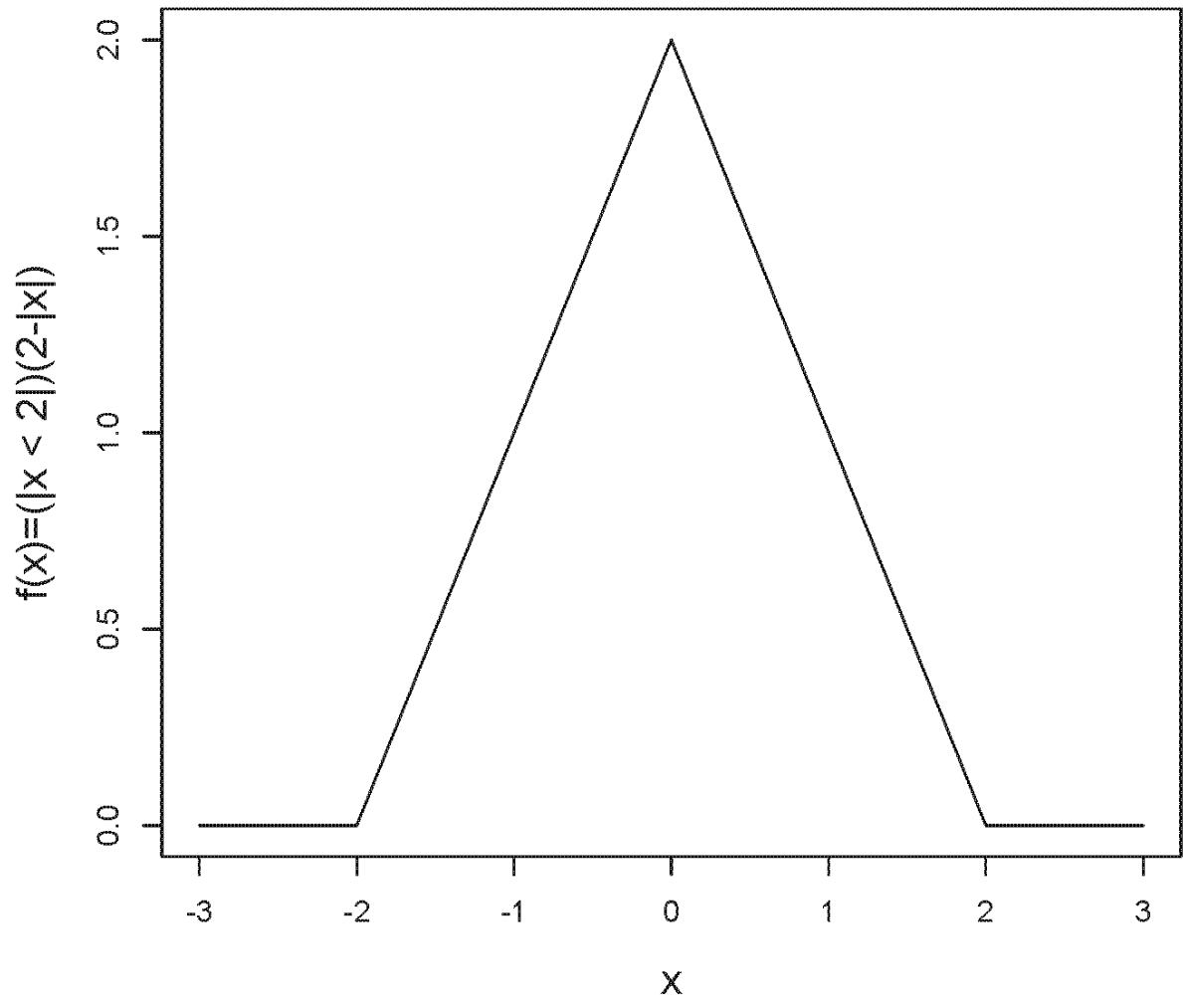
prob.per.hit

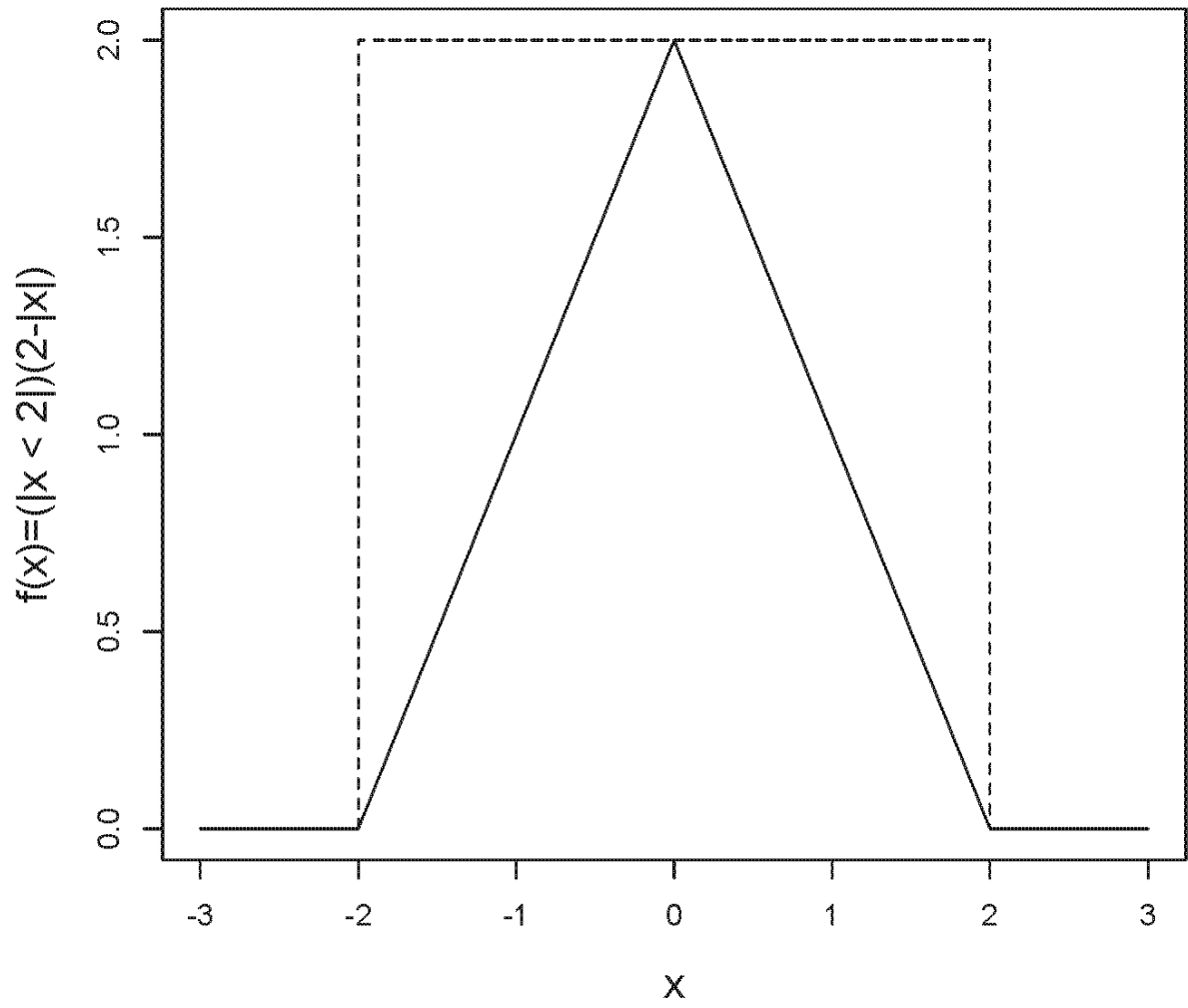


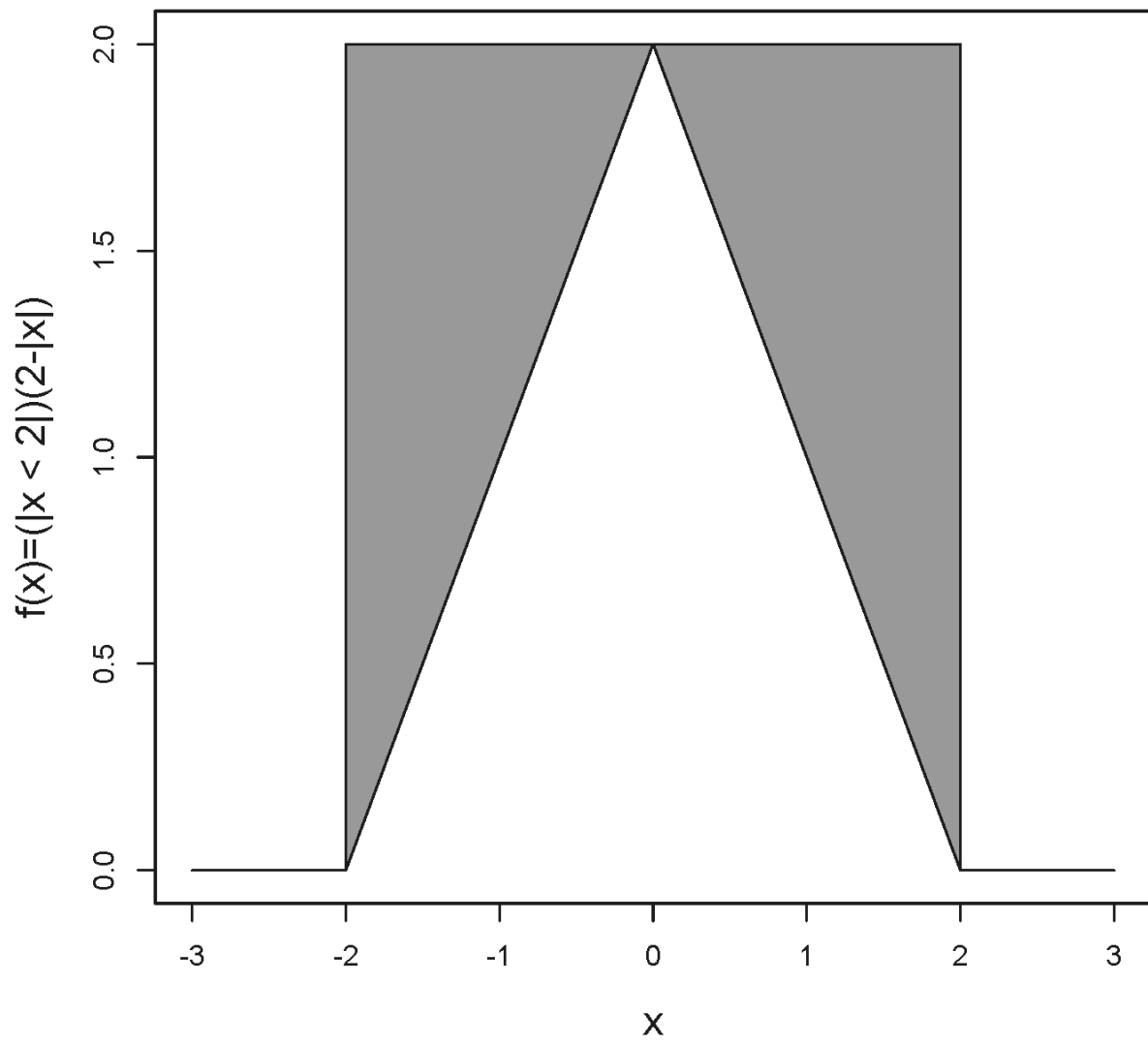
risk

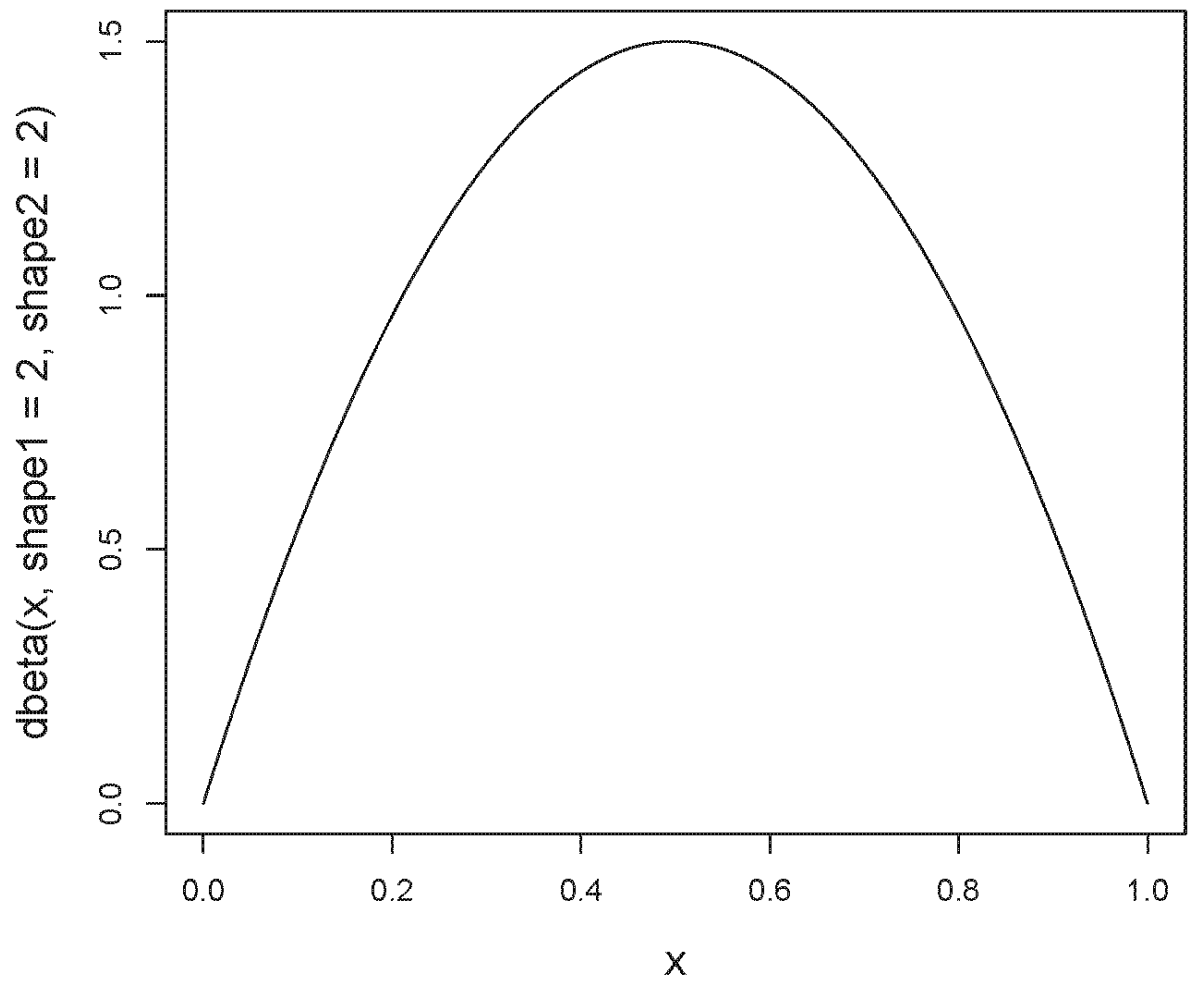


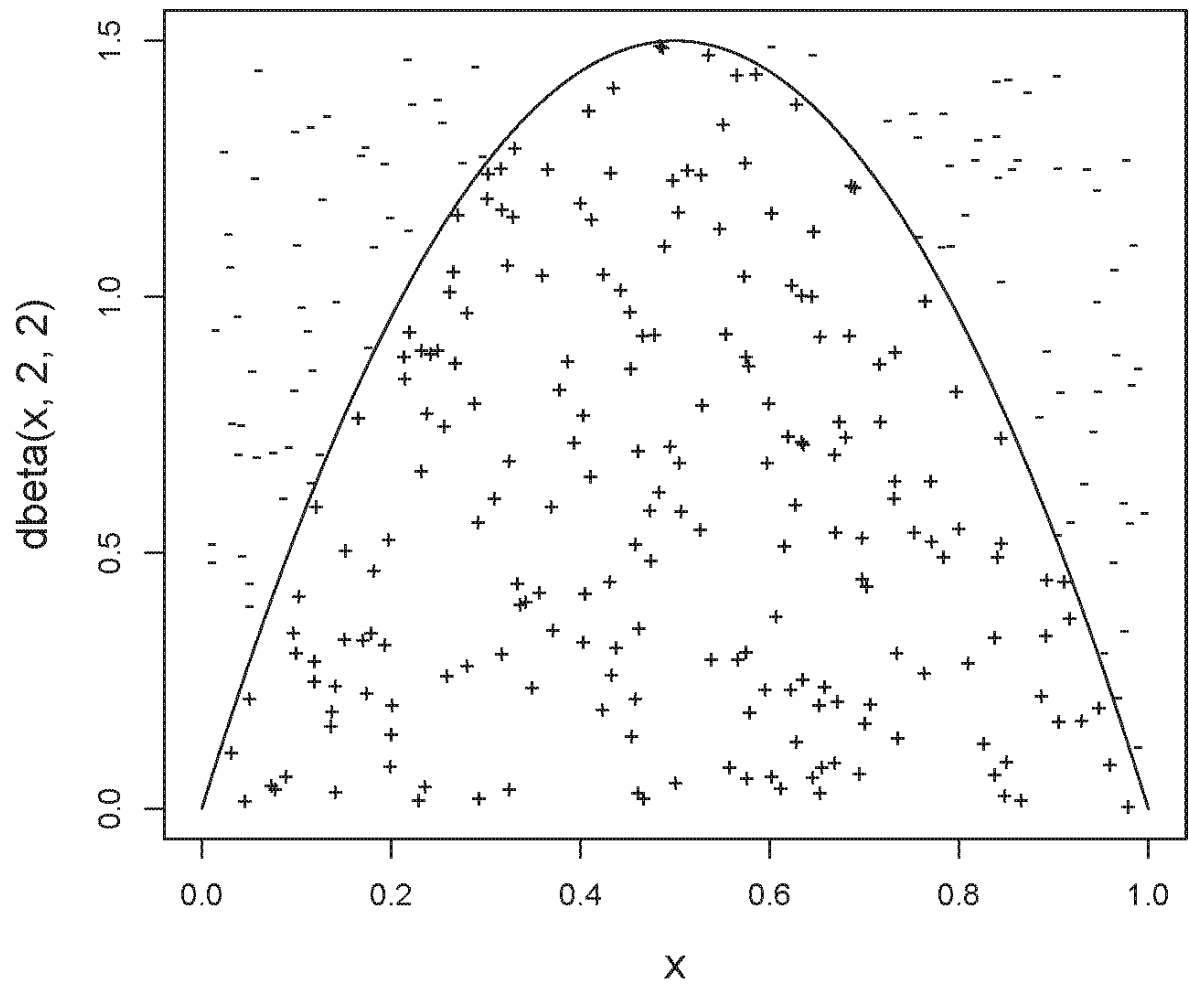




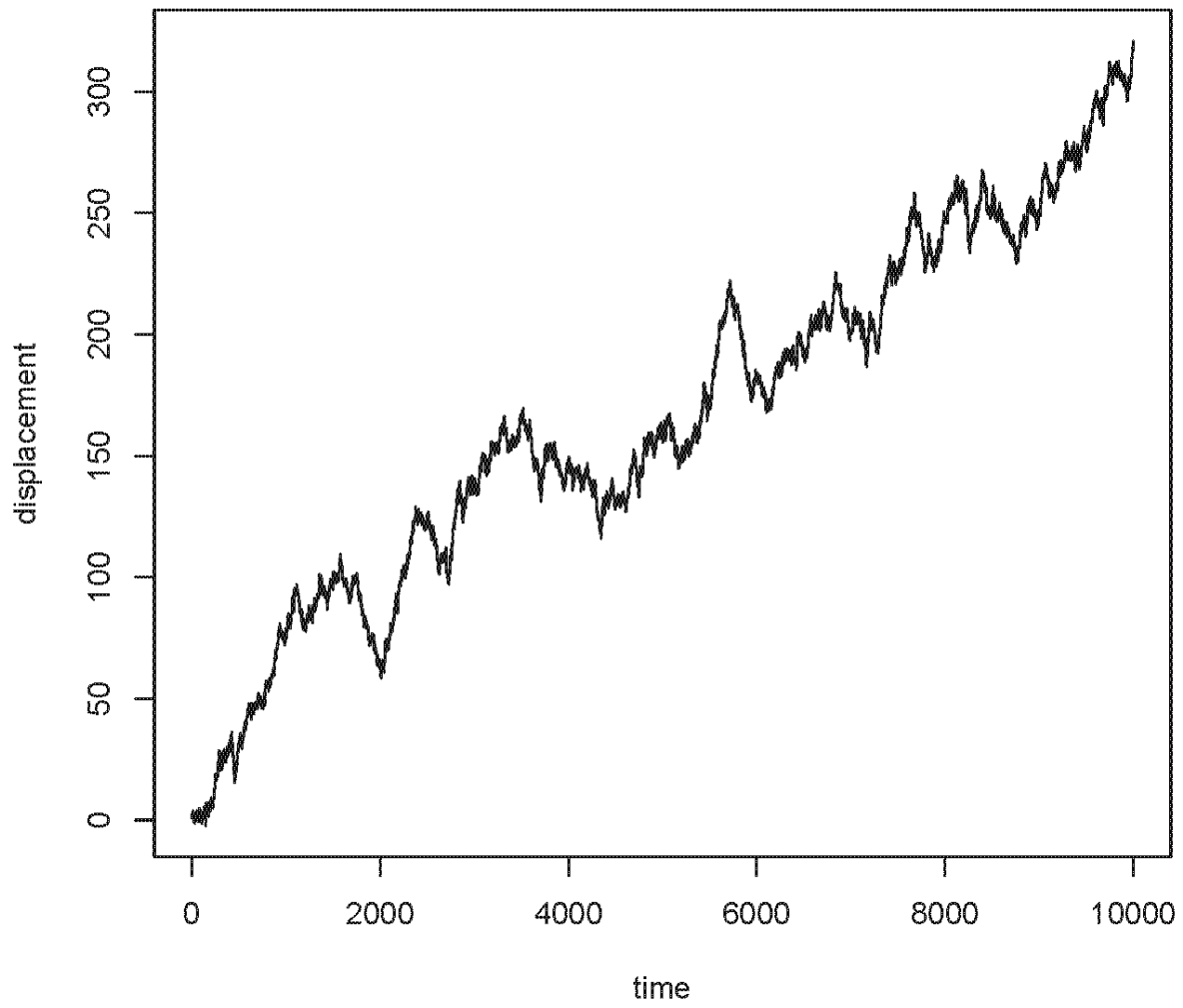




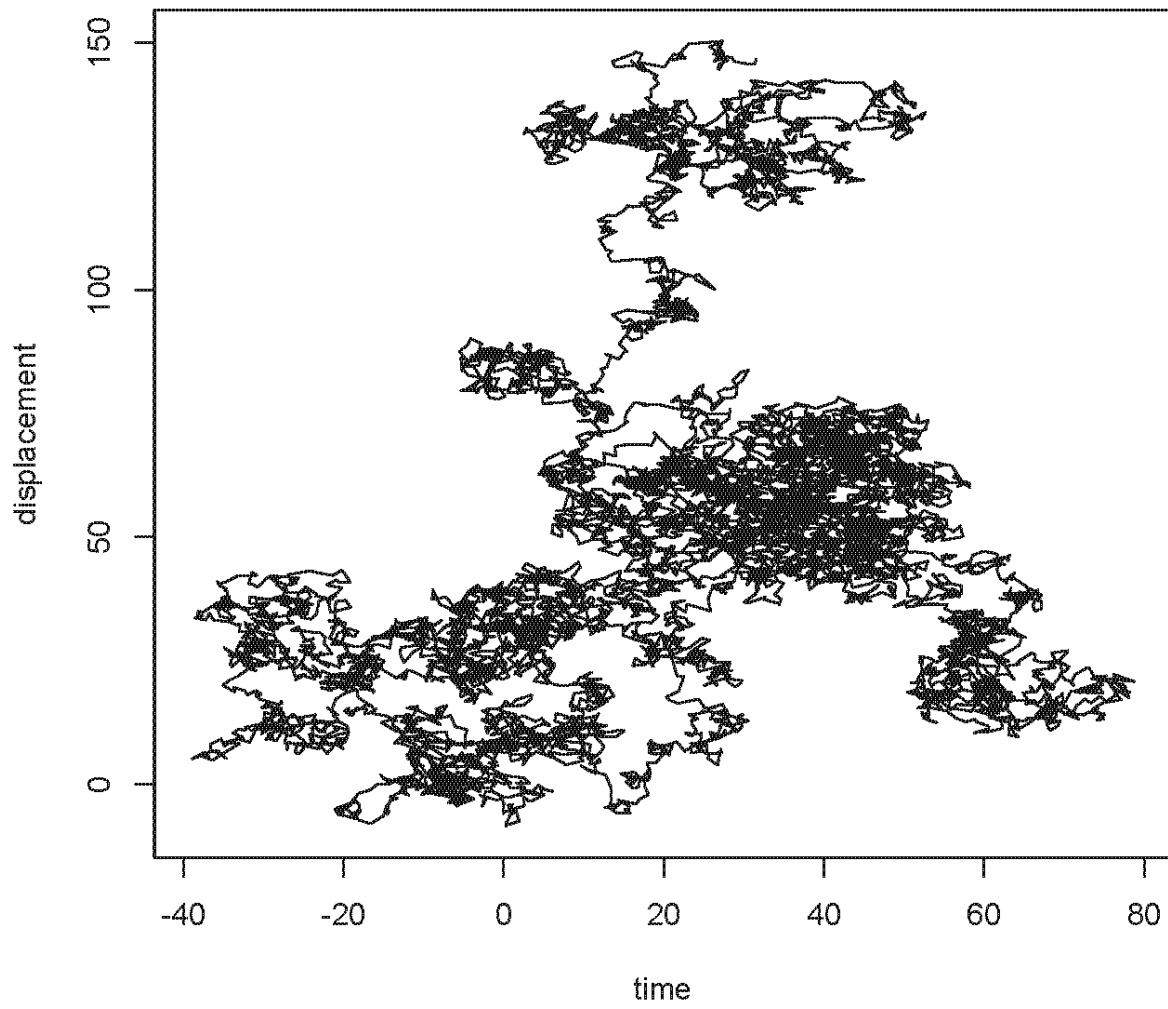


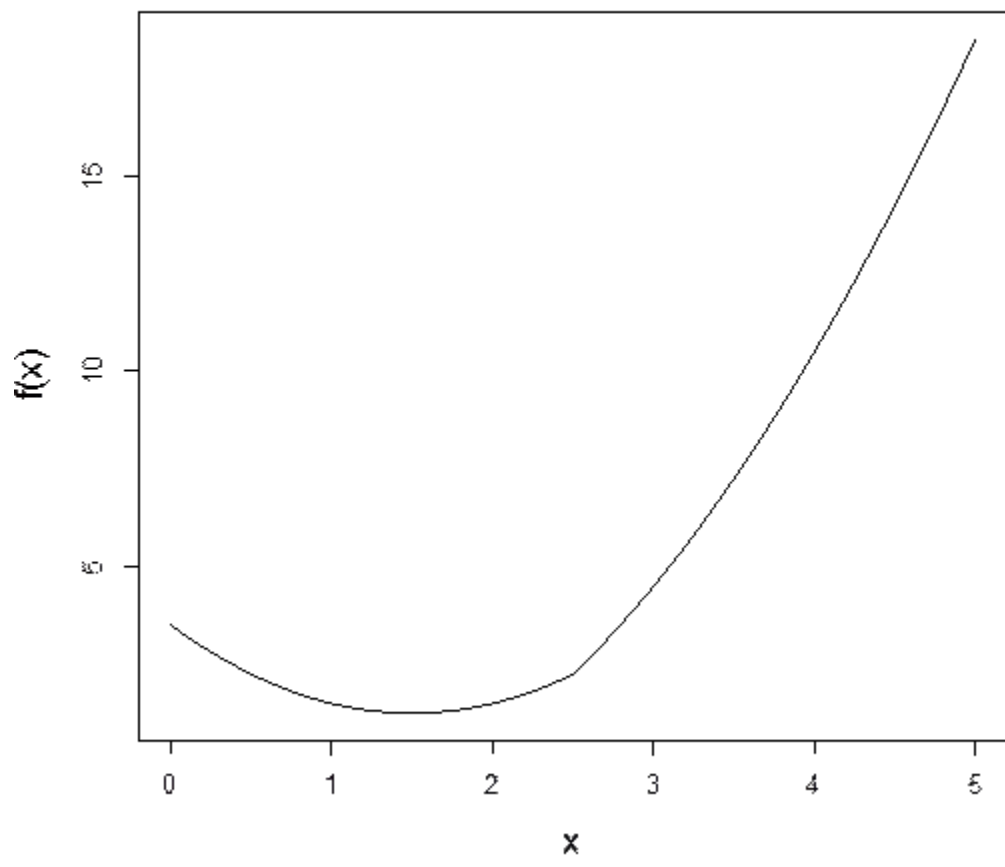


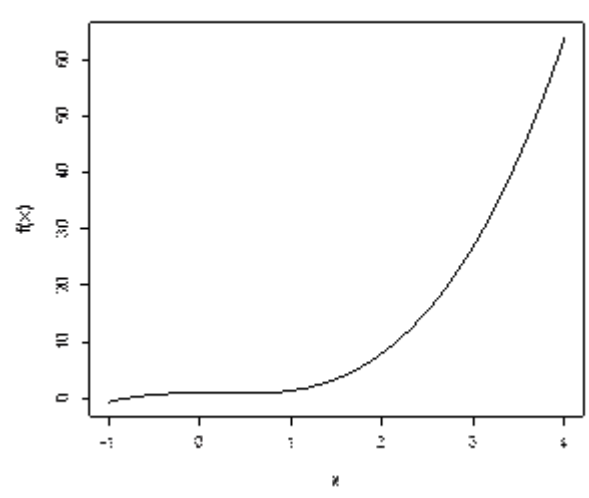
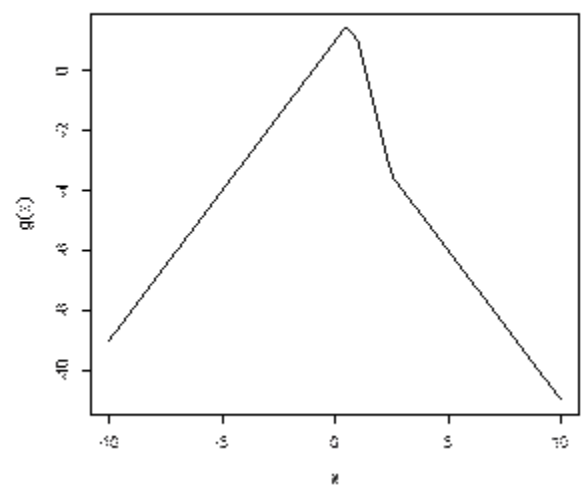
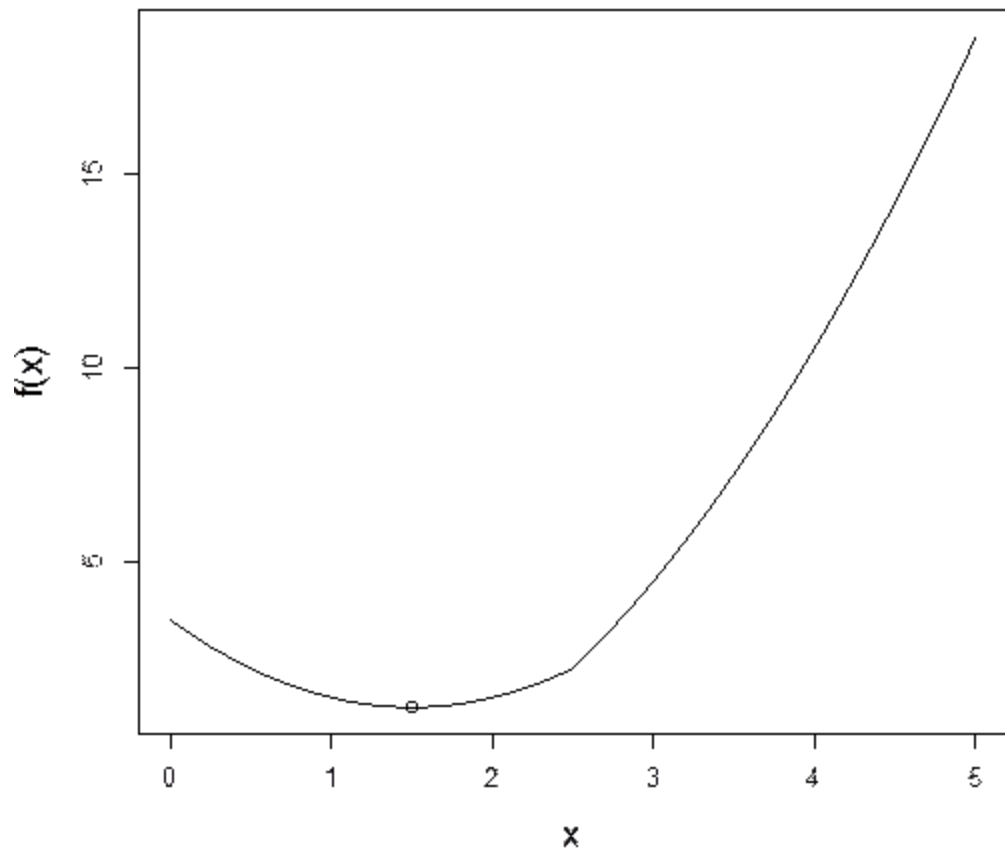
Brownian Motion in 1-Dimension

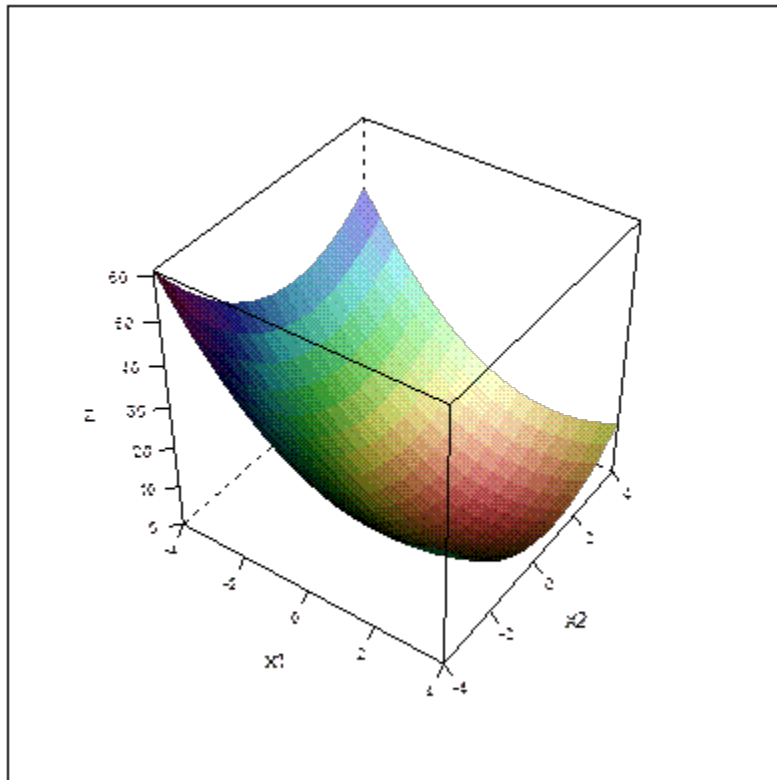
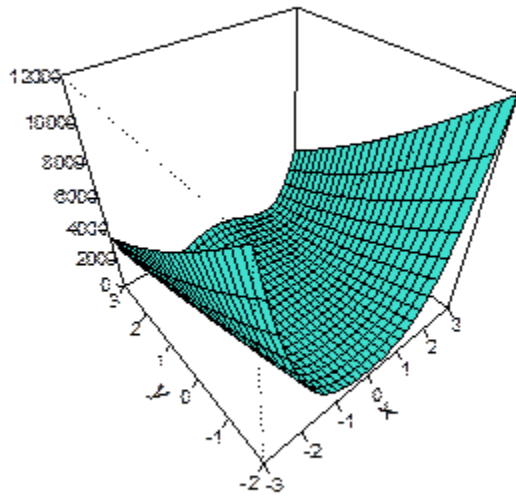


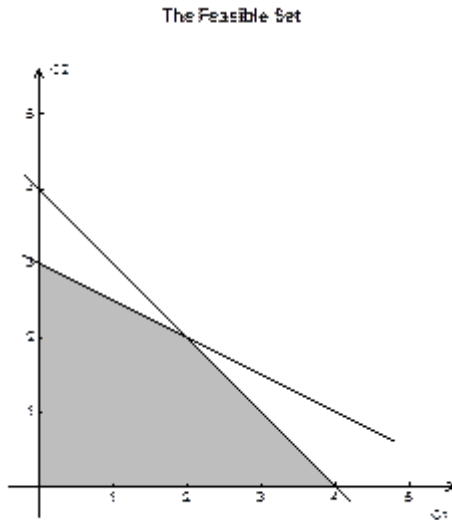
Brownian Motion in 2-Dimensions





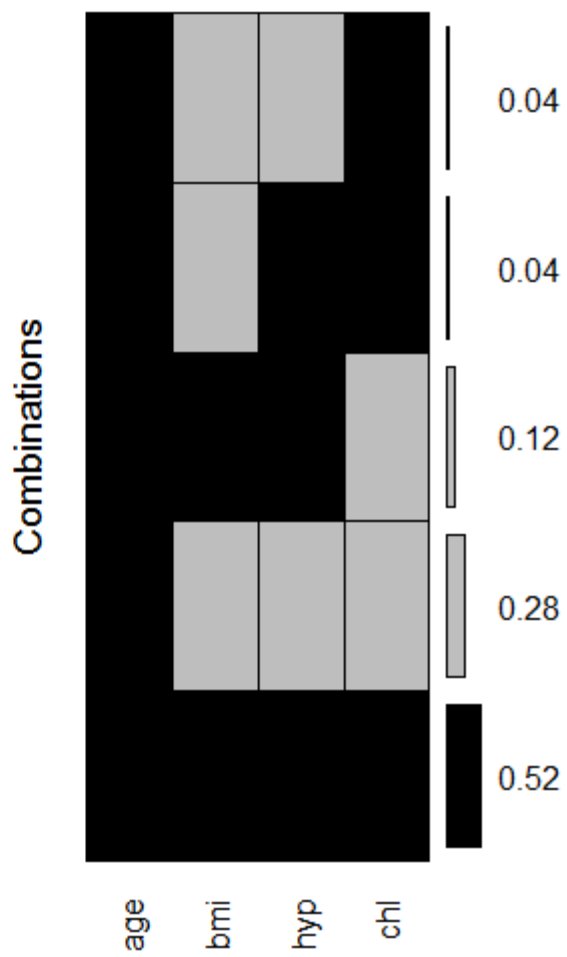
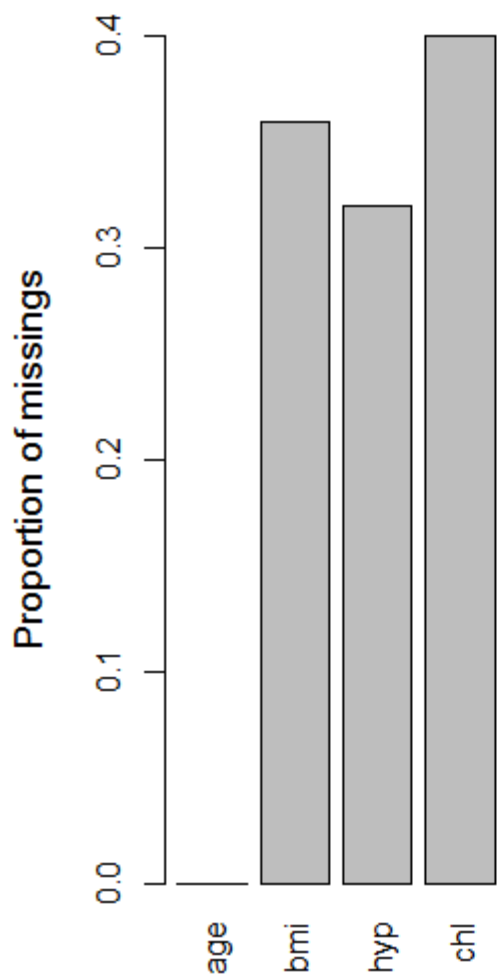




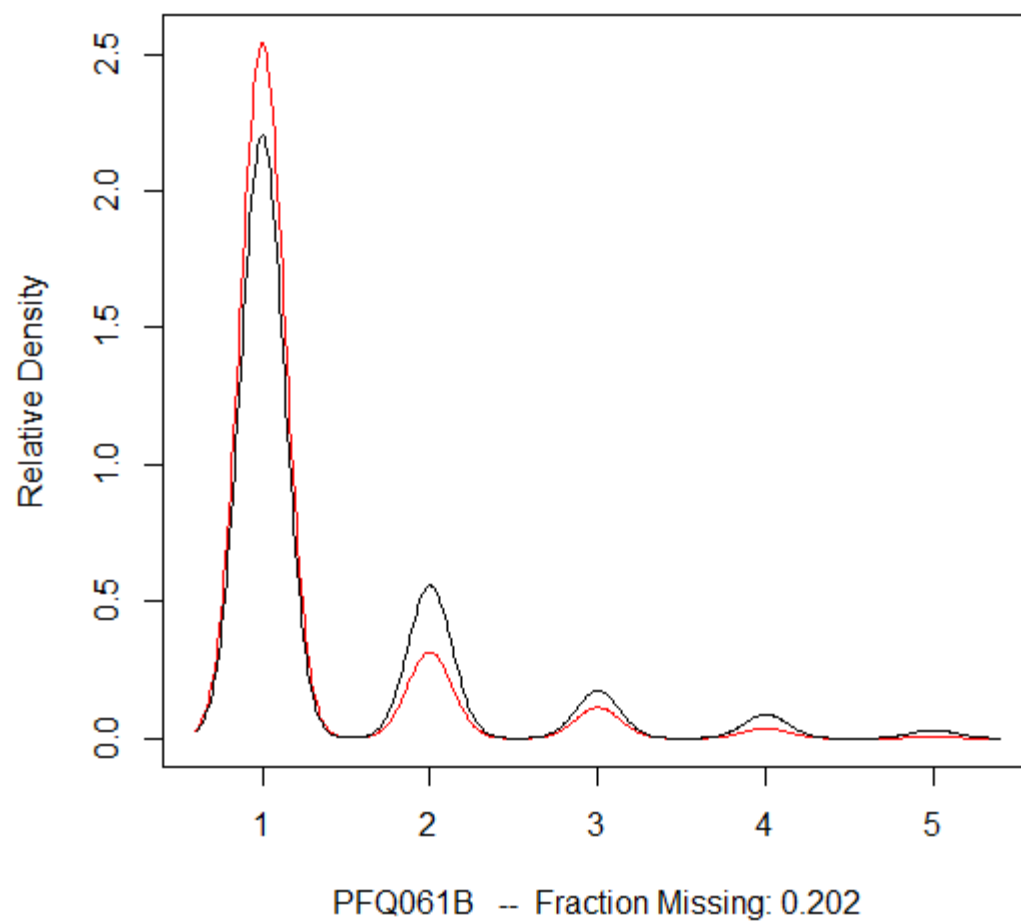


Chapter 10: Advanced Data Management

```
R Console
>
> system.time(a.binary <- condense.to.binary(rep(physical.data[,1],20)))
  user system elapsed
11.80  0.14  11.95
> system.time(a.binary <- condense.to.binary.2(rep(physical.data[,1],20)))
  user system elapsed
 0.31  0.00  0.31
> system.time(a.binary <- condense.to.binary.3(rep(physical.data[,1],20)))
  user system elapsed
  9.7    0.0    9.7
> |
```

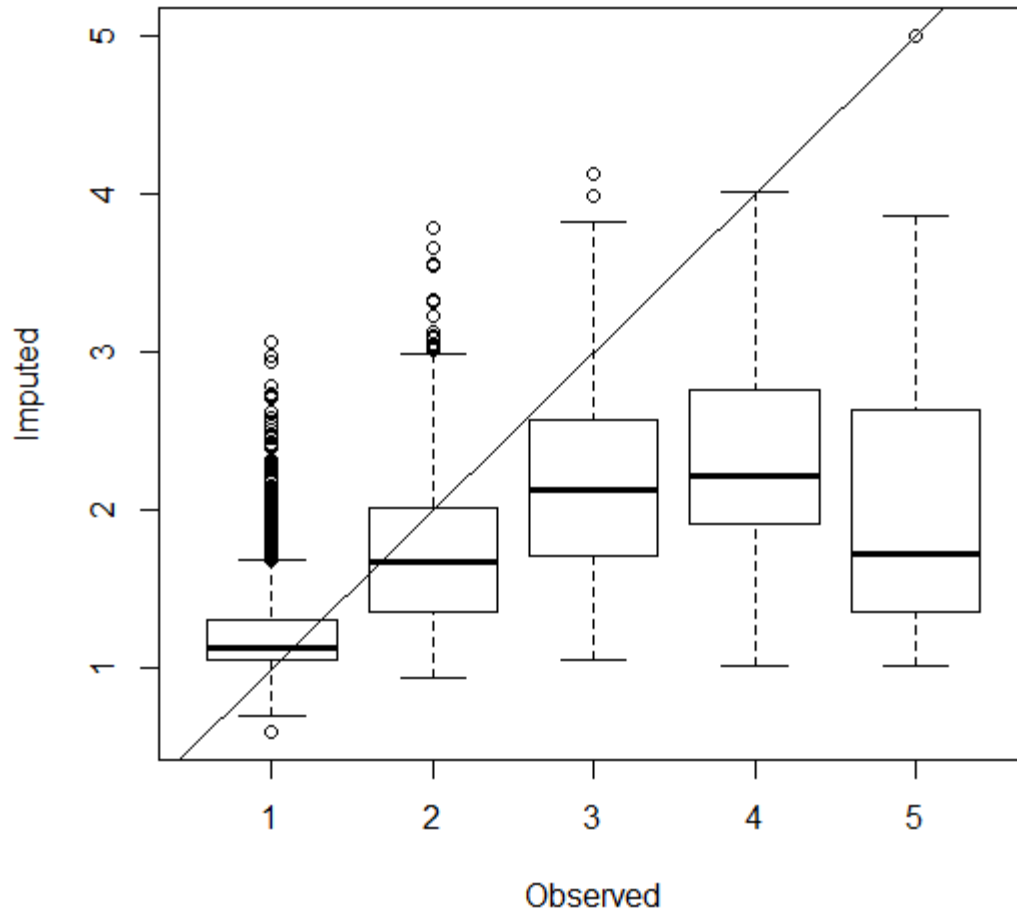


Observed and Imputed values of PFQ061B



```
R R Console
> B.ov.imp <- overimpute(phys.imp, var = 'PFQ061B')
> summary(B.ov.imp)
      row      orig  mean.overimputed
Min.   : 1  Min.   :1.000  Min.   :0.5973
1st Qu.:1134 1st Qu.:1.000  1st Qu.:1.0737
Median :2260 Median :1.000  Median :1.2195
Mean   :2267 Mean   :1.422  Mean   :1.4204
3rd Qu.:3399 3rd Qu.:2.000  3rd Qu.:1.6093
Max.   :4528 Max.   :5.000  Max.   :4.8089
lower.overimputed upper.overimputed prcntmiss
Min.   :-0.69836  Min.   :1.615  Min.   :0.05
1st Qu.: 0.01989  1st Qu.:2.063  1st Qu.:0.20
Median : 0.18296  Median :2.248  Median :0.25
Mean   : 0.35850  Mean   :2.419  Mean   :0.24
3rd Qu.: 0.55965  3rd Qu.:2.624  3rd Qu.:0.30
Max.   : 3.26388  Max.   :6.004  Max.   :0.60
> |
```

Overimputed vs Observed Values



```

R Console
> imputation.means <- sapply(phys.imp$imputations, colMeans)
> imputation.means
      imp1      imp2      imp3      imp4      imp5
PFQ061A 1.314267 1.309629 1.316917 1.312721 1.318463
PFQ061B 1.456714 1.448322 1.441034 1.445230 1.440150
PFQ061C 1.272747 1.282906 1.277164 1.275398 1.275177
PFQ061D 1.612191 1.616829 1.606228 1.610424 1.614620
PFQ061E 1.399956 1.396422 1.389355 1.383834 1.386263
PFQ061F 1.378534 1.385601 1.374337 1.382730 1.384055
PFQ061G 1.292624 1.299691 1.290194 1.289532 1.289532
PFQ061H 1.045936 1.040415 1.045716 1.041961 1.043507
PFQ061I 1.199205 1.212898 1.204726 1.207597 1.199647
PFQ061J 1.170936 1.169390 1.169832 1.171820 1.170053
PFQ061K 1.060733 1.059408 1.065592 1.062058 1.061617
PFQ061L 1.119258 1.130521 1.127208 1.127208 1.126767
PFQ061M 1.667182 1.668507 1.677120 1.675574 1.681316
PFQ061N 1.321334 1.312500 1.309629 1.317800 1.317800
PFQ061O 1.231670 1.234320 1.227473 1.230124 1.224602
PFQ061P 1.187058 1.177783 1.191254 1.184408 1.184850
PFQ061Q 1.302562 1.313383 1.310512 1.306095 1.303224
PFQ061R 1.319788 1.322659 1.323542 1.319125 1.310733
PFQ061S 1.088339 1.084806 1.080389 1.086572 1.084143
PFQ061T 1.636042 1.626325 1.624337 1.619920 1.633834
> rowMeans(imputation.means)
PFQ061A PFQ061B PFQ061C PFQ061D PFQ061E PFQ061F PFQ061G PFQ061H
1.314399 1.446290 1.276678 1.612058 1.391166 1.381051 1.292314 1.043507
PFQ061I PFQ061J PFQ061K PFQ061L PFQ061M PFQ061N PFQ061O PFQ061P
1.204814 1.170406 1.061882 1.126193 1.673940 1.315813 1.229638 1.185071
PFQ061Q PFQ061R PFQ061S PFQ061T
1.307155 1.319170 1.084850 1.628092
> |

```

```

R Console
>
> legs.v.arms.models <- with(imputed.phys.func, lm( I(B+C+D+H+I+J+M+N) ~ I(E+F+G$
> leg.v.arm.pool <- pool(legs.v.arms.models)
> summary(leg.v.arm.pool)
              est          se          t          df
(Intercept)  4.5689783 0.14504376 31.50069 145.12519
I(E + F + G + K + L + O + P + T) 0.5965278 0.01409243 42.32966 83.49253
              Pr(>|t|)      lo 95      hi 95      nmis      fmi
(Intercept)                0 4.2823073 4.8556494    NA 0.1741114
I(E + F + G + K + L + O + P + T) 0 0.5685009 0.6245546    NA 0.2344096
              lambda
(Intercept)  0.1628075
I(E + F + G + K + L + O + P + T) 0.2162875
> |

```