

# Lab1 - Jin Kweon (3032235207)

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```
#matrix -> rows = number of objects/observations & cols = number of features/variables
```

```
"+"(2,3) #fancy version of summation
```

```
## [1] 5
```

```
apropos("log") #list out all the functions that include "log"
```

```
## [1] ".__C__logical"      ".__C__logLik"
## [3] ".__T__Logic:base"   "as.data.frame.logical"
## [5] "as.logical"         "as.logical.factor"
## [7] "dlogis"             "is.logical"
## [9] "log"                "log10"
## [11] "log1p"              "log2"
## [13] "logb"               "Logic"
## [15] "logical"            "logLik"
## [17] "loglin"             "plogis"
## [19] "qlogis"             "rlogis"
## [21] "SSlogis"
```

```
x <- 1:9 #vector
```

```
mat_x <- matrix(x, 3, 3) #by column
```

```
mat_x2 <- matrix(x, 3, 3, byrow = T) #by row
```

```
diag(5) #Create an 5*5 identity matrix
```

```
##      [,1] [,2] [,3] [,4] [,5]
## [1,]    1    0    0    0    0
## [2,]    0    1    0    0    0
## [3,]    0    0    1    0    0
## [4,]    0    0    0    1    0
## [5,]    0    0    0    0    1
```

```
#create three vectors and combine it together to form a matrix.
```

```
a1 <- c(2, 3, 6, 7, 10)
```

```
a2 <- c(1.88, 2.05, 1.70, 1.60, 1.78)
```

```
a3 <- c(80, 90, 70, 50, 75)
```

```
A <- cbind(a1, a2, a3)
```

```
#create three vectors and combine it together to form a matrix.
```

```
b1 <- c(1, 4, 5, 8, 9)
```

```
b2 <- c(1.22, 1.05, 3.60, 0.40, 2.54)
```

```
b3 <- c(20, 40, 30, 80, 100)
```

```
B <- rbind(b1, b2, b3)
```

```
# matrix multiplication
```

```
A %*% B
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 1604.294 3209.974 2416.768 6416.752 8022.775
## [2,] 1805.501 3614.153 2722.380 7224.820 9032.207
## [3,] 1408.074 2825.785 2136.120 5648.680 7058.318
## [4,] 1008.952 2029.680 1540.760 4056.640 5067.064
## [5,] 1512.172 3041.869 2306.408 6080.712 7594.521
```

```
B %*% A
```

```
##           a1          a2          a3
## b1  190.00  47.4000  1865.0
## b2   55.39  15.7273   654.6
## b3 1900.00 476.6000 18800.0
```

```
t(A) %*% t(B)
```

```
##           b1          b2          b3
## a1  190.0  55.3900  1900.0
## a2   47.4  15.7273   476.6
## a3 1865.0 654.6000 18800.0
```

```
#t(B) %*% crossprod(A, B) #Not computable!!!
```

```
#Play around with the R-embedded data, iris.
```

```
iris
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1           5.1           3.5           1.4           0.2    setosa
## 2           4.9           3.0           1.4           0.2    setosa
## 3           4.7           3.2           1.3           0.2    setosa
## 4           4.6           3.1           1.5           0.2    setosa
## 5           5.0           3.6           1.4           0.2    setosa
## 6           5.4           3.9           1.7           0.4    setosa
## 7           4.6           3.4           1.4           0.3    setosa
## 8           5.0           3.4           1.5           0.2    setosa
## 9           4.4           2.9           1.4           0.2    setosa
## 10          4.9           3.1           1.5           0.1    setosa
## 11          5.4           3.7           1.5           0.2    setosa
## 12          4.8           3.4           1.6           0.2    setosa
## 13          4.8           3.0           1.4           0.1    setosa
## 14          4.3           3.0           1.1           0.1    setosa
## 15          5.8           4.0           1.2           0.2    setosa
## 16          5.7           4.4           1.5           0.4    setosa
## 17          5.4           3.9           1.3           0.4    setosa
## 18          5.1           3.5           1.4           0.3    setosa
## 19          5.7           3.8           1.7           0.3    setosa
```

## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa
## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 55	6.5	2.8	4.6	1.5	versicolor
## 56	5.7	2.8	4.5	1.3	versicolor
## 57	6.3	3.3	4.7	1.6	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 60	5.2	2.7	3.9	1.4	versicolor
## 61	5.0	2.0	3.5	1.0	versicolor
## 62	5.9	3.0	4.2	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 64	6.1	2.9	4.7	1.4	versicolor
## 65	5.6	2.9	3.6	1.3	versicolor
## 66	6.7	3.1	4.4	1.4	versicolor
## 67	5.6	3.0	4.5	1.5	versicolor
## 68	5.8	2.7	4.1	1.0	versicolor
## 69	6.2	2.2	4.5	1.5	versicolor
## 70	5.6	2.5	3.9	1.1	versicolor
## 71	5.9	3.2	4.8	1.8	versicolor
## 72	6.1	2.8	4.0	1.3	versicolor
## 73	6.3	2.5	4.9	1.5	versicolor

## 74	6.1	2.8	4.7	1.2 versicolor
## 75	6.4	2.9	4.3	1.3 versicolor
## 76	6.6	3.0	4.4	1.4 versicolor
## 77	6.8	2.8	4.8	1.4 versicolor
## 78	6.7	3.0	5.0	1.7 versicolor
## 79	6.0	2.9	4.5	1.5 versicolor
## 80	5.7	2.6	3.5	1.0 versicolor
## 81	5.5	2.4	3.8	1.1 versicolor
## 82	5.5	2.4	3.7	1.0 versicolor
## 83	5.8	2.7	3.9	1.2 versicolor
## 84	6.0	2.7	5.1	1.6 versicolor
## 85	5.4	3.0	4.5	1.5 versicolor
## 86	6.0	3.4	4.5	1.6 versicolor
## 87	6.7	3.1	4.7	1.5 versicolor
## 88	6.3	2.3	4.4	1.3 versicolor
## 89	5.6	3.0	4.1	1.3 versicolor
## 90	5.5	2.5	4.0	1.3 versicolor
## 91	5.5	2.6	4.4	1.2 versicolor
## 92	6.1	3.0	4.6	1.4 versicolor
## 93	5.8	2.6	4.0	1.2 versicolor
## 94	5.0	2.3	3.3	1.0 versicolor
## 95	5.6	2.7	4.2	1.3 versicolor
## 96	5.7	3.0	4.2	1.2 versicolor
## 97	5.7	2.9	4.2	1.3 versicolor
## 98	6.2	2.9	4.3	1.3 versicolor
## 99	5.1	2.5	3.0	1.1 versicolor
## 100	5.7	2.8	4.1	1.3 versicolor
## 101	6.3	3.3	6.0	2.5 virginica
## 102	5.8	2.7	5.1	1.9 virginica
## 103	7.1	3.0	5.9	2.1 virginica
## 104	6.3	2.9	5.6	1.8 virginica
## 105	6.5	3.0	5.8	2.2 virginica
## 106	7.6	3.0	6.6	2.1 virginica
## 107	4.9	2.5	4.5	1.7 virginica
## 108	7.3	2.9	6.3	1.8 virginica
## 109	6.7	2.5	5.8	1.8 virginica
## 110	7.2	3.6	6.1	2.5 virginica
## 111	6.5	3.2	5.1	2.0 virginica
## 112	6.4	2.7	5.3	1.9 virginica
## 113	6.8	3.0	5.5	2.1 virginica
## 114	5.7	2.5	5.0	2.0 virginica
## 115	5.8	2.8	5.1	2.4 virginica
## 116	6.4	3.2	5.3	2.3 virginica
## 117	6.5	3.0	5.5	1.8 virginica
## 118	7.7	3.8	6.7	2.2 virginica
## 119	7.7	2.6	6.9	2.3 virginica
## 120	6.0	2.2	5.0	1.5 virginica
## 121	6.9	3.2	5.7	2.3 virginica
## 122	5.6	2.8	4.9	2.0 virginica
## 123	7.7	2.8	6.7	2.0 virginica
## 124	6.3	2.7	4.9	1.8 virginica
## 125	6.7	3.3	5.7	2.1 virginica
## 126	7.2	3.2	6.0	1.8 virginica
## 127	6.2	2.8	4.8	1.8 virginica

```
## 128      6.1      3.0      4.9      1.8 virginica
## 129      6.4      2.8      5.6      2.1 virginica
## 130      7.2      3.0      5.8      1.6 virginica
## 131      7.4      2.8      6.1      1.9 virginica
## 132      7.9      3.8      6.4      2.0 virginica
## 133      6.4      2.8      5.6      2.2 virginica
## 134      6.3      2.8      5.1      1.5 virginica
## 135      6.1      2.6      5.6      1.4 virginica
## 136      7.7      3.0      6.1      2.3 virginica
## 137      6.3      3.4      5.6      2.4 virginica
## 138      6.4      3.1      5.5      1.8 virginica
## 139      6.0      3.0      4.8      1.8 virginica
## 140      6.9      3.1      5.4      2.1 virginica
## 141      6.7      3.1      5.6      2.4 virginica
## 142      6.9      3.1      5.1      2.3 virginica
## 143      5.8      2.7      5.1      1.9 virginica
## 144      6.8      3.2      5.9      2.3 virginica
## 145      6.7      3.3      5.7      2.5 virginica
## 146      6.7      3.0      5.2      2.3 virginica
## 147      6.3      2.5      5.0      1.9 virginica
## 148      6.5      3.0      5.2      2.0 virginica
## 149      6.2      3.4      5.4      2.3 virginica
## 150      5.9      3.0      5.1      1.8 virginica
```

```
summary(iris)
```

```
##      Sepal.Length      Sepal.Width      Petal.Length      Petal.Width
## Min.      :4.300   Min.      :2.000   Min.      :1.000   Min.      :0.100
## 1st Qu.:5.100   1st Qu.:2.800   1st Qu.:1.600   1st Qu.:0.300
## Median :5.800   Median :3.000   Median :4.350   Median :1.300
## Mean    :5.843   Mean    :3.057   Mean    :3.758   Mean    :1.199
## 3rd Qu.:6.400   3rd Qu.:3.300   3rd Qu.:5.100   3rd Qu.:1.800
## Max.    :7.900   Max.    :4.400   Max.    :6.900   Max.    :2.500
##           Species
## setosa      :50
## versicolor:50
## virginica   :50
##
##
##
```

```
dim(iris)
```

```
## [1] 150  5
```

```
head(iris)
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1      5.1      3.5      1.4      0.2 setosa
## 2      4.9      3.0      1.4      0.2 setosa
## 3      4.7      3.2      1.3      0.2 setosa
## 4      4.6      3.1      1.5      0.2 setosa
## 5      5.0      3.6      1.4      0.2 setosa
## 6      5.4      3.9      1.7      0.4 setosa
```

```
lincomb <- as.matrix(iris[1:4]) %*% c(1:4)
```

```

lincomb2 <- (1 * diag(150) %*% iris$Sepal.Length) + (2 * diag(150) %*% iris$Sepal.Width) + (3 * diag(150) %*% iris$Petal.Length)

# Check out rdist help to get more of these!!!!
# 2-norm == euclidean norm == L2 norm
# 1-norm == manhattan norm
# max norm == chebyshev

# dot product ~ inner product ==> shows similarity (As angle between x and y gets closer, its bigger and more similar)
# X t(X) -> observation view & t(X) X -> feature/variable view ==> Then, each element  $x_{ij}$  shows the similarity between observation i and feature j

#Define vnorm.
v <- 1:5
vnorm <-function(x){
  sqrt(t(x) %*% x)
}
vnorm(v)

##           [,1]
## [1,] 7.416198

u <- v / vnorm(v) #unit vector

#Check whether a matrix is square.
is_square <- function(A){
  if(dim(A)[1] == dim(A)[2]){
    T
  }else{
    F
  }
}

#Trace is only defined for square matrix.
mtrace <- function(A){
  if(is_square(A) == T){
    sum(diag(A))
  }else{
    NA
  }
}

```

```
}
}
```

```
#Definte p and q to check mtrace is a linear mapping.
```

```
p <- matrix(1:9, 3, 3)
q <- matrix(10:18, 3, 3)
```

```
#Check it is a linear, by confirming additivity and homogeneity
mtrace(p + q) == mtrace(p) + mtrace(q)
```

```
## [1] TRUE
```

```
c <- runif(1, 1, 500) #randomly pick number.
mtrace(c * p) == c * mtrace(p)
```

```
## [1] TRUE
```

```
#Verify all traces are equal.
```

```
tr1 <- mtrace(crossprod(p, q))
tr2 <- mtrace(tcrossprod(p, q))
tr3 <- mtrace(crossprod(q, p))
tr4 <- mtrace(tcrossprod(q, p))
```

```
lapply(list(tr1, tr2, tr3, tr4), identical, tr1)
```

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

```
## [1] TRUE
```

```
##
```

```
## [[2]]
```

```
## [1] TRUE
```

```
##
```

```
## [[3]]
```

```
## [1] TRUE
```

```
##
```

```
## [[4]]
```

```
## [1] TRUE
```

```
sapply(list(tr1, tr2, tr3, tr4), identical, tr1)
```

```
## [1] TRUE TRUE TRUE TRUE
```

```
all(sapply(list(tr1, tr2, tr3, tr4), identical, tr1)) #If you apply "all" into lapply, it shows warning
```

```
## [1] TRUE
```

```
#know how to get a determinant mathematically -> only defined in square matrix.
```

*Here is a proof of four traces being equal :*

$$\text{tr}(X^T Y) = \text{tr}(X Y^T) = \text{tr}(Y^T X) = \text{tr}(Y X^T)$$

*Let the element of matrix X be  $x_{ij}$  and element of matrix Y be  $y_{ij}$  where X and Y are both square with the length of row a*

$$\text{tr}(X^T Y) = (X^T Y)_{11} + \dots + (X^T Y)_{nn} = x_{11}^T y_{11} + \dots + x_{n1}^T y_{1n} + \dots + x_{n1}^T y_{1n} + \dots + x_{nn}^T y_{nn} = x_{11} y_{11}^T + \dots + x_{n1} y_{1n}^T + \dots + x_{n1} y_{1n}^T + \dots + x_{nn} y_{nn}^T = \text{tr}(XY^T).$$

```
#See what mtcars a
class(mtcars)

## [1] "data.frame"

typeof(mtcars)

## [1] "list"

mode(mtcars)

## [1] "list"

head(mtcars)

##           mpg  cyl  disp  hp  drat    wt  qsec vs  am  gear  carb
## Mazda RX4      21.0   6  160 110 3.90 2.620 16.46 0  1    4    4
## Mazda RX4 Wag  21.0   6  160 110 3.90 2.875 17.02 0  1    4    4
## Datsun 710      22.8   4  108  93 3.85 2.320 18.61 1  1    4    1
## Hornet 4 Drive  21.4   6  258 110 3.08 3.215 19.44 1  0    3    1
## Hornet Sportabout 18.7   8  360 175 3.15 3.440 17.02 0  0    3    2
## Valiant         18.1   6  225 105 2.76 3.460 20.22 1  0    3    1

#Make a new matrix.
M <- cbind(mtcars$mpg, mtcars$disp, mtcars$hp, mtcars$drat, mtcars$wt)
apply(M, MARGIN = 2, FUN = mean) #Get column means

## [1] 20.090625 230.721875 146.687500 3.596563 3.217250

#Mean centered matrix
Mc <- scale(M, center = T, scale = F) #or, do sweep(M, MARGIN = 2, colMeans(M), FUN = "-")

#check Mc
colMeans(Mc) #all closed enough to zero, so it is mean-centered matrix

## [1] 4.440892e-16 -1.199041e-14 0.000000e+00 -1.526557e-16 3.469447e-17

#Another version of mean central
sweep(M, MARGIN = 2, colMeans(M), FUN = "-")

##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 0.909375 -70.721875 -36.6875 0.3034375 -0.59725
## [2,] 0.909375 -70.721875 -36.6875 0.3034375 -0.34225
## [3,] 2.709375 -122.721875 -53.6875 0.2534375 -0.89725
## [4,] 1.309375 27.278125 -36.6875 -0.5165625 -0.00225
## [5,] -1.390625 129.278125 28.3125 -0.4465625 0.22275
## [6,] -1.990625 -5.721875 -41.6875 -0.8365625 0.24275
## [7,] -5.790625 129.278125 98.3125 -0.3865625 0.35275
## [8,] 4.309375 -84.021875 -84.6875 0.0934375 -0.02725
## [9,] 2.709375 -89.921875 -51.6875 0.3234375 -0.06725
## [10,] -0.890625 -63.121875 -23.6875 0.3234375 0.22275
## [11,] -2.290625 -63.121875 -23.6875 0.3234375 0.22275
## [12,] -3.690625 45.078125 33.3125 -0.5265625 0.85275
```



```
## [13,] -2.790625  45.078125  33.3125 -0.5265625  0.51275
## [14,] -4.890625  45.078125  33.3125 -0.5265625  0.56275
## [15,] -9.690625 241.278125  58.3125 -0.6665625  2.03275
## [16,] -9.690625 229.278125  68.3125 -0.5965625  2.20675
## [17,] -5.390625 209.278125  83.3125 -0.3665625  2.12775
## [18,] 12.309375 -152.021875 -80.6875  0.4834375 -1.01725
## [19,] 10.309375 -155.021875 -94.6875  1.3334375 -1.60225
## [20,] 13.809375 -159.621875 -81.6875  0.6234375 -1.38225
## [21,]  1.409375 -110.621875 -49.6875  0.1034375 -0.75225
## [22,] -4.590625  87.278125   3.3125 -0.8365625  0.30275
## [23,] -4.890625  73.278125   3.3125 -0.4465625  0.21775
## [24,] -6.790625 119.278125  98.3125  0.1334375  0.62275
## [25,] -0.890625 169.278125  28.3125 -0.5165625  0.62775
## [26,]  7.209375 -151.721875 -80.6875  0.4834375 -1.28225
## [27,]  5.909375 -110.421875 -55.6875  0.8334375 -1.07725
## [28,] 10.309375 -135.621875 -33.6875  0.1734375 -1.70425
## [29,] -4.290625 120.278125 117.3125  0.6234375 -0.04725
## [30,] -0.390625 -85.721875  28.3125  0.0234375 -0.44725
## [31,] -5.090625  70.278125 188.3125 -0.0565625  0.35275
## [32,]  1.309375 -109.721875 -37.6875  0.5134375 -0.43725
```

*#Find a column maxima*

```
colmax <- apply(M, 2, max)
```

*#Scale by column maxima*

```
sweep(M, 2, colmax, FUN = "/") #or, do scale(M, F, colmax)
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 0.6194690 0.3389831 0.3283582 0.7910751 0.4830383
## [2,] 0.6194690 0.3389831 0.3283582 0.7910751 0.5300516
## [3,] 0.6725664 0.2288136 0.2776119 0.7809331 0.4277286
## [4,] 0.6312684 0.5466102 0.3283582 0.6247465 0.5927360
## [5,] 0.5516224 0.7627119 0.5223881 0.6389452 0.6342183
## [6,] 0.5339233 0.4766949 0.3134328 0.5598377 0.6379056
## [7,] 0.4218289 0.7627119 0.7313433 0.6511156 0.6581858
## [8,] 0.7197640 0.3108051 0.1850746 0.7484787 0.5881268
## [9,] 0.6725664 0.2983051 0.2835821 0.7951318 0.5807522
## [10,] 0.5663717 0.3550847 0.3671642 0.7951318 0.6342183
## [11,] 0.5250737 0.3550847 0.3671642 0.7951318 0.6342183
## [12,] 0.4837758 0.5843220 0.5373134 0.6227181 0.7503687
## [13,] 0.5103245 0.5843220 0.5373134 0.6227181 0.6876844
## [14,] 0.4483776 0.5843220 0.5373134 0.6227181 0.6969027
## [15,] 0.3067847 1.0000000 0.6119403 0.5943205 0.9679204
## [16,] 0.3067847 0.9745763 0.6417910 0.6085193 1.0000000
## [17,] 0.4336283 0.9322034 0.6865672 0.6551724 0.9854351
## [18,] 0.9557522 0.1667373 0.1970149 0.8275862 0.4056047
## [19,] 0.8967552 0.1603814 0.1552239 1.0000000 0.2977507
## [20,] 1.0000000 0.1506356 0.1940299 0.8559838 0.3383112
## [21,] 0.6342183 0.2544492 0.2895522 0.7505071 0.4544617
## [22,] 0.4572271 0.6737288 0.4477612 0.5598377 0.6489676
## [23,] 0.4483776 0.6440678 0.4477612 0.6389452 0.6332965
## [24,] 0.3923304 0.7415254 0.7313433 0.7565923 0.7079646
```

```
## [25,] 0.5663717 0.8474576 0.5223881 0.6247465 0.7088864
## [26,] 0.8053097 0.1673729 0.1970149 0.8275862 0.3567478
## [27,] 0.7669617 0.2548729 0.2716418 0.8985801 0.3945428
## [28,] 0.8967552 0.2014831 0.3373134 0.7647059 0.2789454
## [29,] 0.4660767 0.7436441 0.7880597 0.8559838 0.5844395
## [30,] 0.5811209 0.3072034 0.5223881 0.7342799 0.5106932
## [31,] 0.4424779 0.6377119 1.0000000 0.7180527 0.6581858
## [32,] 0.6312684 0.2563559 0.3253731 0.8336714 0.5125369
```

```
#scale to have min = 0 and max = 1
scaling <- function(x){
  (x - min(x))/diff(range(x))
}
```

```
apply(M, 2, scaling)
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 0.4510638 0.22175106 0.20494700 0.52534562 0.28304781
## [2,] 0.4510638 0.22175106 0.20494700 0.52534562 0.34824853
## [3,] 0.5276596 0.09204290 0.14487633 0.50230415 0.20634109
## [4,] 0.4680851 0.46620105 0.20494700 0.14746544 0.43518282
## [5,] 0.3531915 0.72062859 0.43462898 0.17972350 0.49271286
## [6,] 0.3276596 0.38388626 0.18727915 0.00000000 0.49782664
## [7,] 0.1659574 0.72062859 0.68197880 0.20737327 0.52595244
## [8,] 0.5957447 0.18857570 0.03533569 0.42857143 0.42879059
## [9,] 0.5276596 0.17385882 0.15194346 0.53456221 0.41856303
## [10,] 0.3744681 0.24070841 0.25088339 0.53456221 0.49271286
## [11,] 0.3148936 0.24070841 0.25088339 0.53456221 0.49271286
## [12,] 0.2553191 0.51060115 0.45229682 0.14285714 0.65379698
## [13,] 0.2936170 0.51060115 0.45229682 0.14285714 0.56686269
## [14,] 0.2042553 0.51060115 0.45229682 0.14285714 0.57964715
## [15,] 0.0000000 1.00000000 0.54063604 0.07834101 0.95551010
## [16,] 0.0000000 0.97006735 0.57597173 0.11059908 1.00000000
## [17,] 0.1829787 0.92017960 0.62897527 0.21658986 0.97980056
## [18,] 0.9361702 0.01895735 0.04946996 0.60829493 0.17565840
## [19,] 0.8510638 0.01147418 0.00000000 1.00000000 0.02608029
## [20,] 1.0000000 0.00000000 0.04593640 0.67281106 0.08233188
## [21,] 0.4723404 0.12222499 0.15901060 0.43317972 0.24341601
## [22,] 0.2170213 0.61586431 0.34628975 0.00000000 0.51316799
## [23,] 0.2042553 0.58094288 0.34628975 0.17972350 0.49143442
## [24,] 0.1234043 0.69568471 0.68197880 0.44700461 0.59498849
## [25,] 0.3744681 0.82040409 0.43462898 0.14746544 0.59626694
## [26,] 0.7191489 0.01970566 0.04946996 0.60829493 0.10790079
## [27,] 0.6638298 0.12272387 0.13780919 0.76958525 0.16031705
## [28,] 0.8510638 0.05986530 0.21554770 0.46543779 0.00000000
## [29,] 0.2297872 0.69817910 0.74911661 0.67281106 0.42367681
## [30,] 0.3957447 0.18433525 0.43462898 0.39631336 0.32140118
## [31,] 0.1957447 0.57345972 1.00000000 0.35944700 0.52595244
## [32,] 0.4680851 0.12446994 0.20141343 0.62211982 0.32395807
```

```
#There is a way to get covariance in slide 5. Divided by n-1 as it is a sample.
#First
```

```
cov_m <- (1/(nrow(M) - 1)) * t(Mc) %*% Mc
```

```
cov_m
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,]  36.324103 -633.09721 -320.73206  2.1950635 -5.1166847
## [2,] -633.097208 15360.79983 6721.15867 -47.0640192 107.6842040
## [3,] -320.732056 6721.15867 4700.86694 -16.4511089 44.1926613
## [4,]  2.195064  -47.06402  -16.45111  0.2858814  -0.3727207
## [5,]  -5.116685  107.68420  44.19266  -0.3727207  0.9573790
```

```
#Second
cov(M)
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,]  36.324103 -633.09721 -320.73206  2.1950635 -5.1166847
## [2,] -633.097208 15360.79983 6721.15867 -47.0640192 107.6842040
## [3,] -320.732056 6721.15867 4700.86694 -16.4511089 44.1926613
## [4,]  2.195064  -47.06402  -16.45111  0.2858814  -0.3727207
## [5,]  -5.116685  107.68420  44.19266  -0.3727207  0.9573790
```

*#Although identical function gives "FALSE" but there is only small difference in a decimal point....*

*#There is a way to get correlation in slide 5.*

*#First -> as Mc are mean-centered. Try to use the way on pg 31 of slide 3, but it is not working...*

*#cov\_m / (t(Mc) %\*% Mc) ^2*

*#(t(Mc) %\*% Mc) / sqrt((t(Mc) %\*% Mc)) %\*% sqrt((Mc %\*% t(Mc)))*

*#Second*

```
scaling2 <- sweep(Mc, 2, apply(Mc, 2, FUN = sd), FUN = "/")
cor_m2 <- (1/(nrow(M) - 1)) * t(scaling2) %*% scaling2
```

```
cor_m2
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,]  1.0000000 -0.8475514 -0.7761684  0.6811719 -0.8676594
## [2,] -0.8475514  1.0000000  0.7909486 -0.7102139  0.8879799
## [3,] -0.7761684  0.7909486  1.0000000 -0.4487591  0.6587479
## [4,]  0.6811719 -0.7102139 -0.4487591  1.0000000 -0.7124406
## [5,] -0.8676594  0.8879799  0.6587479 -0.7124406  1.0000000
```

*#Third*

```
scaling <- scale(M, T, T)
cor_m <- (1/(nrow(M) - 1)) * t(scaling) %*% scaling
```

```
cor_m
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,]  1.0000000 -0.8475514 -0.7761684  0.6811719 -0.8676594
## [2,] -0.8475514  1.0000000  0.7909486 -0.7102139  0.8879799
## [3,] -0.7761684  0.7909486  1.0000000 -0.4487591  0.6587479
## [4,]  0.6811719 -0.7102139 -0.4487591  1.0000000 -0.7124406
## [5,] -0.8676594  0.8879799  0.6587479 -0.7124406  1.0000000
```

*#Fourth*

```
cor(M)
```

```
##           [,1]      [,2]      [,3]      [,4]      [,5]
## [1,]  1.0000000 -0.8475514 -0.7761684  0.6811719 -0.8676594
## [2,] -0.8475514  1.0000000  0.7909486 -0.7102139  0.8879799
## [3,] -0.7761684  0.7909486  1.0000000 -0.4487591  0.6587479
## [4,]  0.6811719 -0.7102139 -0.4487591  1.0000000 -0.7124406
## [5,] -0.8676594  0.8879799  0.6587479 -0.7124406  1.0000000
```

*#Dummify function*

```
dummify <-function(x, all = T){
  if(all == T){
    new_matrix<- matrix(1, length(x), length(levels(x))) #create an arbitrary matrix that is made of 1
    colnames(new_matrix) <- c(levels(x))
    rownames(new_matrix) <- x
    #assign 0 binary if row and column numbers are not the same
    for (i in 1:length(x)){
      for (j in 1:length(levels(x))){
        if(rownames(new_matrix)[i] != colnames(new_matrix)[j]){
          new_matrix[i,j] <- 0
        }
      }
    }
    return(new_matrix)
  }else{
    new_matrix<- matrix(1, length(x), length(levels(x))) #create an arbitrary matrix that is made of 1
    colnames(new_matrix) <- c(levels(x))
    rownames(new_matrix) <- x
    #assign 0 binary if row and column numbers are not the same
    for (i in 1:length(x)){
      for (j in 1:length(levels(x))){
        if(rownames(new_matrix)[i] != colnames(new_matrix)[j]){
          new_matrix[i,j] <- 0
        }
      }
    }
    # IF all=F, then we only extract k-1 indicators, by taking out one column randomly.
    random_number <- sample(1:3,1)
    return(new_matrix[, -(random_number)])
  }
}
```

*#other way*

```
dummify2 <-function(x, all = T){
  x <- as.factor(x)
  #this function compares the set (4, 6, 8) with each element of x.
  sapply(levels(x), function(data_element){as.integer(x == data_element)})
}
```

```
cyl <- factor(mtcars$cyl)
CYL1 <- dummify(cyl, all = T)
CYL2 <- dummify(cyl, all = F)
```

*#Crosstable function*

```

crosstable <-function(x, y){
  dumc <- dummify(x, all = T)
  dummy <- dummify(y, all = T)
  return(t(dumc) %*% dummy)
}

```

```

gear <- factor(mtcars$gear)
xtb <- crosstable(cyl, gear)

```

CYL1

```

##      4 6 8
## 6 0 1 0
## 6 0 1 0
## 4 1 0 0
## 6 0 1 0
## 8 0 0 1
## 6 0 1 0
## 8 0 0 1
## 4 1 0 0
## 4 1 0 0
## 6 0 1 0
## 6 0 1 0
## 8 0 0 1
## 8 0 0 1
## 8 0 0 1
## 8 0 0 1
## 8 0 0 1
## 8 0 0 1
## 4 1 0 0
## 4 1 0 0
## 4 1 0 0
## 4 1 0 0
## 8 0 0 1
## 8 0 0 1
## 8 0 0 1
## 8 0 0 1
## 4 1 0 0
## 4 1 0 0
## 4 1 0 0
## 8 0 0 1
## 6 0 1 0
## 8 0 0 1
## 4 1 0 0

```

```
dummify2(cyl, all = T)
```

```

##      4 6 8
## [1,] 0 1 0
## [2,] 0 1 0
## [3,] 1 0 0
## [4,] 0 1 0
## [5,] 0 0 1
## [6,] 0 1 0

```

```

## [7,] 0 0 1
## [8,] 1 0 0
## [9,] 1 0 0
## [10,] 0 1 0
## [11,] 0 1 0
## [12,] 0 0 1
## [13,] 0 0 1
## [14,] 0 0 1
## [15,] 0 0 1
## [16,] 0 0 1
## [17,] 0 0 1
## [18,] 1 0 0
## [19,] 1 0 0
## [20,] 1 0 0
## [21,] 1 0 0
## [22,] 0 0 1
## [23,] 0 0 1
## [24,] 0 0 1
## [25,] 0 0 1
## [26,] 1 0 0
## [27,] 1 0 0
## [28,] 1 0 0
## [29,] 0 0 1
## [30,] 0 1 0
## [31,] 0 0 1
## [32,] 1 0 0

```

```
xtb
```

```

##      3 4 5
## 4    1 8 2
## 6    2 4 1
## 8   12 0 2

```