Data Mining and Machine Learning in Bioinformatics

Exercise Series 3

Group members (Name, Student ID, E-Mail):

- Baldomero Valdez, Valenzuela, 2905175, baldmer.w@gmail.com
- Omar Trinidad Gutierrez Mendez, 2850441, omar.vpa@gmail.com
- Shinho Kang, 2890169, wis.shinho.kang@gmail.com

Task 4

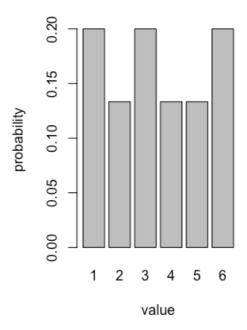
CODE

```
# Parameters
   m: sample matrix
   p: probability vector
  n: number of trials
multinom <- function(m, p, n) {</pre>
     - the expectation value
 ex <- p*n
 print ('======')
 print ('Expectation Value')
  print (ex)
     - the population variance/covariance matrix
  var_m<-n*p*(1-p)</pre>
  print ('=======')
  print ('Population variance')
  print (var_m)
  cov m<-matrix(0,ncol=6,nrow=6)</pre>
  for (i in 1:length(p)) {
   for (j in 1:length(p)) {
     #print (-1 * n * i * j)
     cov_m[i,j] \leftarrow (-1 * n * p[i] * p[j])
   }
  print ('======')
```

```
print ('Population covariance matrix')
  print (cov_m)
      - the sample mean (R functions apply and mean)
  print ('=======')
  print ('Sample Mean')
  print (apply(m, 2, mean))
  \#apply(m,1,function(x) sum(x*c(1,2,3,4,5,6))/n)
     - the sample covariance matrix (R function cov)
 print ('======')
  print ('Sample Covariance Matrix')
  print (cov(m))
     - the value of the probability mass function for each sample (Tip: you can u
se R function dmultinom)
 print ('======')
 print ('Value of the probability mass function')
 print (apply(m, 1, function(x) dmultinom(x, prob=p)))
 # visualizing p
 barplot(p, xlab = "value", ylab="probability", axisnames = T, names.arg = c(1,2,
3,4,5,6))
}
# probability vector
p \leftarrow c(0.2, 0.4/3, 0.2, 0.4/3, 0.4/3, 0.2)
# number of trials
n < -10
# sample matrix - 50 random drawings from the multinomial distribution, transpose
the matrix (row=sample vector)
m<-t(rmultinom(50, n, p))</pre>
multinom(m,p,n)
```

RESULT

```
[1] "======="
[1] "Expectation Value"
[1] 2.000000 1.333333 2.000000 1.333333 1.333333 2.000000
[1] "========="
[1] "Population variance"
[1] 1.600000 1.155556 1.600000 1.155556 1.155556 1.600000
[1] "========="
[1] "Population covariance matrix"
                       [,2]
                                 [,3]
                                         [,4]
                                                        [,5]
                                                                 [,6]
[1,] -0.4000000 -0.2666667 -0.4000000 -0.2666667 -0.2666667 -0.4000000
[2,] -0.2666667 -0.1777778 -0.2666667 -0.1777778 -0.1777778 -0.2666667
[3,] -0.4000000 -0.2666667 -0.4000000 -0.2666667 -0.2666667 -0.4000000
[4,] -0.2666667 -0.1777778 -0.2666667 -0.1777778 -0.1777778 -0.2666667
[5,] -0.2666667 -0.1777778 -0.2666667 -0.1777778 -0.1777778 -0.2666667
\begin{bmatrix} 6, \end{bmatrix} -0.4000000 -0.2666667 -0.4000000 -0.2666667 -0.2666667 -0.4000000
[1] "========="
[1] "Sample Mean"
[1] 2.40 1.28 2.04 1.10 1.26 1.92
[1] "========="
[1] "Sample Covariance Matrix"
                                    [,3]
                                                                [,5]
                        [,2]
                                                [,4]
                                                                              [,6]
\begin{bmatrix} 1, \end{bmatrix} 2.2448980 -0.62448980 -0.26122449 -0.14285714 -0.3510204082 -0.8653061224
\begin{bmatrix} 2, \\ \end{bmatrix} -0.6244898 1.06285714 -0.01142857 -0.08979592 -0.3191836735 -0.0179591837
[3,] -0.2612245 -0.01142857 1.34530612 -0.10612245 -0.4800000000 -0.4865306122
\begin{bmatrix} 4 \end{bmatrix} \begin{bmatrix} -0.1428571 \\ -0.08979592 \\ -0.10612245 \\ 0.94897959 \\ -0.3122448980 \\ -0.2979591837 \\ \end{bmatrix}
\begin{bmatrix} 5 \end{bmatrix} = 0.3510204 = 0.31918367 = 0.48000000 = 0.31224490 = 1.4616326531 = 0.0008163265
\begin{bmatrix} 6, \end{bmatrix} - 0.8653061 - 0.01795918 - 0.48653061 - 0.29795918 0.0008163265 1.6669387755
[1] "========="
[1] "Value of the probability mass function"
[1] 8.601600e-04 7.645867e-05 7.645867e-04 1.146880e-03 4.587520e-03 2.293760e-03
3.058347e-03 2.293760e-03 9.061768e-05
[10] 2.293760e-03 3.398163e-04 1.019449e-03 3.058347e-03 7.645867e-04 1.146880e-03
8.601600e-04 5.097244e-04 3.058347e-03
[19] 1.529173e-03 1.529173e-03 1.529173e-03 3.440640e-04 8.601600e-04 1.146880e-03
 6.881280e-04 1.529173e-03 1.019449e-03
[28] 2.548622e-04 1.529173e-03 1.146880e-03 5.734400e-05 1.699081e-04 1.146880e-03
1.146880e-03 1.019449e-03 1.146880e-03
[37] 1.019449e-03 4.587520e-04 4.530884e-04 3.398163e-04 8.601600e-04 2.548622e-04
1.529173e-03 1.529173e-03 1.146880e-03
[46] 1.699081e-04 2.293760e-04 2.293760e-03 4.530884e-04 1.359265e-04
```



Discussion

Everytime we draw random samples, it differed mean, variance and covariances. With bigger number of samples, we get more closed results to the theoretically expected results.