18.s096 Problem Set 5 - Dimitris Koutentakis

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1 Problem 1

1.1 (a)

$$M_y(t) = e^{ta} \sum_i \sum_j P(U_i) P(X_j) e^{tU_i X_j}$$
(1)

$$= e^{ta} \sum_{i} \sum_{j} \frac{1}{2} \left(e^{btX_{i}} + e^{-btX_{i}} \right) P(X_{i})$$
 (2)

$$=e^{ta}\frac{1}{2}\left(\frac{1}{1-bt}+\frac{1}{1+bt}\right)$$
 (3)

$$=e^{ta}\frac{1}{2}\left(\frac{2}{1-(bt)^2}\right) \tag{4}$$

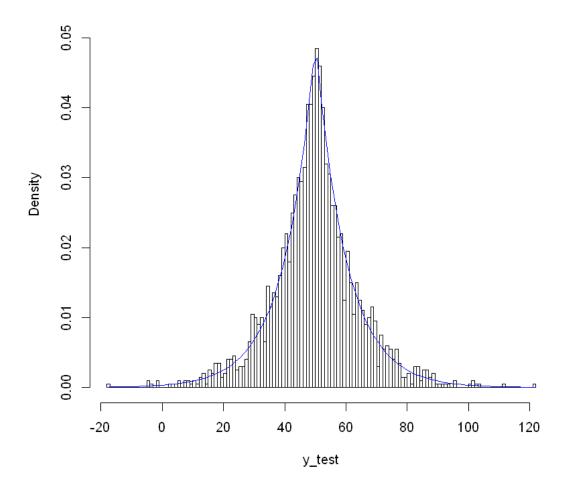
$$=\frac{e^{ta}}{1-h^2t^2}\tag{5}$$

We also know that the MGF of a Laplace distribution with location a and scale b is: $\frac{e^{ta}}{1-b^2t^2}$. Hence the two distributions are the same.

1.2 (b)

In order to find a distribution, we need to use the inverse of the CDF of the Laplace Distribution, or use the definition of Y explained in Problem 1(a).

Histogram of y_test



49.9460331816226 10.1533533619423

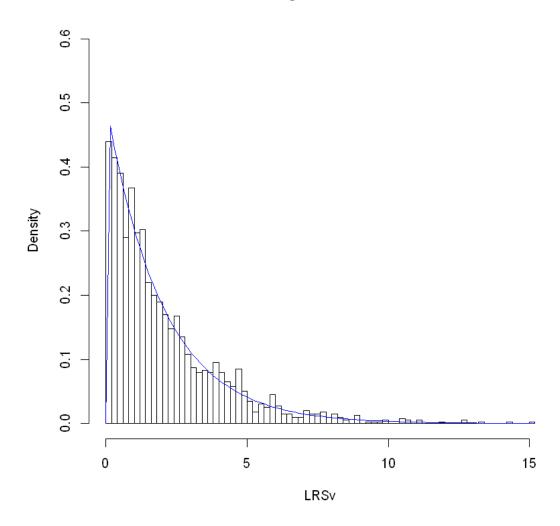
mean(abs(y_test - median(y_test)))

The sample median is the MLE estimate of the parameter a, which very close to the actual value of a=50. The sample mean absolute deviation from the median is the MLE of \hat{b} , which again is very close to the actual value of b=10.

2 Problem 2

```
2.1 (a)
```

```
In [65]: # install.packages("zoo")
         library(zoo)
         setwd("../dkout/Downloads/")
         SP500<- read.zoo(file="SP500.csv")
         y <- diff(log(SP500))
         n0 = length(y)
         location.mle = median(y)
         scale.mle = mean(abs(y-location.mle))
         cat("\n ", n0, location.mle, scale.mle, "\n")
        Error in setwd("../dkout/Downloads/"): cannot change working directory
    Traceback:
        1. setwd("../dkout/Downloads/")
In [66]: LRSv = c()
         for (i in 1:2000){
             n0 = 439
             a0 = 0.00035
             b0 = 0.006029
             y_star = myrlaplace(n0, a0, b0)
             a_star = median(y_star)
             b_star = mean(abs(y_star - median(y_star)))
             theta_star = c(a_star, b_star)
               theta\_star
             l_star = sum(log(dlaplace(y_star, a_star, b_star)))
             1_0 = sum(log(dlaplace(y_star, a0, b0)))
             LRStat = 2*(1_star - 1_0)
               LRStat
             LRSv[i] = LRStat
         }
In [67]: hist(LRSv, freq = FALSE, nclass = 100, ylim = c(0, 0.6))
         curve(dchisq(x, df=2), add = TRUE, col = "blue")
```

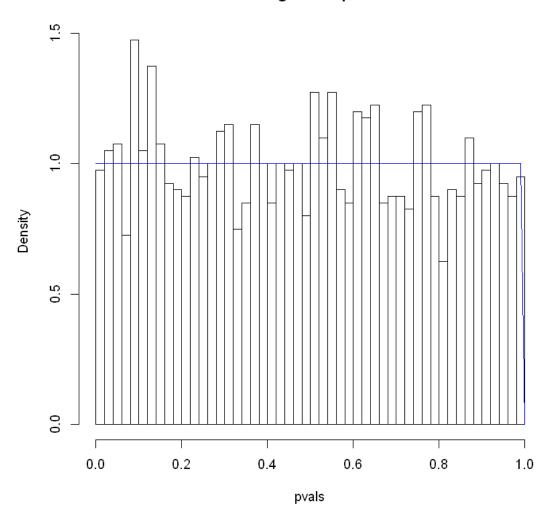


Since the blue curve follows the histogram relatively closely, we can say that the density of a Chi-squared distribution with 2 degrees of freedom is a good fit.

2.2 (c)

```
In [68]: pvals = 1-pchisq(LRSv, df = 2)
    hist(pvals, freq = FALSE, nclass = 50)
    curve(dunif(x, 0, 1), col = "blue", add = TRUE)
```

Histogram of pvals



3 Problem 3

```
m0 = 0.0000436
    sd0 = 0.0084
    y_star = rnorm(n0, m0, sd0)

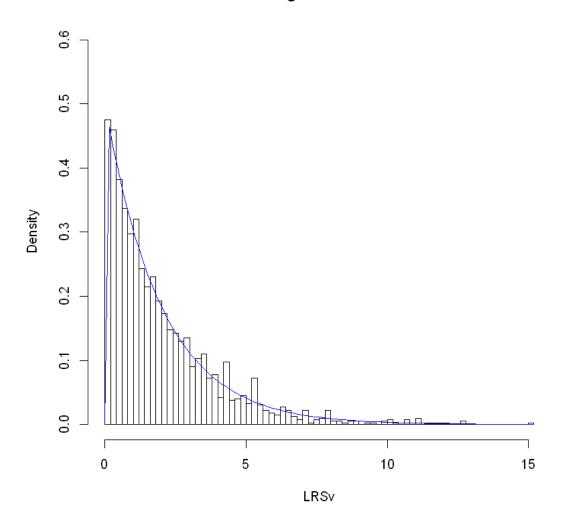
m_star = mean(y_star)
    sd_star = sqrt(mean((y_star-m_star)^2))
    theta_star = c(a_star, b_star)

# theta_star

l_star = sum(log(dnorm(y_star, m_star, sd_star)))
    l_0 = sum(log(dnorm(y_star, m0, sd0)))
    LRStat = 2*(l_star - l_0)

# LRStat
    LRSv[i] = LRStat
    }

In [71]: hist(LRSv, freq = FALSE, nclass = 100, ylim = c(0, 0.6))
    curve(dchisq(x, df=2), add = TRUE, col = "blue")
```

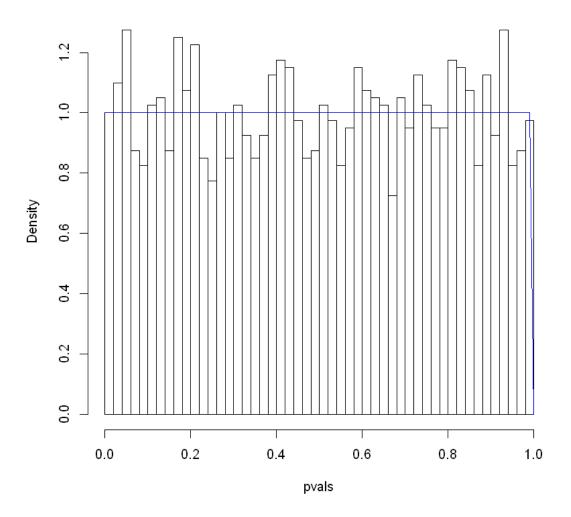


Since the blue curve very closely follows the histogram, we can say that the density of a Chi-squared distribution with 2 degrees of freedom is a good fit.

3.1 (c)

```
In [72]: pvals = pchisq(LRSv, df = 2)
    hist(pvals, freq = FALSE, nclass = 50)
    curve(dunif(x, 0, 1), col = "blue", add = TRUE)
```

Histogram of pvals



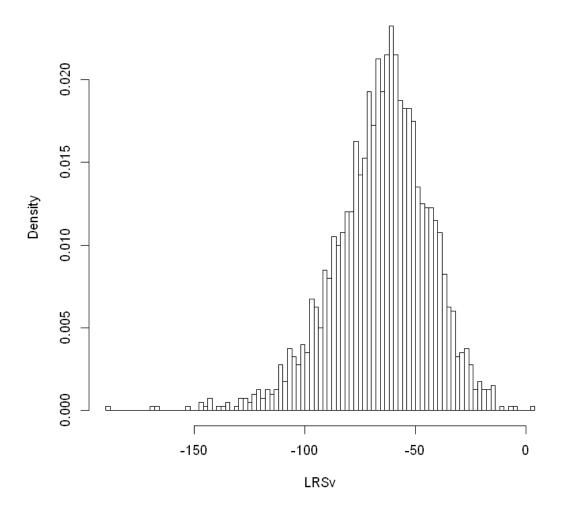
4 Problem 4

4.1 (a)

```
b0 = 0.006029
           y_star = myrlaplace(n0, a0, b0)
           a_star = median(y_star)
           b_star = mean(abs(y_star - median(y_star)))
          theta_star = c(a_star, b_star)
            theta\_star
           1_theta = sum(log(dlaplace(y_star, a_star, b_star)))
           m_star = mean(y_star)
           sd_star = sqrt(mean((y-m_star)^2))
           l_phi = sum(log(dnorm(y_star, m_star, sd_star)))
           LRStat = 2*(1_phi - 1_theta)
            LRStat
          LRSv[i] = LRStat
       }
439 0.0003495251 0.006028806
```

In [74]: hist(LRSv, nclass = 100, freq = FALSE)

4.2 (b)



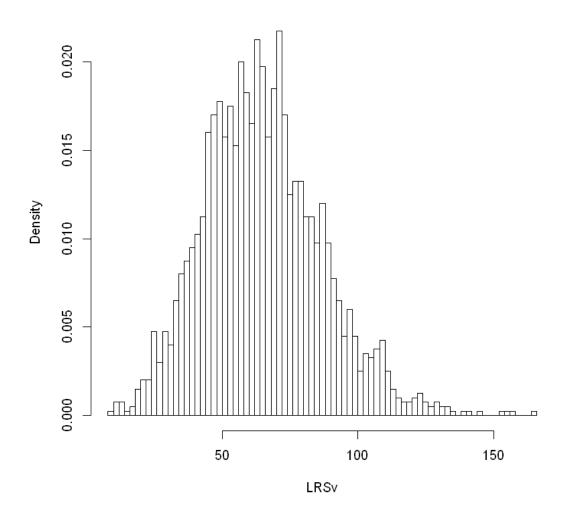
4.3 (c)

-50.7360197480102

4.4 (d)

```
cat("\n ", n0, location.mle, scale.mle, "\n")
LRSv = c()
for (i in 1:2000){
   n0 = 439
    a0 = 0.00035
    b0 = 0.006029
    y_star = myrlaplace(n0, a0, b0)
    a_star = median(y_star)
    b_star = mean(abs(y_star - median(y_star)))
    theta_star = c(a_star, b_star)
      theta\_star
    1_theta = sum(log(dlaplace(y_star, a_star, b_star)))
    m_star = mean(y_star)
    sd_star = sqrt(mean((y-m_star)^2))
    l_phi = sum(log(dnorm(y_star, m_star, sd_star)))
   LRStat = 2*(1_theta - 1_phi)
     LRStat
   LRSv[i] = LRStat
}
hist(LRSv, nclass = 100, freq = FALSE)
```

439 0.0003495251 0.006028806



5 Problem 5

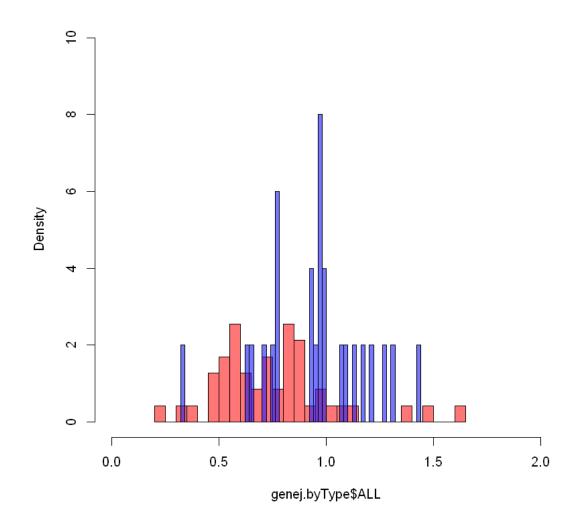
5.1 (a)

1.7128 2.72

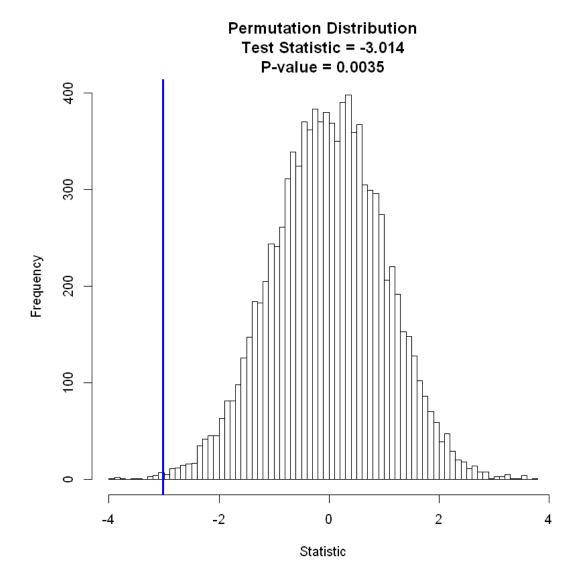
1. 'ALL' 2. 'ALL.1' 3. 'ALL.2' 4. 'ALL.3' 5. 'ALL.4' 6. 'ALL.5' 7. 'ALL.6' 8. 'ALL.7' 9. 'ALL.8' 10. 'ALL.9' 11. 'ALL.10' 12. 'ALL.11' 13. 'ALL.12' 14. 'ALL.13' 15. 'ALL.14' 16. 'ALL.15' 17. 'ALL.16' 18. 'ALL.17' 19. 'ALL.18' 20. 'ALL.19' 21. 'AML' 22. 'AML.1' 23. 'AML.2' 24. 'AML.3' 25. 'AML.4' 26. 'AML.5' 27. 'AML.6' 28. 'AML.7' 29. 'AML.8' 30. 'AML.9' 31. 'AML.10' 32. 'AML.11'

- 33. 'AML.12' 34. 'AML.13' 35. 'ALL.20' 36. 'ALL.21' 37. 'ALL.22' 38. 'ALL.23' 39. 'ALL.24'
- 40. 'ALL.25' 41. 'ALL.26' 42. 'ALL.27' 43. 'ALL.28' 44. 'ALL.29' 45. 'ALL.30' 46. 'ALL.31'
- 47. 'ALL.32' 48. 'ALL.33' 49. 'ALL.34' 50. 'ALL.35' 51. 'ALL.36' 52. 'ALL.37' 53. 'ALL.38'
- 54. 'ALL.39' 55. 'ALL.40' 56. 'ALL.41' 57. 'ALL.42' 58. 'ALL.43' 59. 'ALL.44' 60. 'ALL.45'
- 61. 'ALL.46' 62. 'AML.14' 63. 'AML.15' 64. 'AML.16' 65. 'AML.17' 66. 'AML.18' 67. 'AML.19'
- 68. 'AML.20' 69. 'AML.21' 70. 'AML.22' 71. 'AML.23' 72. 'AML.24'

Histogram of genej.byType\$ALL



```
5.2 (b)
In [80]: genej.ttest <- t.test(genej.byType[[1]], genej.byType[[2]], var.equal = TRUE)</pre>
         genej.ttest
         names(genej.ttest)
Two Sample t-test
data: genej.byType[[1]] and genej.byType[[2]]
t = -3.014, df = 70, p-value = 0.003589
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
-0.32817995 -0.06680742
sample estimates:
mean of x mean of y
0.7524794 0.9499731
   1. 'statistic' 2. 'parameter' 3. 'p.value' 4. 'conf.int' 5. 'estimate' 6. 'null.value' 7. 'alternative'
8. 'method' 9. 'data.name'
In [81]: library(perm)
         mytwosample.exact.mc<-function (gene1, gene2, alternative = "two.sided", nmc = 10^4,
             t0.test<-t.test(gene1, gene2,var.equal=TRUE)</pre>
             t0 <-t0.test$statistic
             N < - nmc
             if (setSEED)
                  set.seed(seed)
             ti <- rep(NA, N)
             for (i in 1:N) {
                  perms = sample(c(gene1, gene2))
                  ti.test<-t.test(perms[1:length(gene1)], perms[length(gene1)+1:length(gene2)],</pre>
                  ti[i] <- ti.test$statistic</pre>
                    print(ti.test)
         #
             out <- calcPvalsMC(ti, t0, digits, alternative, FALSE, p.conf.level)</pre>
             result<-list(t0=t0, ti=ti, out=out)</pre>
         }
In [82]: genej.permTest <- mytwosample.exact.mc(genej.byType[[1]], genej.byType[[2]])</pre>
In [83]: genej.pvalue<-mean(abs(genej.permTest$ti) >=abs(genej.permTest$t0))
         hist(genej.permTest$ti,nclass=100,xlab="Statistic",main=paste(c("Permutation Distribu
         abline(v=genej.permTest$t0,col='blue',lwd=3)
```



5.3 (c)

Asymptotic General Independence Test

data: genej by leukemiaType (ALL, AML) Z = -2.8558, p-value = 0.004293 alternative hypothesis: two.sided

5.4 (d)

```
In [85]: permTS(genej~leukemiaType)

Permutation Test using Asymptotic Approximation

data: genej by leukemiaType

Z = -2.8558, p-value = 0.004293

alternative hypothesis: true mean leukemiaType=ALL - mean leukemiaType=AML is not equal to 0 sample estimates:

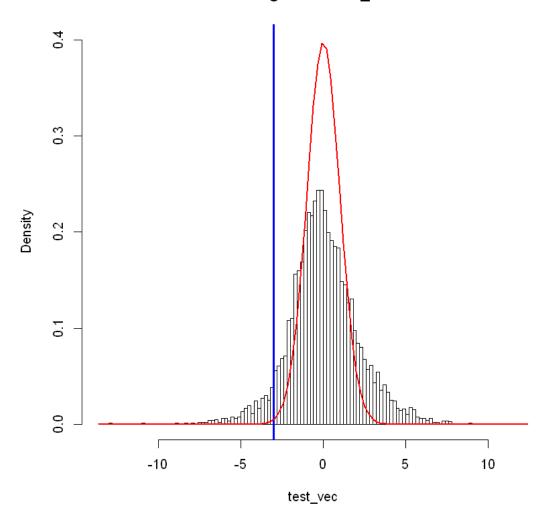
mean leukemiaType=ALL - mean leukemiaType=AML - 0.1974937
```

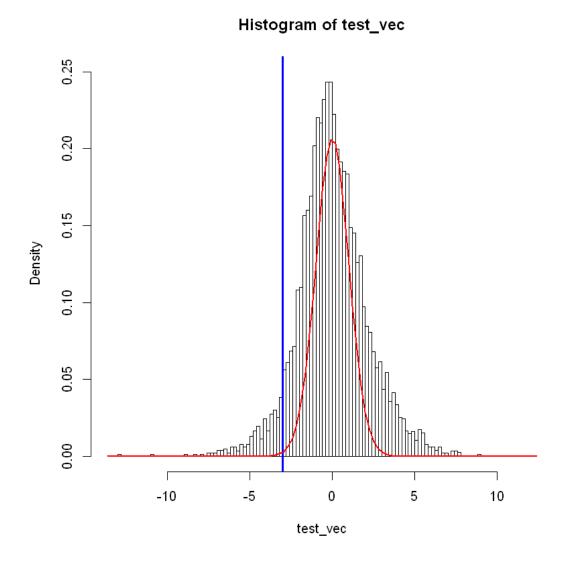
5.5 (e)

It seems that the pvalues are close enough, thus the results are relatively consistent.

5.6 (f)

Histogram of test_vec





From the graph above, it seems the value of π_0 can be at most around 0.5.