MANA130083.01 Nonparametric

Spring 2023

Solution to Homework 2

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Warning: This note is only used as a reference solution for the homework, and the solution to each question is not unique. The solution may contain factual and/or typographic errors and comments and criticism are kindly welcomed.

Problem 1 Recall that the Kolomogorov-Smirnov One-Sample Statistic $D_n = \sup_x |F_X(x) - \hat{F}_n(x)|$ is distribution free $(\hat{F}_n$ is the empirical CDF). Suppose n = 100 and the observed $D_n = 0.04$. Would you reject the null hypothesis $H_0: X_1, \ldots, X_n \sim F_X$ at level $\alpha = 0.05$. Write a simulation to justify your answer.

Solution:

```
# D: Kolomogorov-Smirnov One-Sample Statistic
   # n: the number of samples
   # nrep: the number of samples for simulation
   kspvalue<-function(D,n,nrep) {</pre>
     Dn < - c()
 5
     for (i in 1:nrep) {
        x.obs <- rnorm(n, 0, 1)
 7
       Fn.hat <- ecdf(x.obs)</pre>
       Dn[i] <- max(abs(Fn.hat(x.obs) - pnorm(x.obs)))</pre>
10
11
     return(sum(Dn > D)/nrep)
12
   # Calculate the p-value of the statistics under the problem setting
```

```
14 D = 0.04; n = 100; nrep = 5000
15 kspvalue(D,n,nrep)
```

The result shows that the p-value is 0.983 > 0.05 and we fail to reject the null hypothesis.

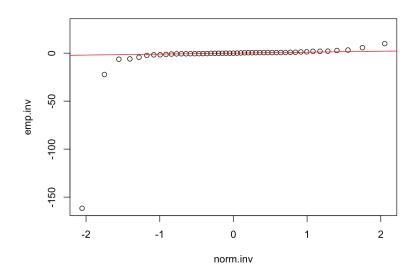
Problem 2 *Use the following r code to generate 50 random numbers from Cauchy distribution.*

```
1 set.seed (2)
2 d <- reauchy (50)</pre>
```

Use a QQ plot to see how good (or bad) your data fits a normal distribution.

Solution:

```
1 emp.inv <- sort(d)
2 sel <- seq(1, 50, by=1)/50
3 norm.inv <- qnorm(sel)
4 plot(norm.inv, emp.inv)
5 abline(0,1,col="red")</pre>
```



where we only compare the data to the standard normal distribution. If you want to show the goodness of fit for the normal family, kindly use 'abline(a=mean(d), b=sd(d), col="red")'.

Problem 3 Use a test discussed in class to test whether the following sequence is random at level $\alpha = 0.1$,

Use simulation to justify your answer.

Solution: (Solution is not unique and the following code is based on the number of run)

```
1 # n: the number of samples
2 # nrep: the number of samples for simulation
3 # sample: the original sample
4 # run: the number of run in the samples
5 randomtest.run<-function(n,nrep,sam,run){</pre>
     runs <- c()
6
7
     for (i in 1:nrep) {
       count <-1
8
       temp <- sample(sam, 11, replace = F)</pre>
9
10
       for (j in 1:(n-1)){
11
         count <- count + abs(temp[j]-temp[j+1])</pre>
12
       }
       runs[i] <- count</pre>
13
14
     return(2*min(sum(runs < run), sum(runs > run))/nrep) # two-sided p-value
15
16 }
17 # Calculate the p-value of the statistics under the problem setting
18 n=11; nrep = 5000; run=5
19 sam \leftarrow c(1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1)
20 randomtest.run(n,nrep,samp,run)
```

The corresponding p-value is 0.489 > 0.1 and we fail to reject the null.

Remark Please note that randomness implies independency but *does not suggest* sampling with equal probability (i.e. Bernoulli(p) with p=0.5). Thus, in the simulation of Problem 3, you shall sample locally rather than creat new sequence with equal probability.