

# HW7 Nonparametric Statistics

Fan Heng fh2294  
Columbia University

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## Question 1

(a)

The independent hypothesis is:

$H_0: P(X=a, Y=b) = P(X=a)P(Y=b)$   $a, b \in -1, 0, 1$  vs  $H_1: P(X=a, Y=b) \neq P(X=a)P(Y=b)$  for some  $a$  and  $b$ .

(b)

The  $\chi^2$  statistic for this test is

$$nQ_n = n \sum_{a,b \in (-1,0,1)} \frac{(P(X=a, Y=b) - P(X=a)P(Y=b))^2}{P(X=a)P(Y=b)}$$

(c)

The  $\chi^2$  test in R is

```
XY <- array(c(1, 2, 4, 4, 3, 10, 10, 16, 50), c(3, 3), )
rownames(XY) <- c("X=-1", "X=0", "X=1")
colnames(XY) <- c("Y=-1", "Y=0", "Y=1")
XY
##           Y=-1 Y=0 Y=1
## X=-1         1  4 10
## X=0          2  3 16
## X=1          4 10 50
chisq.test(XY)
## Warning: Chi-squared approximation may be incorrect
```

```
##
## Pearson's Chi-squared test
##
## data: XY
## X-squared = 1.442, df = 4, p-value = 0.8369
```

P value is 0.8369 which is very large, so we accept  $H_0$ .

## Question 2

For  $y_1 \leq y_2 \leq \dots \leq y_n$   
 $P(Y_1 \leq y_1, Y_1 \leq y_1, \dots, Y_n \leq y_n)$   
 $= P(F_x(X_{(1)}) \leq y_1, F_x(X_{(2)}) \leq y_2, \dots, F_x(X_{(n)}) \leq y_n)$   
 $= P(X_{(1)} \leq F_x^{-1}(y_1), X_{(2)} \leq F_x^{-1}(y_2), \dots, X_{(n)} \leq F_x^{-1}(y_n))$

Each  $X_i$  has the same probability to be chosen as  $X_{(1)}, \dots, \text{or } X_{(n)}$ . So there are  $n!$  permutation from  $(X_1, X_2, \dots, X_n)$  to  $(X_{(1)}, X_{(2)}, \dots, X_{(n)})$  with each having the same probability of  $\frac{1}{n!}$ . Therefore,

$P(X_{(1)} \leq F_x^{-1}(y_1), X_{(2)} \leq F_x^{-1}(y_2), \dots, X_{(n)} \leq F_x^{-1}(y_n)) = n! P(X_1 \leq F_x^{-1}(y_1), X_2 \leq F_x^{-1}(y_2), \dots, X_n \leq F_x^{-1}(y_n))$   
 $= n! P(X_1 \leq F_x^{-1}(y_1)) P(X_2 \leq F_x^{-1}(y_2)), \dots, P(X_n \leq F_x^{-1}(y_n))$   
 $= n! F_x(F_x^{-1}(y_1)) F_x(F_x^{-1}(y_2)), \dots, F_x(F_x^{-1}(y_n))$   
 $= n! y_1 y_2, \dots, y_n$

In other conditions that  $y_1 \leq y_2 \leq \dots \leq y_n$  doesn't hold,  $P(Y_1 \leq y_1, Y_1 \leq y_1, \dots, Y_n \leq y_n) = 0$ .

The joint probability density function of  $Y_1, Y_2, \dots, Y_n$  is

$f_{Y_1, Y_2, \dots, Y_n} = \frac{\partial^n P(Y_1 \leq y_1, Y_1 \leq y_1, \dots, Y_n \leq y_n)}{\partial y_1 \partial y_2 \dots \partial y_n}$   
 $= n!$

$f_{Y_1, Y_2, \dots, Y_n} =$

## Question 3

Characterizing the distribution of K through Monte Carlo simulation.

Since the distribution of K is free of  $F_0$ , we can generate sample data from any distribution. In this question, we generate data from uniform(0,1).

```
K <- rep(0, 10000)
Ind <- seq(1, 10000, 1)
for (i in 1:10000) {
  x <- runif(5, 0, 1) # generate 5 data from Uniform(0,1)
  x.ord <- x[order(x)]

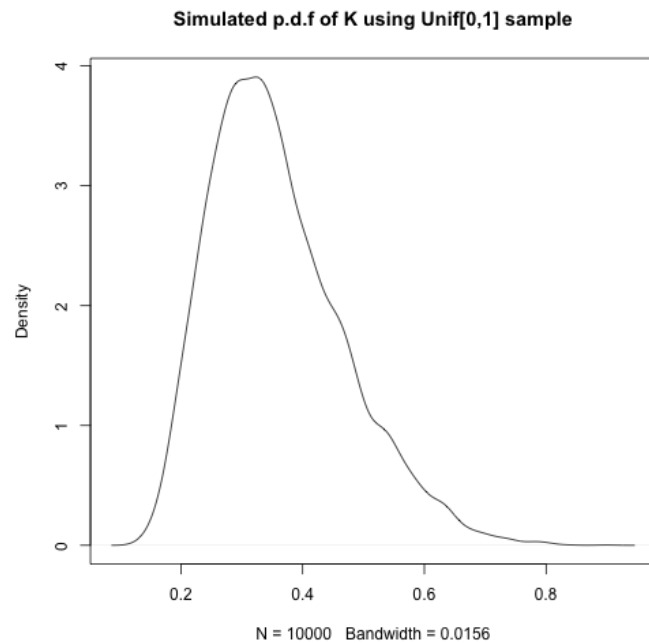
  K[i] <- max(abs((order(x.ord) - 1)/5 - x.ord), abs(order(x.ord)/5 - x.ord),
```

```

    abs(x.ord[1]), abs(1 - x.ord[5]))
}

# the density function of K
plot(density(K), main = "Simulated p.d.f of K using Unif[0,1] sample")

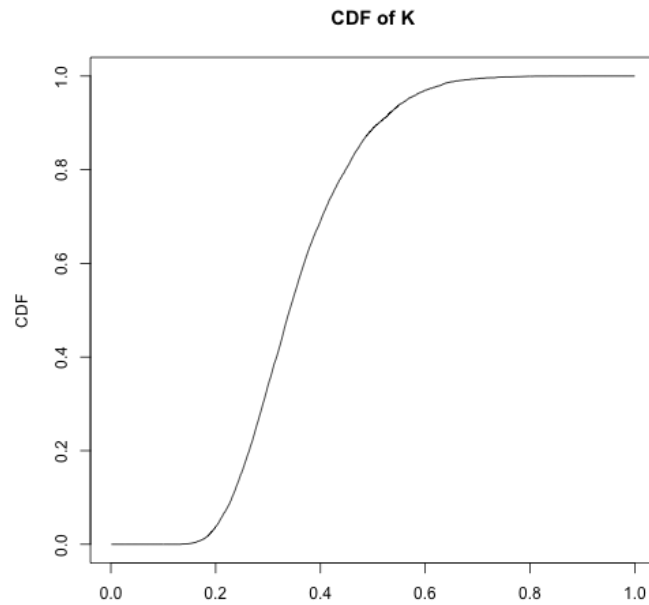
```



```

# the cumulative distribution function of K
quantileK <- seq(0.001, 1, 0.001)
cdfK <- rep(0, 1000)
for (i in 1:1000) {
  cdfK[i] <- sum(K <= quantileK[i])/10000
}
plot(quantileK, cdfK, xlab = "", ylab = "CDF", main = "CDF of K", type = "l")

```



```
mean(K)
## [1] 0.3583
var(K)
## [1] 0.01206
```

The empirical distribution of K is  $N(0.427, 0.018)$ .

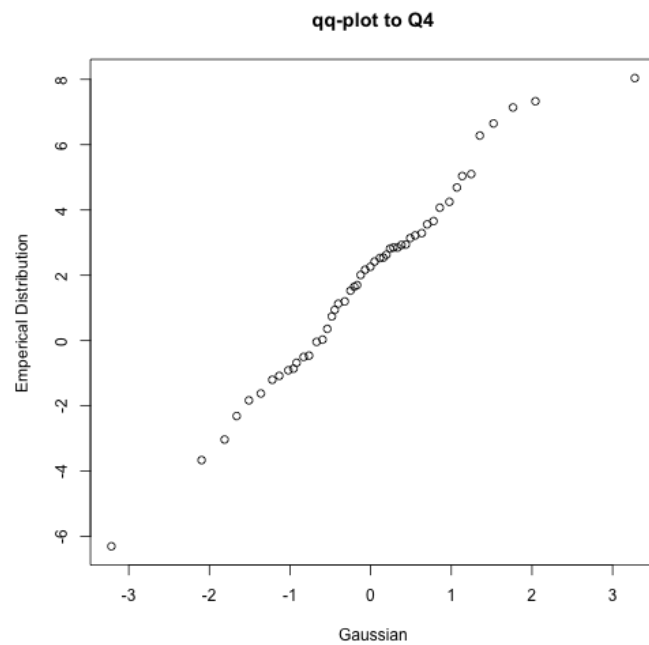
## Question 4

```
x <- c(-6.3, 2.94, 2.53, -0.86, 5.04, 3.22, -1.62, 3.56, 1.13, 2.63, -1.08,
       3.66, 4.07, -3.66, 0.74, 2.85, 2.85, 1.7, 1.53, 7.33, 2.82, -2.31, 0.94,
       -0.04, -1.2, 1.2, 5.1, 4.69, -0.46, 2.17, 2.01, 0.36, 3.14, 8.04, 7.14,
       2.54, -3.03, 4.25, -0.91, 1.65, 2.26, -1.83, -0.68, 6.28, 2.93, -0.5, 2.42,
       3.29, 0.03, 6.65)
```

(a)

Our assumption is that data  $X_1, \dots, X_{50}$  is drawn from a Gaussian distribution.

```
# Generate a new data set y from Gaussian N(0,1)
yy <- rnorm(1000, 0, 1)
qq <- qqplot(yy, x, xlab = "Gaussian", ylab = "Emperical Distribution", main = "qq-plot to Q4")
```

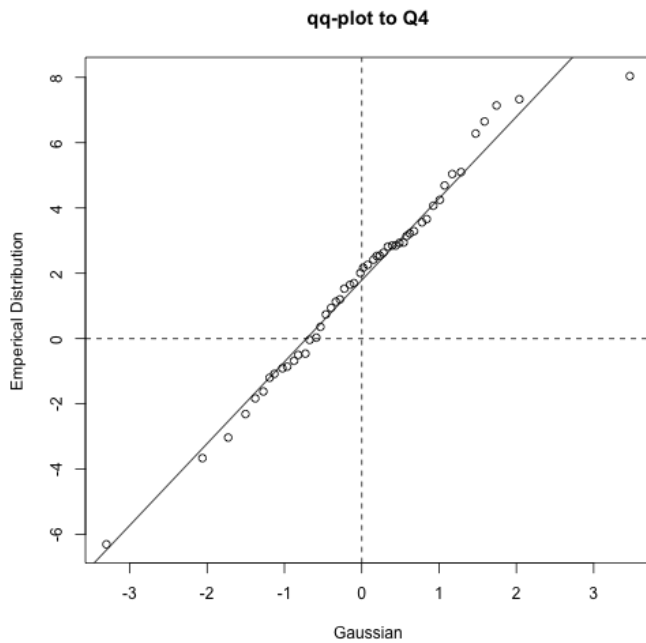


The qqplot seems like a straight line except a few points. It supports our assumption.

(b)

Fit a line to the qq-plot data and estimate mean and variance of distribution F.

```
yy <- rnorm(1000, 0, 1)
qq <- qqplot(yy, x, xlab = "Gaussian", ylab = "Emperical Distribution", main = "qq-plot to Q4")
qqfit <- lm(qq$y ~ qq$x)
abline(qqfit, lty = 1)
abline(v = 0, h = 0, lty = 2)
```



```
qqfit$coef
## (Intercept)      qq$x
##      1.790      2.502
```

QQ-plot shows the linear relationship of  $t_0(a) \equiv F_0^{-1}(\alpha)$  and  $t(a) \equiv F^{-1}(\alpha)$ . The linear equation is  $t(a) = \sigma t_0(a) + \mu$  where  $\mu$  is the mean of distribution  $F$  and  $\sigma$  is the variance.

From the fitted line to the  $qq$ -plot data,  $\hat{\sigma} = 2.637498$ , and  $\hat{\mu} = 1.823786$ .

**Question 5 , 6** on a separate page

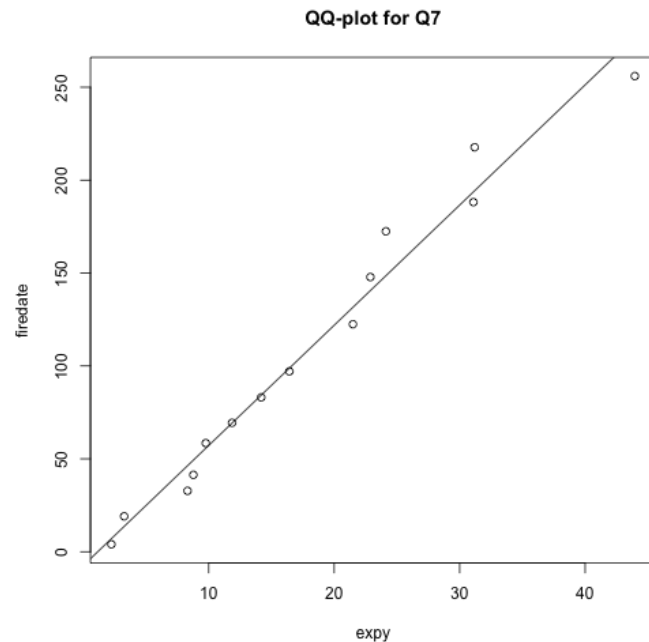
## Question 7

The Null hypothesis for this question is that  $H_0$  : the time between the occurrence of fire in the reserve follows  $\text{Exp}(1/15)$  and the time between the occurrence of the fire in the reserve does not follow  $\text{Exp}(1/15)$ .

(a)

```
firedate <- c(4, 18, 32, 37, 56, 64, 78, 89, 104, 134, 154, 178, 190, 220, 256)
fireinter <- firedate[-1] - firedate[-15]
# Generate dataset from Exp(1/15)
expy <- rexp(length(fireinter), rate = 1/15)
```

```
# qqplot of the data
qqfire <- qqplot(expy, firedate, main = "QQ-plot for Q7")
abline(lm(qqfire$y ~ qqfire$x))
```



The qqplot doesn't seem to be a straight line so the claim is not justified.

(b)

Kolmogorov-Smirnov test

```
# install.packages('exptest')
require(exptest)

## Loading required package: exptest
ks.exp.test(fireinter)

##
## Kolmogorov-Smirnov test for exponentiality
##
## data: fireinter
## KSn = 0.3144, p-value = 0.0195
```

The P-value of Kolmogorov-Smirnov test is 0.0225 which is significant under 5% significant level. We reject the Null.

(c)

The Anderson-Darling test

```
# install.packages('ADGofTest')
require(ADGofTest)

## Loading required package: ADGofTest

ad.test(fireinter, pexp)

##
## Anderson-Darling GoF Test
##
## data: fireinter and pexp
## AD = 170.4, p-value = 4.286e-05
## alternative hypothesis: NA
```

The P-value of Anderson-Darling test is 4.286e-05 which is significant under 5% significant level. We reject the Null.