```
install.packages('MASS')
library('MASS')
#Part 1
\#P(x;p,n)=(n x)(p)^x(1-p)(n-x)
#P(X=0)=(4!/(0!*4!))*(0.5)^0*(0.5)^4=0.0625
\#P(X=1)=(4!/(1!*3!))*(0.5)^1*(0.5)^3=0.25
\#P(X=2)=(4!/(2!*2!))*(0.5)^2*(0.5)^2=0.375
\#P(X=3)=(4!/(3!*1!))*(0.5)^3*(0.5)^1=0.25
\#P(X=4)=(4!/(4!*0!))*(0.5)^4*(0.5)^0=0.0625
dbinom(0:4,4,.5)
#b
\#P(X<0)
#P(X<0)=P(X=0)=0.0625
pbinom(0,4,.5)
#P(0 < X < 1)
\#P(0<X<1)=P(X=1)+P(X=0)-P(X=0)=0.25+0.0625-0.0625=0.25
pbinom(1,4..5) - pbinom(0,4..5)
#P(1<X<2)
#P(1<X<2)=P(X=2)+P(X=1)+P(X=0)-P(X=1)-P(X=0)=0.375
pbinom(2,4,.5) - pbinom(1,4,.5)
#P(2<X<3)
#P(2<X<3)=P(X=3)+P(X=2)+P(X=1)+P(X=0)-P(X=2)-P(X=1)-P(X=0)=0.25
pbinom(3,4,.5) - pbinom(2,4,.5)
#P(3<X<4)
\#P(3<X<4)=P(X=4)+P(X=3)+P(X=2)+P(X=1)+P(X=0)-P(X=3)-P(X=2)-P(X=1)-P(X=0)=0.0625
pbinom(4,4..5) - pbinom(3,4..5)
\#P(X>4)
\#P(X>4)=1-P(X<4); P(X<4)=P(X=4)+P(X=3)+P(X=2)+P(X=1)+P(X=0)=1; P(X>4)=0
1 - pbinom(4,4,.5)
#c
plot(0:4,dbinom(0:4,4,.5),type='h',ylab='f(x)',xlab='x',main='PMF of Bin(4,0.5)',ylim=c(0.0,0.4))
plot(-2:6,pbinom(-2:6,4,.5),type='s',ylab='F(x)',xlab='x',main='CDF of
Bin(4,0.5)',ylim=c(0.0,1.0),xlim=c(-1,5)
#Part 2
#a
#(5.3-4.7)/(8-2)=0.6/6=0.1=10%
#b
z <- runif(100000,2,8)
length(z[z>4.7&z<5.3])/length(z)
#c
#The P(4.7<x̄<5.3) should be around 76%
#d
ptm <- proc.time()
w <- c()
for (n in 1:100000){
 w <- c(w,mean(runif(48,2,8)))
```

```
proc.time() - ptm
#time to run code was 15.552
length(w[w>4.7&w<5.3])/length(w)
#The P(4.7 < \bar{x} < 5.3) is about 76%
#e
ptm <- proc.time()
colMeans(matrix(runif(4800000,2,8),nrow=48,ncol=100000))
proc.time() - ptm
#time to run code was 0.283
y3 <- colMeans(matrix(runif(4800000,2,8),nrow=48,ncol=100000))
length(y3[y3>4.7&y3<5.3])/length(y3)
#The P(4.7<x<5.3) is about 76%
#f
ptm <- proc.time()
matrix1 <- matrix(data=1,nrow=48,ncol=100000)
matrix1
for (j in 1:100000){
 for (i in 1:48){
  matrix1[i,j]=runif(1,2,8)
colMeans(matrix1)
proc.time() - ptm
#time to run code was 9.214
length(colMeans(matrix1)|colMeans(matrix1)>4.7&colMeans(matrix1)<5.3|)/
length(colMeans(matrix1))
#The P(4.7 < \bar{x} < 5.3) is about 76%
#g
#The most efficient in terms of execution time is clearly approach 2 where a
#matrix of values is created and the columns are averaged.
#Part 3
a <- 1:10
b <- 50:55
riffle <- function(x,y){
 i=1
 v <- c()
 if (\operatorname{length}(x) >= \operatorname{length}(y)) \operatorname{length}(1=y) else \operatorname{length}(1=x)
 if (\operatorname{length}(x) >= \operatorname{length}(y)) \operatorname{length}2 = x else \operatorname{length}2 = b
 while (i-1<length(length1)){
  v \leftarrow c(v,x[i])
  v <- c(v,y[i])
  i < -i + 1
 v <- c(v,length2[i:length(length2)])
riffle(a,b)
```

```
#Part 4
insert <- function(x, where, what){
 if (where \leq length(x)+1) append(x,what,where-1) else{
  print('Warning Message: index exceeds the dimensions of the vector')
  c(x,rep(NA,(where-length(a))-1),what)
insert(a, 15, 100)
insert(a,6,100)
#Part 5
#H0:\mu1=\mu2; Ha:\mu1 \neq \mu2; Ha:\mu1 < \mu2; Ha:\mu1 > \mu2
#The pooled variance estimate is represented by Sp=sqrt(((n1-1)*s1+(n2-1)*s2))/(n1+n2-2)
#where n1 and n2 are the sample sizes and s1 and s2 are the sample variances
#The test statistic is equal to T=\bar{X}1-\bar{X}2/(Sp*sqrt(1/n1+1/n2))
#The null distribution is |T| < t1-\alpha/2, v
#d
#Ha:\mu 1 \neq \mu 2: t(1-\alpha/2,\nu) P-value needs to be greater than T to accept null hypothesis
#Ha:\mu1 < \mu2; t(\alpha,\nu) P-value needs to be greater than T to accept null hypothesis
#Ha:\mu1 > \mu2; t(1-\alpha,\nu) P-value needs to be less than T to accept null hypothesis
test <- function(y1,y2,alt='two-sided',lev=0.95){
 if (alt=='two-sided'){
  add <- 0
  for (a in y1){
    add <- add + (a-mean(y1))^2
  add2 <- 0
  for (b in y2){
    add2 \leftarrow add2 + (b-mean(y2))^2
  test statistic <- (mean(y1)-mean(y2))/(sqrt((add2/(length(y2)-1))/length(y2)+(add/
(length(y1)-1))/length(y1)))
  critical_value <- qt(1-((1-lev)/2),length(y1)+length(y2)-2)
  sd1 <- sqrt(add/length(y1))
  sd2 <- sqrt(add2/length(y2))
  p_value <- (mean(y1)-mean(y2))/sqrt((sd1/length(y1))+sd2/length(y2))
  SE \leftarrow sqrt((((add/(length(y1)-1))/length(y1)) + (add2/(length(y2)-1))/length(y2)))
  CI <- c(mean(y1)-mean(y2)-critical_value*SE,mean(y1)-mean(y2)+critical_value*SE)
  print(CI)
  print(test_statistic)
 }
```

```
if (alt=='less'){
  add <- 0
  for (a in y1){
   add <- add + (a-mean(y1))^2
  add2 <- 0
  for (b in y2){
    add2 <- add2 + (b-mean(y2))^2
  test_statistic <- (mean(y1)-mean(y2))/(sqrt((add2/(length(y2)-1))/length(y2)+(add/
(length(y1)-1))/length(y1)))
  critical_value <- qt(1-((1-lev)),length(y1)+length(y2)-2)
  sd1 <- sqrt(add/length(y1))
  sd2 <- sqrt(add2/length(y2))
  p_value <- (mean(y1)-mean(y2))/sqrt((sd1/length(y1))+sd2/length(y2))
  SE \leftarrow sqrt((((add/(length(y1)-1))/length(y1)) + (add2/(length(y2)-1))/length(y2)))
  CI <- c(mean(y1)-mean(y2)-critical_value*SE,mean(y1)-mean(y2)+critical_value*SE)
  print(CI)
  print(test_statistic)
 if (alt=='greater'){
  add <- 0
  for (a in y1){
   add <- add + (a-mean(y1))^2
  add2 <- 0
  for (b in y2){
   add2 <- add2 + (b-mean(y2))^2
  test_statistic <- (mean(y1)-mean(y2))/(sqrt((add2/(length(y2)-1))/length(y2)+(add/
(length(y1)-1))/length(y1)))
  critical_value <- qt(((1-lev)),length(y1)+length(y2)-2)
  sd1 <- sqrt(add/length(v1))
  sd2 <- sqrt(add2/length(y2))
  p_value <- (mean(y1)-mean(y2))/sqrt((sd1/length(y1))+sd2/length(y2))
  SE \leftarrow sqrt((((add/(length(y1)-1))/length(y1)) + (add2/(length(y2)-1))/length(y2)))
  CI <- c(mean(y1)-mean(y2)-critical_value*SE,mean(y1)-mean(y2)+critical_value*SE)
  print(CI)
  print(test_statistic)
  print(p_value)
}
usa <- subset(Cars93,Origin %in% c('USA'))[['MPG.city']]
non_usa <- subset(Cars93,Origin %in% c('non-USA'))[['MPG.city']]
```

test(c(non\_usa),c(usa),alt='two-sided',lev=0.95)