

# Homework #07

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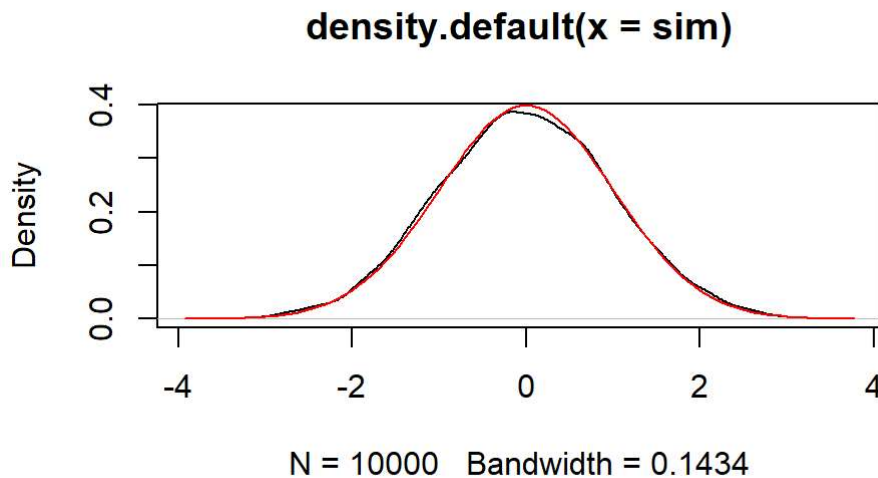
31 Mar 2021

## Chapter 05

### Problem 25

N = 8

```
n<-8
trials<-10000
mu<-.5*(0+1)
sigma<- sqrt((1/12)*(1-0)^2)
sim<-replicate(trials, {x<-runif(n,0,1);
(mean(x)-mu) / (sigma/sqrt(n))
})
plot(density(sim))
curve(dnorm(x),add=TRUE,col="red")
```

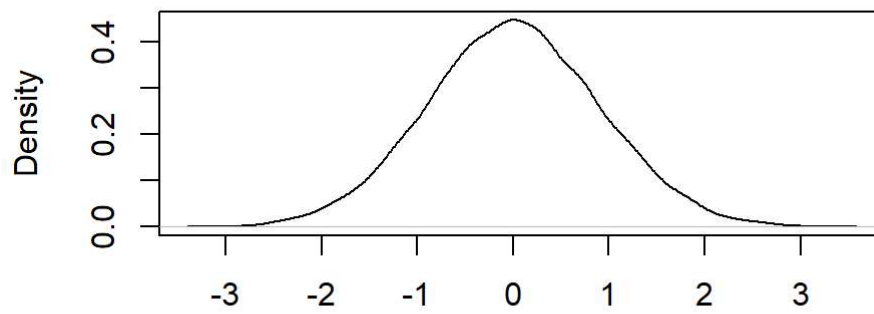


### Problem 26

By inspection,  $\mu = 10$ ,  $\sigma = 2.5$

```
mu<-10
sigma<-2.5
n<-100
sumNorm<-replicate(10000,{x <- replicate(n, x<-sum(rexp(20, 2)));
(mean(x)-mu)/(sigma/sqrt(n))})
plot(density(sumNorm))
```

### density.default(x = sumNorm)



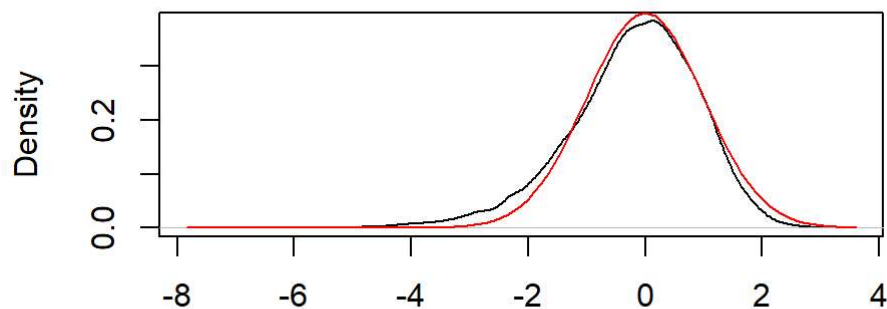
N = 10000 Bandwidth = 0.1285

## Problem 28

N = 36

```
n<-36
trials<-10000
k<-2
mu<-k
sigma<- sqrt(2*k)
sim<-replicate(trials, {x<-rchisq(n,df=k);
(mean(x)-mu) / (sd(x)/sqrt(n))
})
plot(density(sim))
curve(dnorm(x),add=TRUE,col="red")
```

### density.default(x = sim)

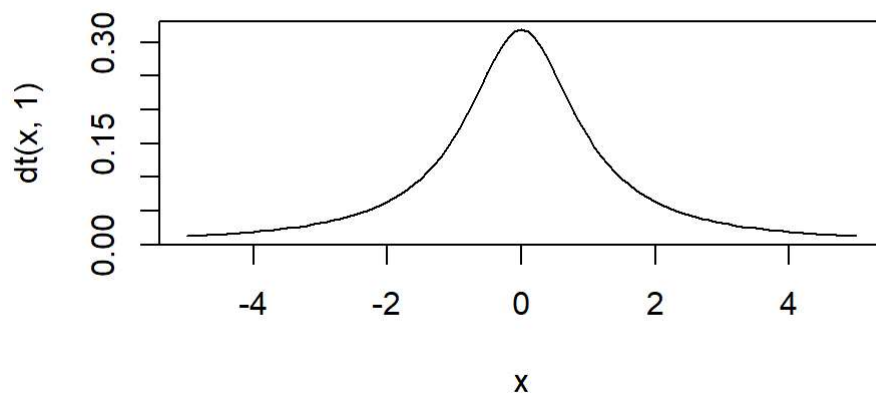


N = 10000 Bandwidth = 0.1509

## Problem 30

a.

```
curve(dt(x, 1), from = -5, to = 5)
```



b.

```
x<-replicate(10,mean(rt(100,1)))
y<-replicate(10,mean(rt(1000,1)))
z<-replicate(10,mean(rt(10000,1)))
x;y;z
```

```
## [1] -10.15873479  0.06135676  0.23603428  0.37500493  0.21981890
## [6]  1.57487308  0.55444057  2.92937889  0.48798601  0.31203998
```

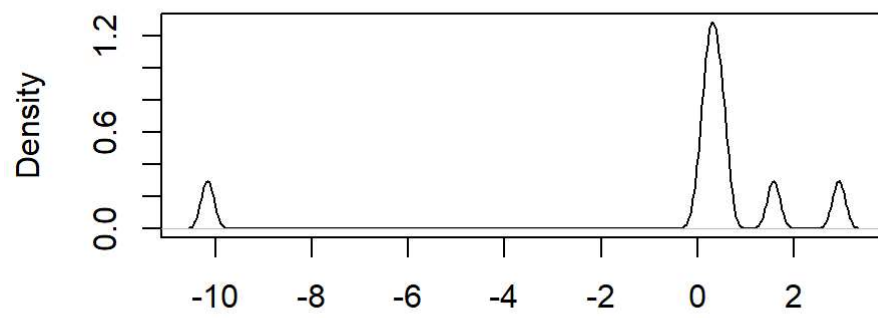
```
## [1] -0.1040474 -2.3421840  1.1200097  0.4597604 -1.3562335  0.2768023
## [7]  1.0946895 -12.7685502 -1.0600909 10.4177791
```

```
## [1] -2.8119947  1.1671858  0.2167064 36.4373808  0.6693422  1.0989255
## [7] -0.3780790 -0.1529971  0.6115447  0.9078466
```

c.

```
plot(density(x))
```

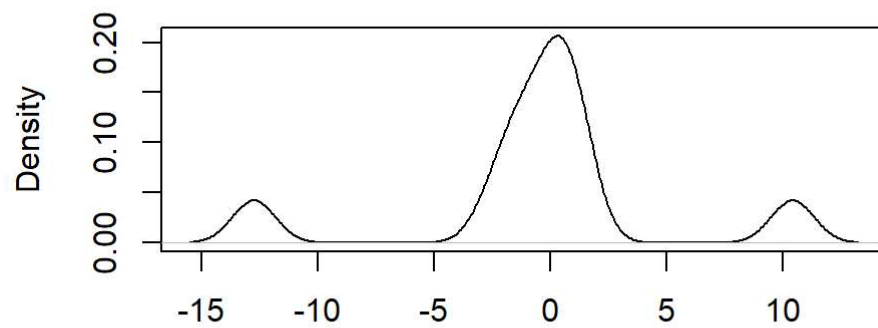
**density.default(x = x)**



N = 10 Bandwidth = 0.133

```
plot(density(y))
```

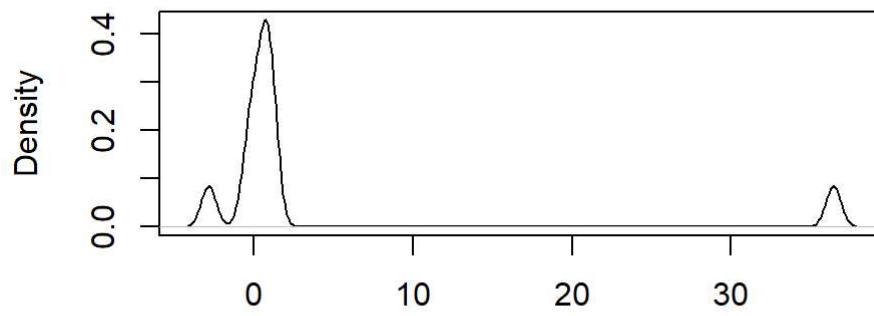
**density.default(x = y)**



N = 10 Bandwidth = 0.94

```
plot(density(z))
```

### density.default(x = z)

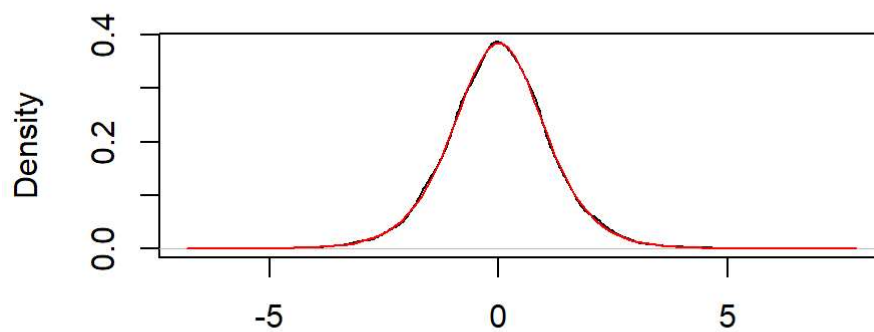


N = 10 Bandwidth = 0.4711

## Problem 34

```
n<-8
mu<-2
sim<-replicate(10000, {x<-rnorm(n,2,3);
(mean(x)-mu) / (sd(x)/sqrt(n))
})
plot(density(sim))
curve(dt(x,df=7),add=TRUE,col="red")
```

### density.default(x = sim)



N = 10000 Bandwidth = 0.1515

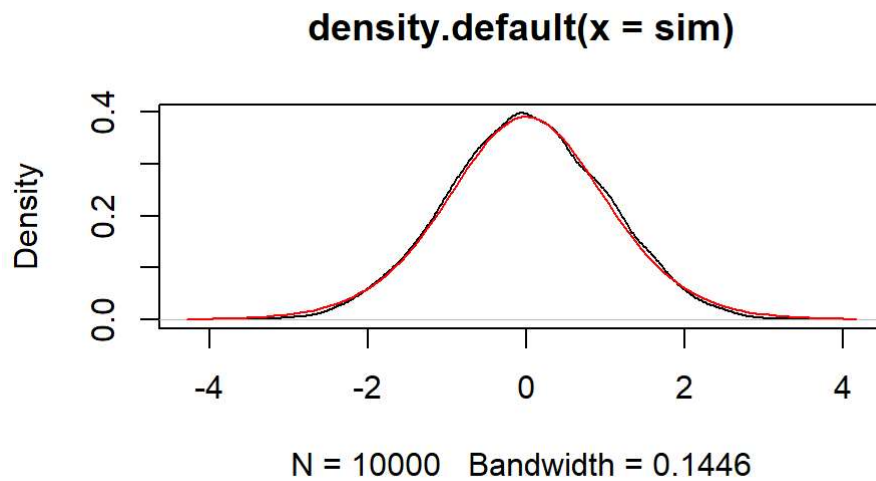
## Chapter 08

### Problem 01

```

n<-12
mu<-1
sigma<-3
sim<-replicate(10000, {x<-rnorm(n,1,sigma);
(mean(x)-mu) / (sigma/sqrt(n))
})
plot(density(sim))
curve(dt(x,df=11),add=TRUE,col="red")

```



## Problem 03

C = 1.439756

```
qt(0.1, df=6)
```

```
## [1] -1.439756
```

## Problem 04

The smaller the degrees of freedom, the more area in the tail, the greater the variance.

```
1-pnorm(2)
```

```
## [1] 0.02275013
```

```
1-pt(2,df=40)
```

```
## [1] 0.02616117
```

```
1-pt(2,df=20)
```

```
## [1] 0.02963277
```

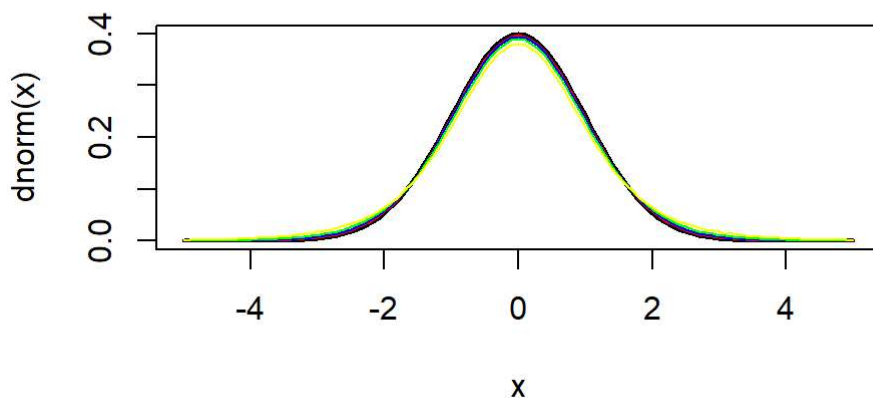
```
1-pt(2,df=10)
```

```
## [1] 0.03669402
```

```
1-pt(2,df=5)
```

```
## [1] 0.05096974
```

```
curve(dnorm(x), from = -5, to = 5, lwd=2)  
curve(dt(x,df=40),add=TRUE,col="red")  
curve(dt(x,df=20),add=TRUE,col="blue")  
curve(dt(x,df=10),add=TRUE,col="green")  
curve(dt(x,df=5),add=TRUE,col="yellow")
```



## Problem 06

99% confidence interval [16.49394,20.0219]

```
plastics <- fosdata::plastics %>% filter(!is.na(diameter))  
diam <- plastics$diameter  
xbar<-mean(diam)  
s<-sd(diam)  
stderror<- s/sqrt(length(diam))  
tcritical = -qt(0.005, df=(length(diam)-1))  
xbar - tcritical * stderror
```

```
## [1] 16.49394
```

```
xbar + tcritical * stderror
```

```
## [1] 20.0219
```

## Problem 07

95% Confidence Interval: [299836.7, 299868.1]

This interval does not include 299729m/s

```
light <- morley$Speed + 299000
xbar<-mean(light)
s<-sd(light)
stderror<- s/sqrt(length(light))
tcritical = -qt(0.025, df=(length(light)-1))
xbar - tcritical * stderror
```

```
## [1] 299836.7
```

```
xbar + tcritical * stderror
```

```
## [1] 299868.1
```

## Problem 11

a -> II.

b -> I.

c -> III.

## Problem 13

a.

Population mean is 2

```
lambda<-0.5
mu <- (1/lambda)
mu
```

```
## [1] 2
```

b.

```
z<-replicate(10000, {y<-t.test(rexp(10,lambda),mu=2)$conf.int;
  (mu >= y[1]) & (mu <=y[2])})
sum(z)/length(z)
```



```
## [1] 0.8996
```

C.

The value is different than 95% because it is pulling a very small sample, which can lead to very skewed means. Upping the sample size (not the replicate trials) increases the accuracy, while the larger number of trials increases precision.

By reducing the replicate trials by a factor of ten and increasing the sample size by the same, we start seeing more accurate but less precise results.

```
z<-replicate(1000, {y<-t.test(rexp(100,lambda),mu=2)$conf.int;  
  (mu >= y[1]) & (mu <=y[2])})  
sum(z)/length(z)
```

```
## [1] 0.95
```