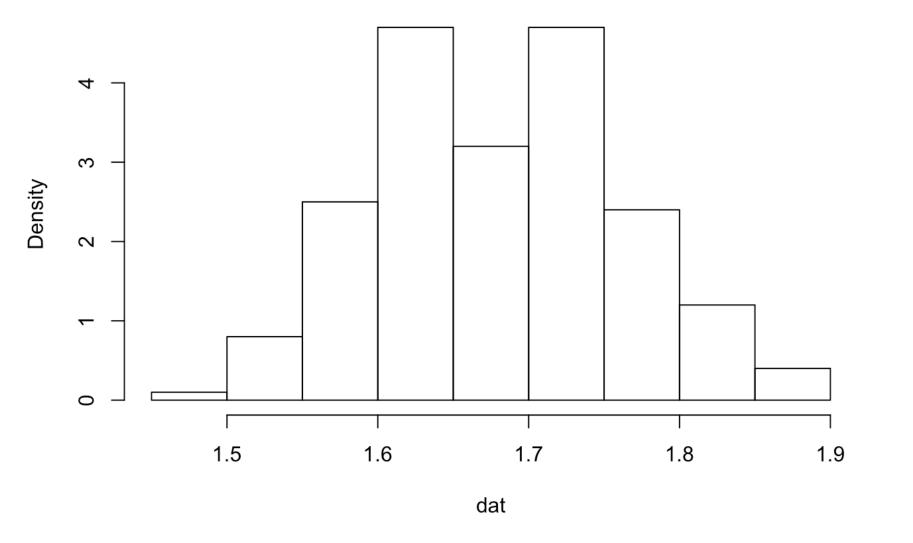
# tutorial3

### Likun Cui(41725041)

### 17 August, 2018

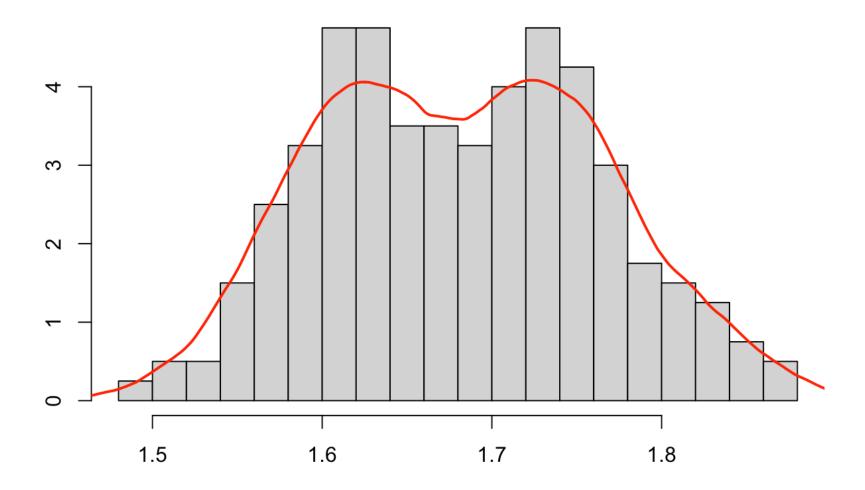
```
#Task1.1 Read in height.txt data
setwd("/Users/likuncui/Downloads/5003/tutorial3/")
dat<- as.numeric(read.table("height.txt",header=T)[,2])
#Task1.2 Generate histogram to summarise and visualise height variable
hist(dat, probability = TRUE, main = "Summarise data density in each bin")</pre>
```

### Summarise data density in each bin



```
#Task2.1 Apply kernel density estimation methods to estimate height density.
#range <- seq(min(dat)-sd(dat), max(dat)+sd(dat), length.out=1000)
# estimate density of x using different kernels
d1 <- density(dat, kernel="epanechnikov")
hist(dat, breaks=20, freq=FALSE, col="lightgray", xlab="", ylab="", main="Epanechnikov( Default h)")
lines(d1, lwd=2, col="red")</pre>
```

### **Epanechnikov( Default h)**

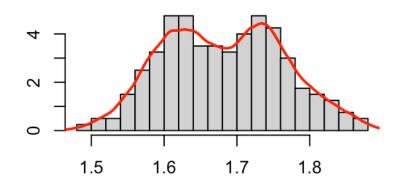


```
h<-0.02
d2<-density(dat,bw= h,kernel="epanechnikov")</pre>
d3<-density(dat,bw=h,kernel="triangular")</pre>
d4<-density(dat,bw=h,kernel="gaussian")</pre>
d5<-density(dat,bw=h,kernel="biweight")</pre>
par(mfrow=c(2,2))
hist(dat, breaks=20, freq=FALSE, col="lightgray", xlab="", ylab="", main="Epanechn
ikov")
lines(d2, lwd=2, col="red")
hist(dat, breaks=20, freq=FALSE, col="lightgray", xlab="", ylab="", main="Triangul
ar")
lines(d3, lwd=2, col="red")
hist(dat, breaks=20, freq=FALSE, col="lightgray", xlab="", ylab="", main="Normal")
lines(d4, lwd=2, col="red")
hist(dat, breaks=20, freq=FALSE, col="lightgray", xlab="", ylab="", main="Biweight
")
lines(d5, lwd=2, col="red")
```

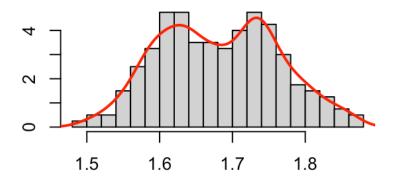
### **Epanechnikov**

# 1.5 1.6 1.7 1.8

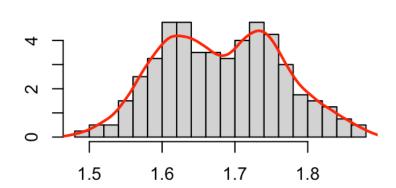
### Triangular



### **Normal**



### **Biweight**



#Task3.1 Use BCV methods to select for optimal bandwidth for each kernel of choice .  $h <- c(0.3,\ 0.625,\ 1.875)$  h.bcv <- bw.bcv(dat)

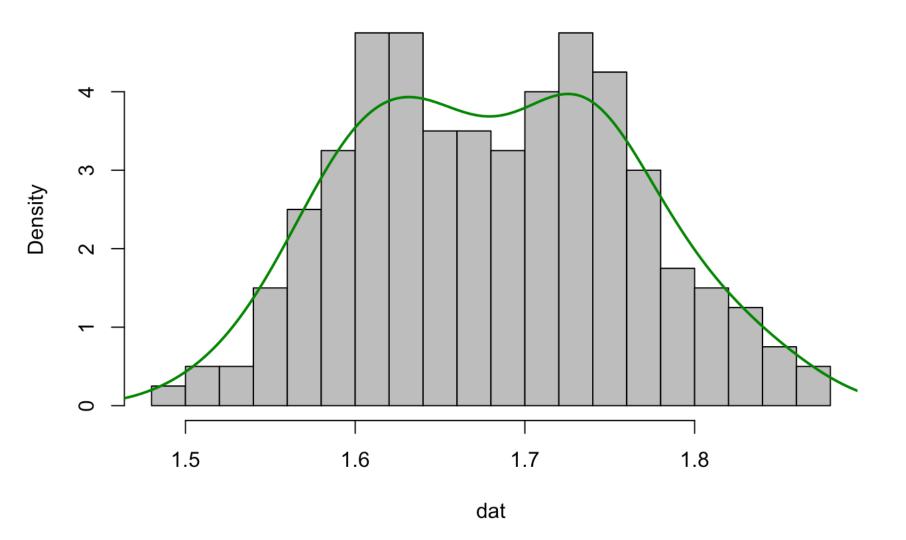
## Warning in bw.bcv(dat): minimum occurred at one end of the range

d.bcv <- density(dat, bw=h.bcv, kernel="gaussian")
h.bcv</pre>

## [1] 0.03196391

## Plot estimation results
par(mfrow=c(1,1))
hist(dat, breaks=20, freq=FALSE, col="gray")
lines(d.bcv, col="green4", lwd=2)

### Histogram of dat



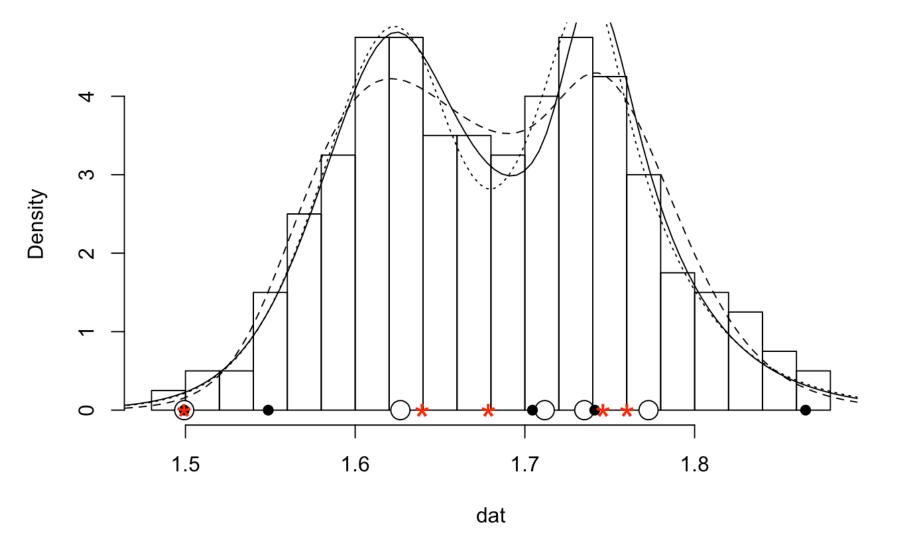
#Task4.1 Apply cubic spline density estimation with different number of knots
fit1 <- logspline(dat)
hist(dat,breaks=20,freq=FALSE)
fit1</pre>

```
##
    knots A(1)/D(2) loglik
                                AIC minimum penalty maximum penalty
                   2 211.95 -408.00
##
        4
                                                26.78
                                                                   Inf
##
        5
                   2 225.34 -429.48
                                                 2.76
                                                                 26.78
##
        6
                   2 225.36 -424.23
                                                   NA
                                                                    NA
##
        7
                   2 228.09 -424.40
                                                 0.43
                                                                  2.76
                   2 228.12 -419.15
##
        8
                                                   NA
                                                                    NA
                   2 228.53 -414.67
                                                 0.31
                                                                  0.43
##
        9
                   2 228.68 -409.68
                                                 0.26
                                                                  0.31
##
       10
                   2 228.81 -404.64
##
       11
                                                 0.02
                                                                  0.26
##
       12
                   2 228.82 -399.37
                                                 0.01
                                                                  0.02
                   1 228.83 -394.08
##
                                                 0.00
       13
                                                                  0.01
## the present optimal number of knots is
                                             5
## penalty(AIC) was the default: BIC=log(samplesize): log( 200 )= 5.3
```

```
plot(fit1, add=T)
points(fit1$knots,rep(0,5),pch=21,cex=2,bg="white")

fit2 <- logspline(dat, nknots=3)
fit3 <- logspline(dat, nknots=7)
plot(fit2, add=T,lty=2)
plot(fit3, add=T,lty=3)
points(fit2$knots,rep(0,5),pch=21,cex=1,bg="black")
points(fit3$knots,rep(0,5),pch="*",cex=2, col="red")</pre>
```

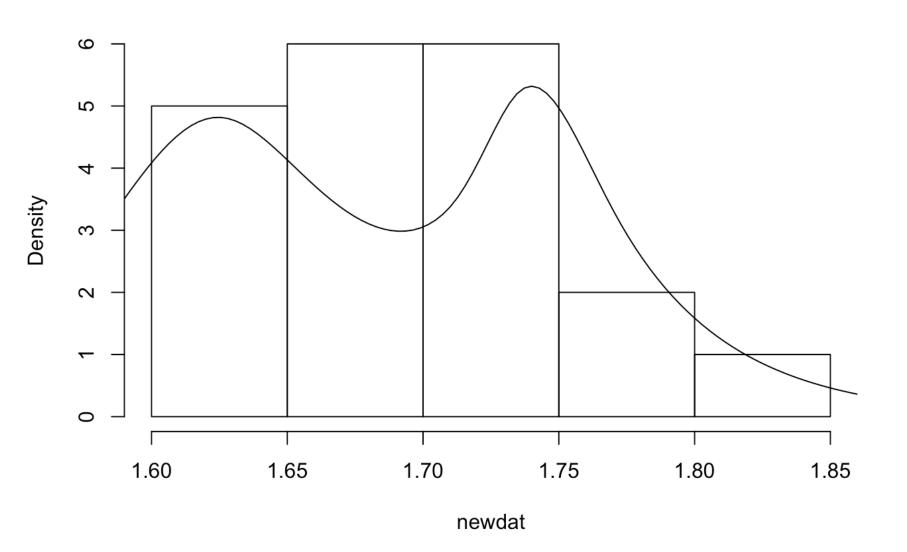
### Histogram of dat



#Task4.2 Compare these results to those from using kernel functions.

#Task5.1 Read in newHight.txt data.
newdat<- as.numeric(read.table("newHeight.txt",header=T)[,2])
hist(newdat, probability = TRUE, main = "Summarise data density using previous mod el")
#classify each of these new samples based on density estimation results from heigh t.txt
plot(fit1, add=T)</pre>

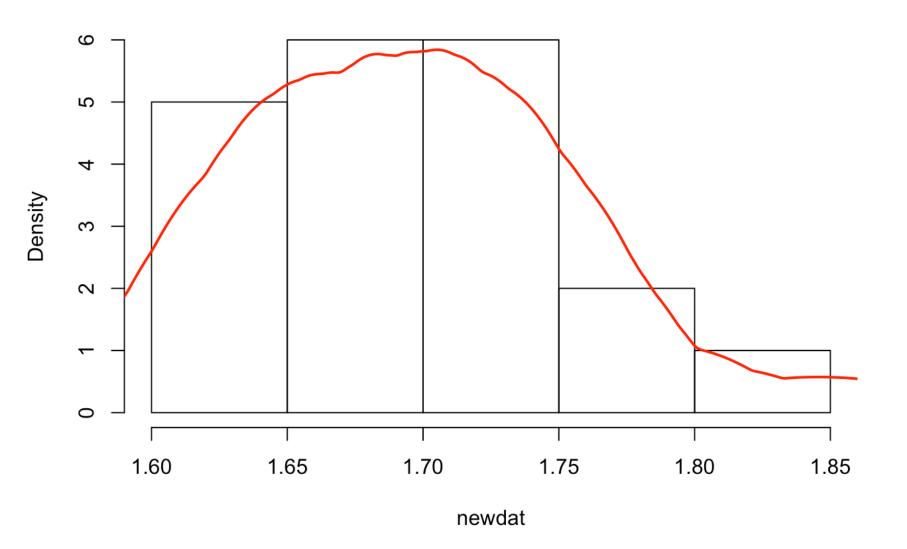
## Summarise data density using previous model



```
#the perfect prediction should be the following plot:
hist(newdat, probability = TRUE, main = "Summarise data density by default functio
n")

d7 <- density(newdat, kernel="epanechnikov")
lines(d7, lwd=2, col="red")</pre>
```

### Summarise data density by default function



print("since the are totally deffences between the two curves, although it could b
e used as classification, it not a proper result.")

## [1] "since the are totally deffences between the two curves, although it could be used as classification, it not a proper result."