

# Lab 9: Likelihoods

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## Part I

1. Loading the data from the MASS package.

```
library(MASS)
```

```
## Warning: package 'MASS' was built under R version 3.1.3
```

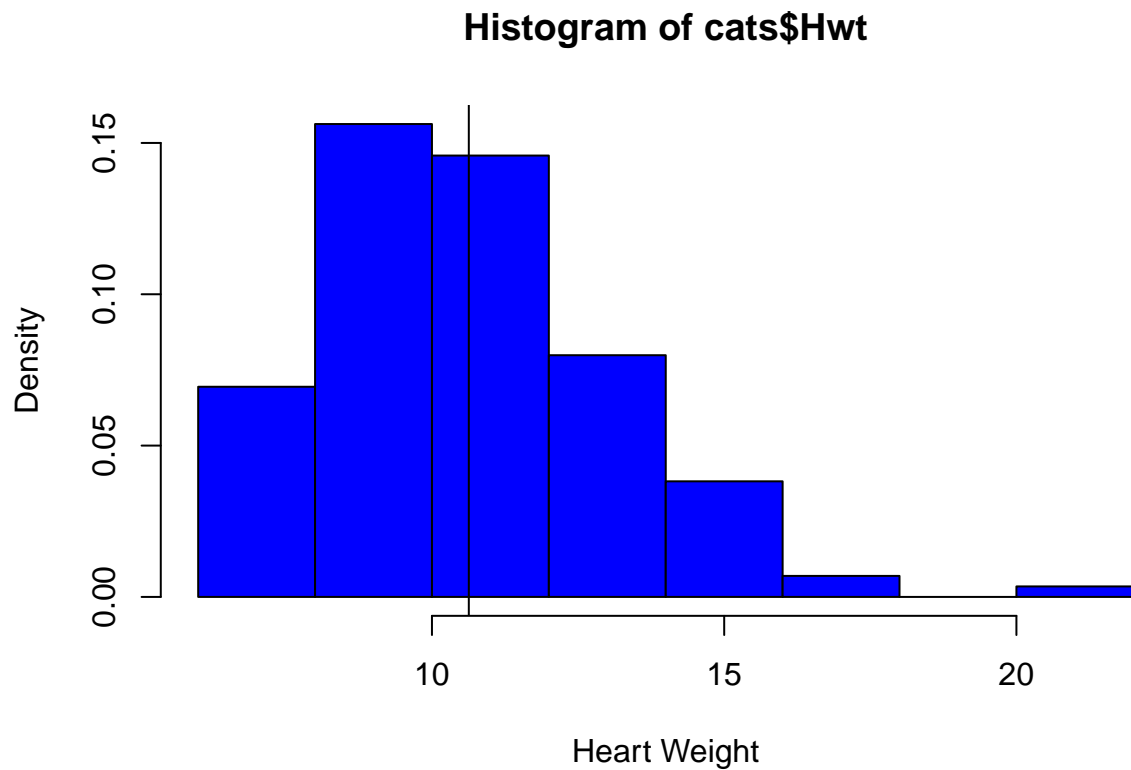
```
data(cats)
summary(cats)
```

```
## Sex      Bwt      Hwt
## F:47  Min.   :2.000  Min.    : 6.30
## M:97  1st Qu.:2.300  1st Qu.: 8.95
##      Median :2.700  Median :10.10
##      Mean   :2.724  Mean   :10.63
##      3rd Qu.:3.025  3rd Qu.:12.12
##      Max.   :3.900  Max.   :20.50
```

## Question 1

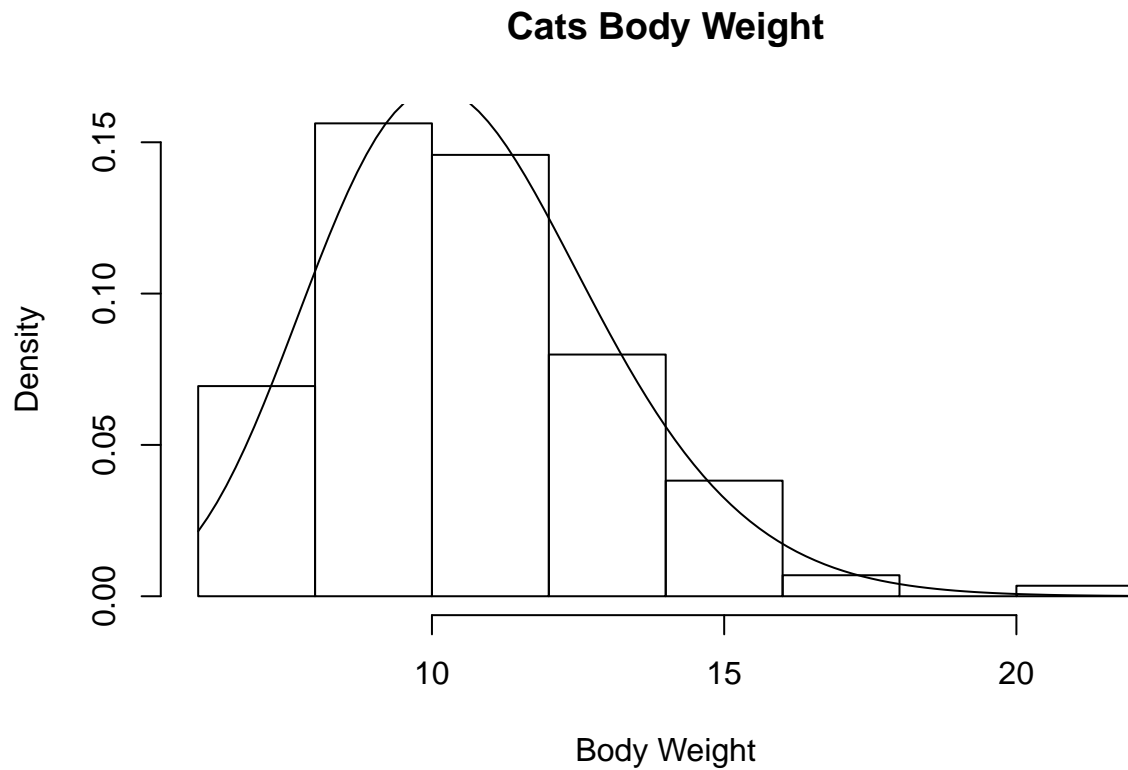
Fitting the gamma distribution to the cats' hearts, using the “method of moments”

```
# making the histogram
hist(cats$Hwt, xlab="Heart Weight", col="blue", probability=TRUE)
abline(v=mean(cats$Hwt))
```



```
# function for getting the shape and the scale of the for the gamma distribution
cat.stats <- function(vec)
{
  m<-mean(vec)
  var <- var(vec)
  a<-(m^2)/var
  s<-var/m
  return(c("Mean"=m,"Variance"=var,"a"=a,"s"=s))
}

# now plot the histogram with the gamma distribution given the shape and the scale
hist(cats$Hwt,main='Cats Body Weight',xlab='Body Weight',probability=TRUE)
curve(dgamma(x,shape=cat.stats(cats$Hwt)[3],scale=cat.stats(cats$Hwt)[4]),add=TRUE)
```



## Question 2

The log-likelihood of the shape and scale parameters we just estimated.

```
# get the log likelihood
sum(dgamma(cats$Hwt,shape=cat.stats(cats$Hwt)[3],scale=cat.stats(cats$Hwt)[4],log=T))
```

```
## [1] -325.6886
```

## Question 3

A function, `gamma.loglike`, which takes in a shape and a scale parameter (in that order), and returns the log-likelihood of the cats' hearts' masses

```
# function that gets the gamma log likelihood given the shape and the scale
gamma.loglike <- function(shape,scale){
  gam.out <- function(x)
  {
    loglik.gam<- sum(dgamma(x,shape=shape,scale=scale,log=T))
    return(loglik.gam)
  }
  return(gam.out)
}

# calling the function on the cats data to make sure we get the same result as question 2
cats.heart<-gamma.loglike(shape=cat.stats(cats$Hwt)[3],scale=cat.stats(cats$Hwt)[4])
body(cats.heart)
```

```
## {
##   loglik.gam <- sum(dgamma(x, shape = shape, scale = scale,
##     log = T))
##   return(loglik.gam)
## }
```

```
loglike.cats.heart<-cats.heart(cats$Hwt)
paste(loglike.cats.heart)
```

```
## [1] "-325.688619634947"
```

## Question 4

In the first plot, we vary the shape parameter over the range 1 to 40, while holding the scale parameter fixed at the value you estimated in problem 1. In the second, we vary the scale parameter over the range 0.01 to 1, holding shape fixed at your estimated value.

```
# set graph environment
par(mfrow=c(1,2))

# vectorize your results so you can call your function on a range of values
ll.v <- Vectorize(function(shape,scale,data=cats$Hwt){sum(dgamma(data,shape=shape,scale=scale,log=T))},v)

# call the function on the shape ranging from 1 to 40 and plot it
max(ll.v(shape=seq(from=1,to=40,by=20),scale=cat.stats(cats$Hwt)[4], data=cats$Hwt))
```

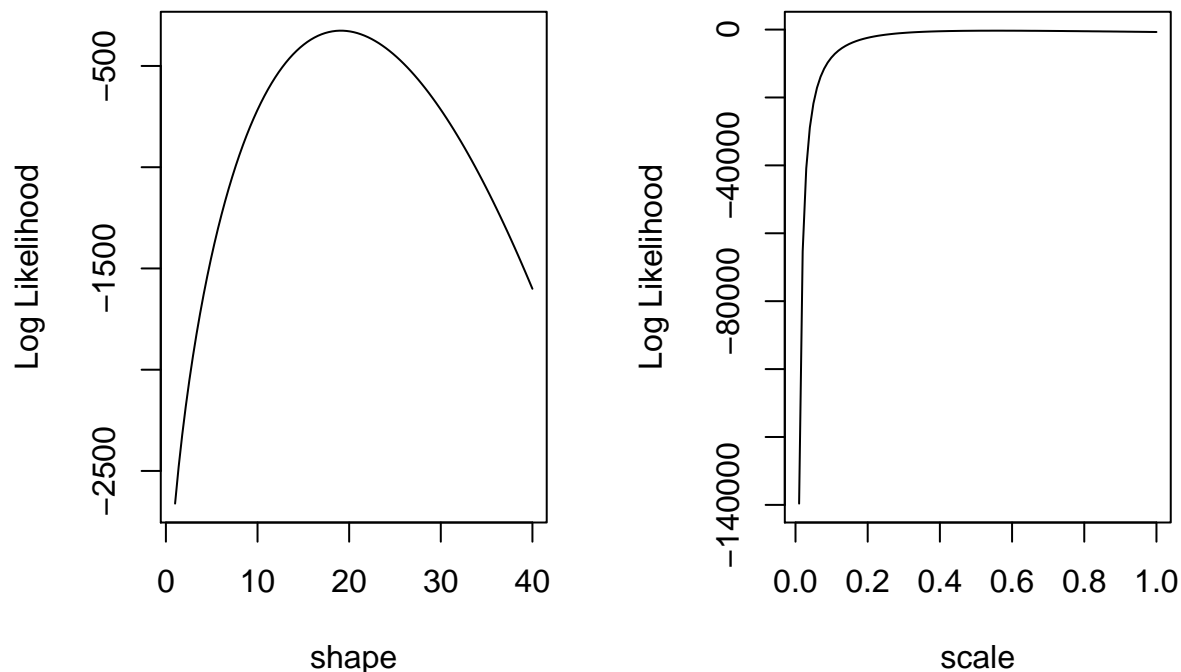
```
## [1] -339.2703
```

```
curve(ll.v(shape=x,scale=cat.stats(cats$Hwt)[4],data=cats$Hwt),from=1,to=40,xlab="shape",ylab="Log Likelihood",lty=1)

# call the function on the scale ranging from 0.01 to 1 and plot it
max(ll.v(scale=seq(from=0.01,to=1,by=0.1),shape=cat.stats(cats$Hwt)[3], data=cats$Hwt))
```

```
## [1] -336.444
```

```
curve(ll.v(scale=x,shape=cat.stats(cats$Hwt)[3],data=cats$Hwt),from=0.01,to=1,xlab="scale",ylab="Log Likelihood",lty=1)
```



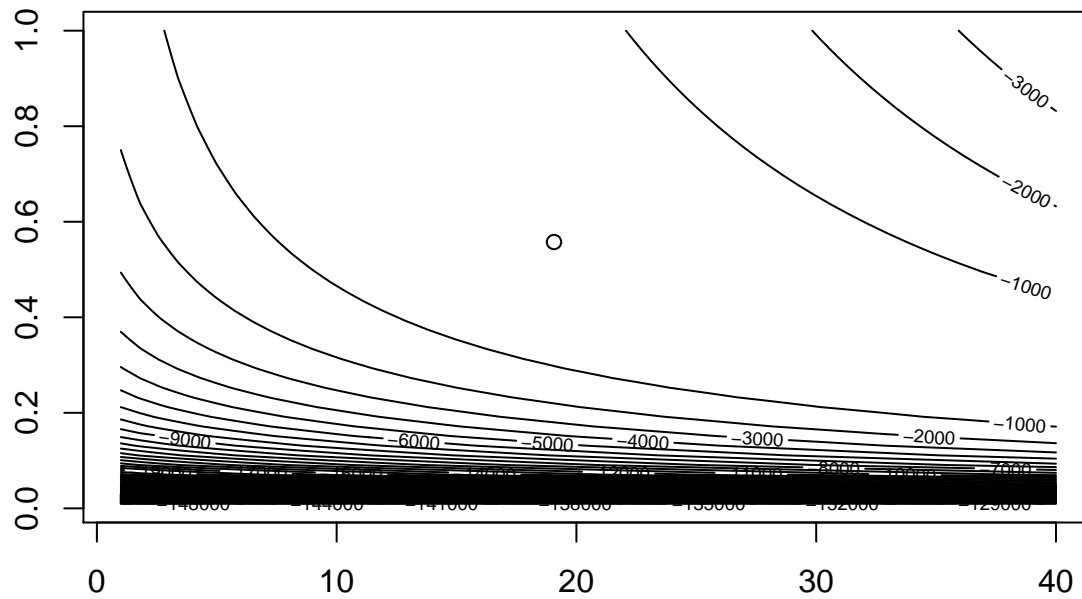
Both the functions peak at around -336 to -339. Yes they should be because that is the maximum of the log likelihood given by the method of moments. The value of the shape corresponding to that peak is around 19-20. The value for the scale seems to be around 0.5.

## Question 5

The function makes a contour plot of the log-likelihood, with the shape parameter on the horizontal axis (range 1 to 40) and the scale parameter on the vertical (range 0.01 to 1). Then I add a point indicating the location of your moment-based estimate from question 1.

```
# function for the contours that evaluates and expression for the given inputs
surface.contour <- function(expr,from.x=0,to.x=40,from.y=0,to.y=1,n.x=50,
  n.y=50,...) {
  x.seq <- seq(from=from.x,to=to.x,length.out=n.x)
  y.seq <- seq(from=from.y,to=to.y,length.out=n.y)
  unevaluated.expression <- substitute(expr)
  z <- function(x,y) {
    return(eval(unevaluated.expression,envir=list(x=x,y=y)))
  }
  z.values <- outer(X=x.seq,Y=y.seq,FUN=z)
  z.matrix <- matrix(z.values,nrow=n.x)
  contour(x=x.seq,y=y.seq,z=z.matrix,...)
  invisible(list(x=x.seq,y=y.seq,z=z.matrix, func=z))
}

# creat the plot
par(mfrow=c(1,1))
surface.contour(ll.v(shape=x,scale=y),from.x=1,to.x=40,from.y=0.01,to.y=1,nlevels=200)
# add the point
points(cat.stats(cats$Hwt)[3],cat.stats(cats$Hwt)[4])
```



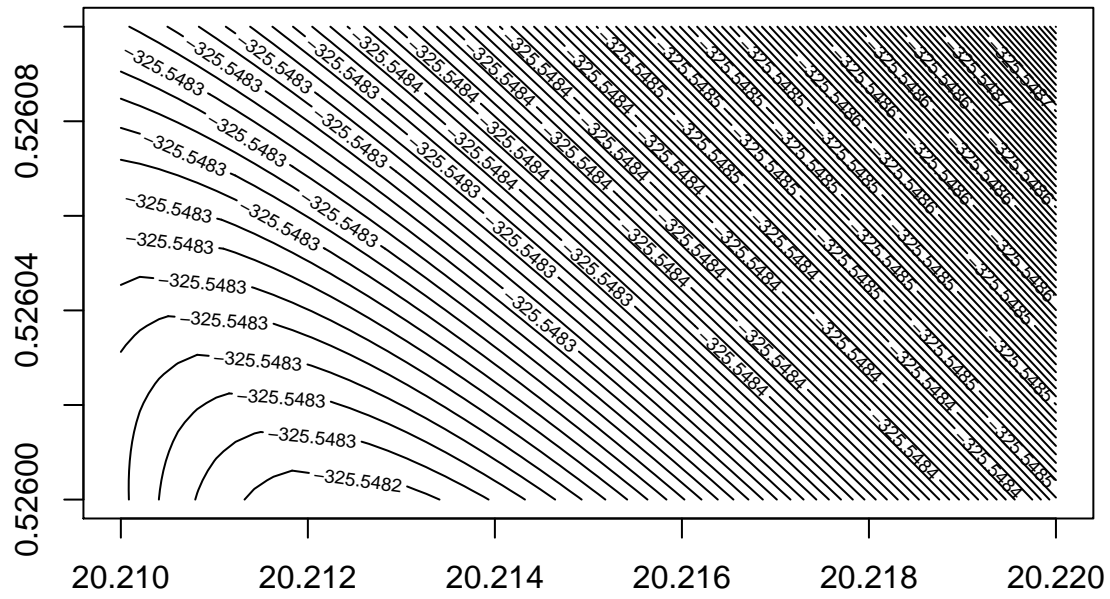
## Question 6

I can't really tell from the contour from the above part but we can see that if we zoom in closer we might be able to see other contours. According to part 4 we should the maximum to occur when  $a \approx 19$  and the  $s \approx 0.56$ . The maximum log likelihood should be around -326. So let's have a closer look.

## Question 7

Locate the region where the likelihood seems to be largest by changing the ranges over which the shape and scale vary.

```
par(mfrow=c(1,1))
surface.contour(ll.v(shape=x,scale=y),from.x=20.21,to.x=20.22,from.y=0.526,to.y=0.5261,nlevels=100)
```



### Question 8

After looking more closely we that we can reach a higher value which would be around -3.25.5482. I believe we can zooming in more and find a closer value. This would be for a  $\approx 20.212$  and  $s \approx 0.526$