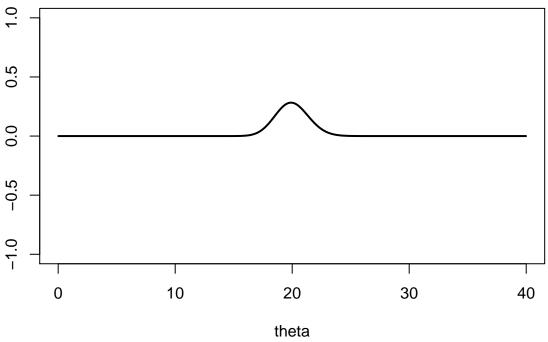
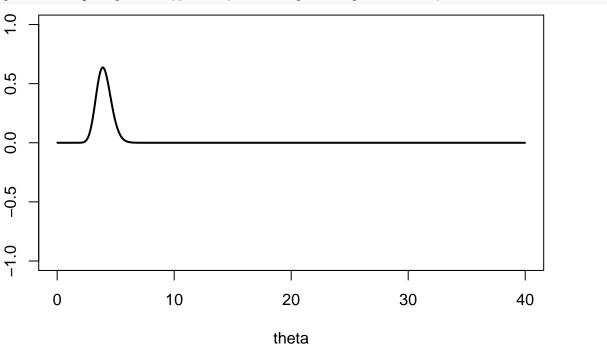
# Question 7 and 10

#### Question 7

```
#####
#Question 7
#####
#a
data <- read.table("data/planes.txt",header=T)</pre>
attach(data)
sumfatal <- sum(fatal)</pre>
n <- length(fatal)</pre>
##### looking at different Gamma priors #####
theta <- ppoints(1000)*40
gammaprior1 <- dgamma(theta, shape=0, rate=0)</pre>
minplot <- min(gammaprior1)</pre>
maxplot <- max(gammaprior1)</pre>
plot(theta,gammaprior1,type="l",ylim=c(minplot,maxplot),lwd=2,ylab="")
0.5
-1.0
       0
                         10
                                            20
                                                              30
                                                                                 40
                                          theta
#set the prior to be stronger
gammaprior1 <- dgamma(theta, shape=200, rate=10)</pre>
plot(theta,gammaprior1,type="l",ylim=c(minplot,maxplot),lwd=2,ylab="")
```



```
#set the prior to be off
gammaprior1 <- dgamma(theta,shape=40,rate=10)
plot(theta,gammaprior1,type="l",ylim=c(minplot,maxplot),lwd=2,ylab="")</pre>
```



```
## now plot the posterior distribution
gammaposterior1 <- dgamma(theta,shape=(sumfatal+0),rate=(n+0))
minplot <- min(gammaposterior1)
maxplot <- max(gammaposterior1)
hist(fatal,xlim=c(0,40),ylim=c(minplot,maxplot),prob=T,col="gray",xlab="",ylab="",main="")
par(new=T)
plot(theta,gammaposterior1,type="l",xlim=c(0,40),ylim=c(minplot,maxplot),lwd=2,xlab="",ylab="",main="")</pre>
```

```
legend("topright",c("Gamma(0,0)"),col=c(1:4),lwd=2)
0.25
                                                                    Gamma(0,0)
0.20
0.15
0.10
0.05
0.00
       0
                         10
                                          20
                                                             30
                                                                              40
#b
###use samples from the posterior distribution to obtain samples from the
#posterior predictive distribution
gammaposterior1 <- dgamma(theta, shape=(sumfatal+0), rate=(n+0))</pre>
minplot <- min(gammaposterior1)</pre>
maxplot <- max(gammaposterior1)</pre>
hist(fatal,xlim=c(0,40),ylim=c(minplot,maxplot),prob=T,col="gray",xlab="",ylab="",main="")
par(new=T)
plot(theta,gammaposterior1,type="1",xlim=c(0,40),ylim=c(minplot,maxplot),lwd=2,xlab="",ylab="",main="")
legend("topright",c("Gamma(0,0)"),col=c(1:4),lwd=2)
0.25
                                                                    Gamma(0,0)
0.20
0.15
0.05
0.00
       0
                         10
                                          20
                                                             30
                                                                              40
##compute a 95% interval
theta <- rgamma(1000, 238)/10
```

```
y1986 <- rpois(1000,theta)
print(sort(y1986)[c(25,976)])</pre>
```

## [1] 14 34

#### Question 10

```
data <- read.table("data/planes.txt",header=T)
attach(data)

## The following objects are masked from data (pos = 3):

##

## deaths, fatal, rate, year

sumfatal <- sum(fatal)

n <- length(fatal)

## a. Choose a non-informative prior

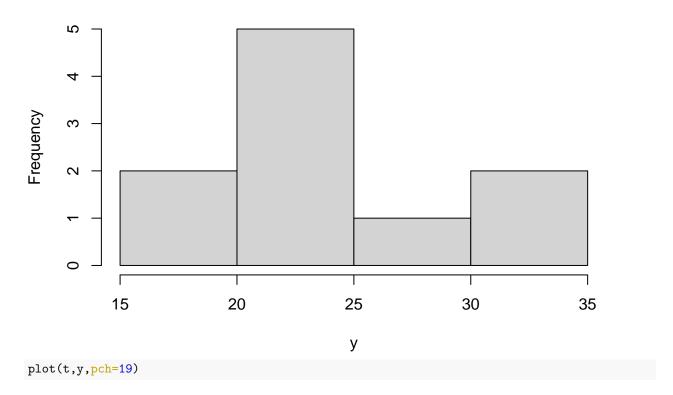
## input data:
data <- read.table("data/planes.txt",skip=1)

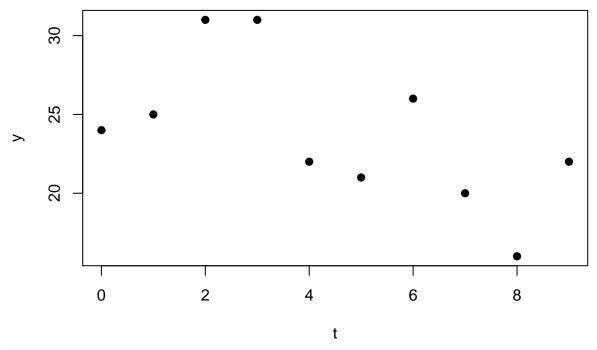
y <- data[,2]

t <- data[,1]-1976

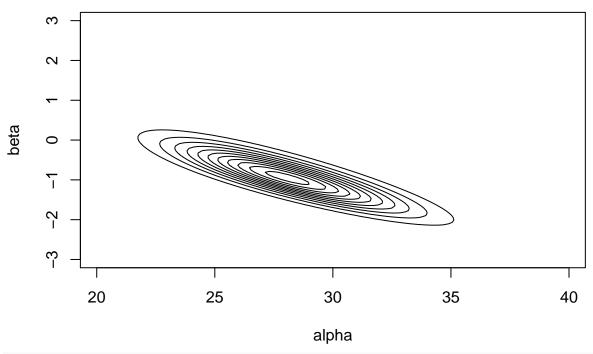
n <- length(y)
hist(y)</pre>
```

# Histogram of y





```
## graphing posterior over range of alpha and beta:
posteriorplanes <- function(alpha, beta){</pre>
  logpost <- -Inf</pre>
  if (alpha + beta*max(t) > 0){
    logpost <- 0</pre>
    for (i in 1:n){
      logpost <- logpost + y[i]*log(alpha+beta*t[i])</pre>
      logpost <- logpost - (alpha+beta*t[i])</pre>
    }
  }
  logpost
}
numgrid <- 100
alpharange <- ppoints(numgrid)*20 # alpha between 0 and 20</pre>
betarange <- ppoints(numgrid)*6 # beta between 0 and 6
numgrid <- 100
alpharange <- ppoints(numgrid)*20+20 # alpha between 20 and 40</pre>
betarange <- ppoints(numgrid)*6-3 # beta between -3 and 3
full <- matrix(NA, nrow=numgrid, ncol=numgrid)</pre>
for (i in 1:numgrid){
  for (j in 1:numgrid){
    full[i,j] <- posteriorplanes(alpharange[i],betarange[j])</pre>
  }
full <- exp(full - max(full))</pre>
full <- full/sum(full)</pre>
contour(alpharange,betarange,full,xlab="alpha",ylab="beta",drawlabels=F)
```

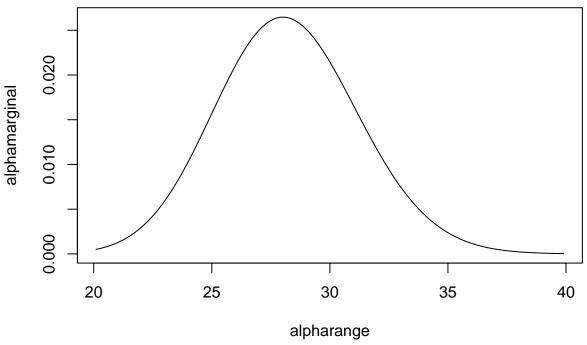


```
## calculating probabilities for grid sampler:

alphamarginal <- rep(NA,numgrid)
for (i in 1:numgrid){
    alphamarginal[i] <- sum(full[i,])
}
betaconditional <- matrix(NA,nrow=numgrid,ncol=numgrid)
for (i in 1:numgrid){
    for (j in 1:numgrid){
        betaconditional[i,j] <- full[i,j]/sum(full[i,])
    }
}

## plotting marginal distribution of alpha
par(mfrow=c(1,1))
plot(alpharange,alphamarginal,type="l",main="marginal dist. of alpha")</pre>
```

### marginal dist. of alpha

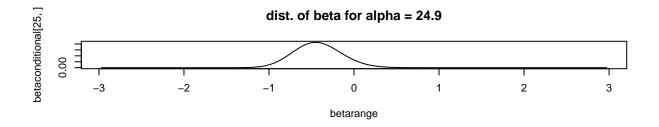


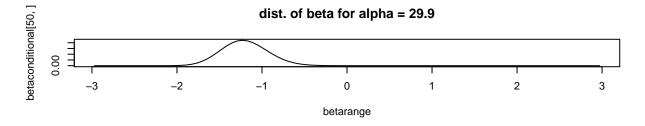
```
## plotting conditional distribution of beta given alpha
alpharange[25]
## [1] 24.9
```

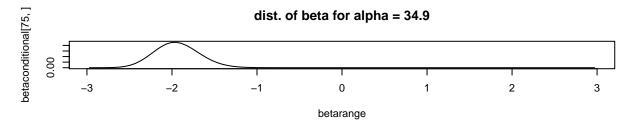
alpharange[50]

## [1] 29.9 alpharange[75]

```
## [1] 34.9
par(mfrow=c(3,1))
plot(betarange,betaconditional[25,],type="l",main="dist. of beta for alpha = 24.9")
plot(betarange,betaconditional[50,],type="l",main="dist. of beta for alpha = 29.9")
plot(betarange,betaconditional[75,],type="l",main="dist. of beta for alpha = 34.9")
```

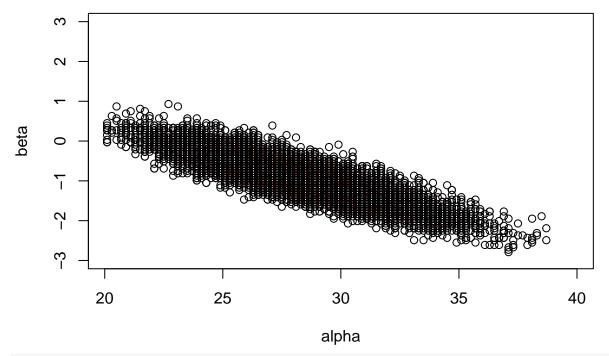






```
## sampling grid values:
alpha.samp <- rep(NA,10000)
beta.samp <- rep(NA,10000)
for (m in 1:10000){
    a <- sample(1:100,size=1,replace=T,prob=alphamarginal)
    b <- sample(1:100,size=1,replace=T,prob=betaconditional[a,])
    alpha.samp[m] <- alpharange[a]
    beta.samp[m] <- betarange[b]
}

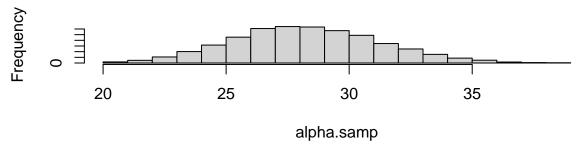
par(mfrow=c(1,1))
contour(alpharange,betarange,full,xlab="alpha",ylab="beta",drawlabels=F,col=2)
points(alpha.samp,beta.samp)</pre>
```



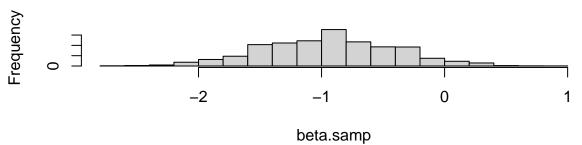
 $\ensuremath{\mbox{\#\#}}$  calculating posterior means/intervals for alpha and beta

```
par(mfrow=c(2,1))
hist(alpha.samp,main="Alpha Samples")
hist(beta.samp,main="Beta Samples")
```

## **Alpha Samples**



## **Beta Samples**



```
mean(alpha.samp)
## [1] 28.26966
mean(beta.samp)
## [1] -0.943464
alpha.sampsort <- sort(alpha.samp)</pre>
beta.sampsort <- sort(beta.samp)</pre>
alpha.sampsort[250]
## [1] 22.5
alpha.sampsort[9750]
## [1] 34.3
beta.sampsort[250]
## [1] -2.01
beta.sampsort[9750]
## [1] 0.09
sum(beta.samp >= 0)/10000
## [1] 0.0408
par(mfrow=c(1,1))
plot(t,y,pch=19)
for (i in 1:1000){
  abline(alpha.samp[i],beta.samp[i],col=3)
points(t,y,pch=19)
     25
     20
                            2
            0
                                            4
                                                            6
                                                                           8
```

t

##b. Use grid sampling to obtain 1000 samples from the joint posterior distribution. Give the 2D counto ##c. Use the samplies from (b) to obtain 1000 samples from the posterior predictive distribution for y\* ##d. Calculate a 95% posterior predictive interval for y\* and compare it to the interval found in Quest ## now plot the posterior distribution