

# Take home quesiton

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a)

```
D_n <- function(theta, q , data ){  
  n<- length(data)  
  return((sum(abs(theta-data)-data)/(2*n) + (1-2*q)*theta/2)*0.05)  
}
```

b)

```
norwegian.fire <- fread('https://www.math.wustl.edu/~nasyring/475/norwegianfire.txt')  
X.old <- norwegian.fire$V1[norwegian.fire$V2 == 89]/500  
X <- norwegian.fire$V1[norwegian.fire$V2 == 90]/500
```

c)

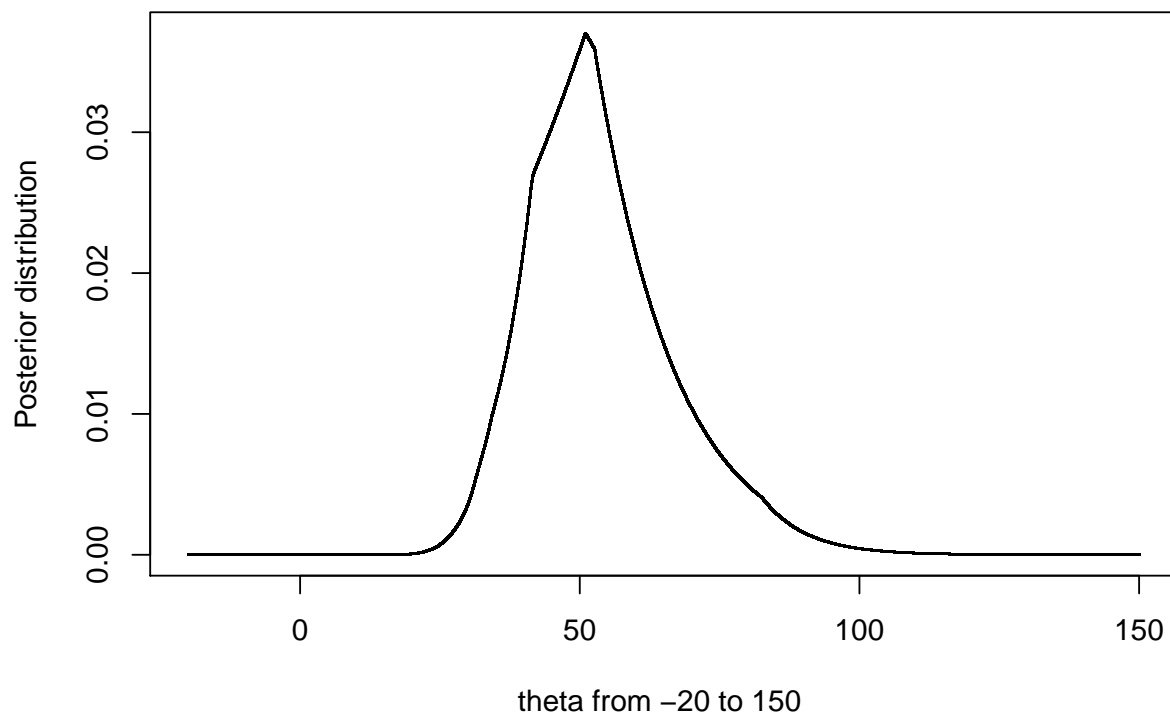
```
p.shape <- 2  
p.scale <- as.numeric(quantile(X.old, 0.995)/2)  
  
numerator <- function(theta,q, data){return(exp(-length(data)*D_n(theta,0.995,data))*  
                                              dgamma(theta,shape = p.shape,scale = p.scale))}  
  
Simpson_rule_exam <- function(fun, up, lo, n, q, data){  
  if(lo>up){  
    c <- up  
    up <- lo  
    lo <- c  
  }  
  xi_s <- lo+ c(1:n)/n *(up-lo)  
  h <- (up-lo)/n  
  sum <- 0  
  for (i in 1:(n-1)) {  
    sum <- sum+ h/6 * (fun(xi_s[i],q,data)+4* fun((xi_s[i]+xi_s[i+1])/2,q,data) +  
                      fun(xi_s[i+1],q,data))  
  }  
  
  return(sum)  
}  
  
denominator <- Simpson_rule_exam(numerator,-1000,1000,10000,q,X)  
  
posterior_exam <- function(theta, q,data) {  
  numerator(theta,q,data)/ denominator  
}
```

d)

```
theta_is <- seq(-20,150,by = 0.001) %>% data.frame()

apply(theta_is,1, function(x){posterior_exam(x,0.995,X)}) -> temp_result

plot(seq(-20,150,by = 0.001),temp_result,pch='.', xlab = 'theta from -20 to 150 ', ylab = 'Posterior di
```



e)

```
# using the gradient descend to find the minimum of negative posterior distribution

gradient_neg_posterior <- function(theta,q,data){

  dn <- D_n(theta,q,data)
  n <- length(data)

  return((-n*exp(-n*dn)*(sum(sign(theta-data))/(2*n) + (1-2*q)/2)*0.05*
    dgamma(theta, scale = p.scale, shape = p.shape)+
    ((p.shape-1)/theta^2 - 1/(theta*p.scale))*
    theta^(p.shape)*exp(-theta/p.scale) *exp(-n*dn) *
    1/(factorial(p.shape-1)*p.scale^p.shape))/denominator)
}
```

```

graden_descend <- function(init_theta,grad,tol, max_it, learning_rate ,q, data){
  old_theta <- init_theta
  n_it <- 0
  rea_tol <- tol*2
  while (rea_tol >tol & n_it < max_it) {
    new_theta <- old_theta - grad(old_theta,q,data)*learning_rate
    rea_tol <- abs(grad(new_theta,q,data))
    n_it <- n_it+1
    old_theta <- new_theta
    if(grad(new_theta,q,data)>grad(old_theta,q,data)){learning_rate = learning_rate/2}
  }
  return(list(solution = new_theta, n_it = n_it, last_tol = rea_tol))
}

```

```

graden_descend(40, gradient_neg_posterior, 10(-4), 10000,1, 0.995, X)

```

```

## $solution
## [1] 51.0183488470972
##
## $n_it
## [1] 10000
##
## $last_tol
## [1] 0.000676835688413607

```