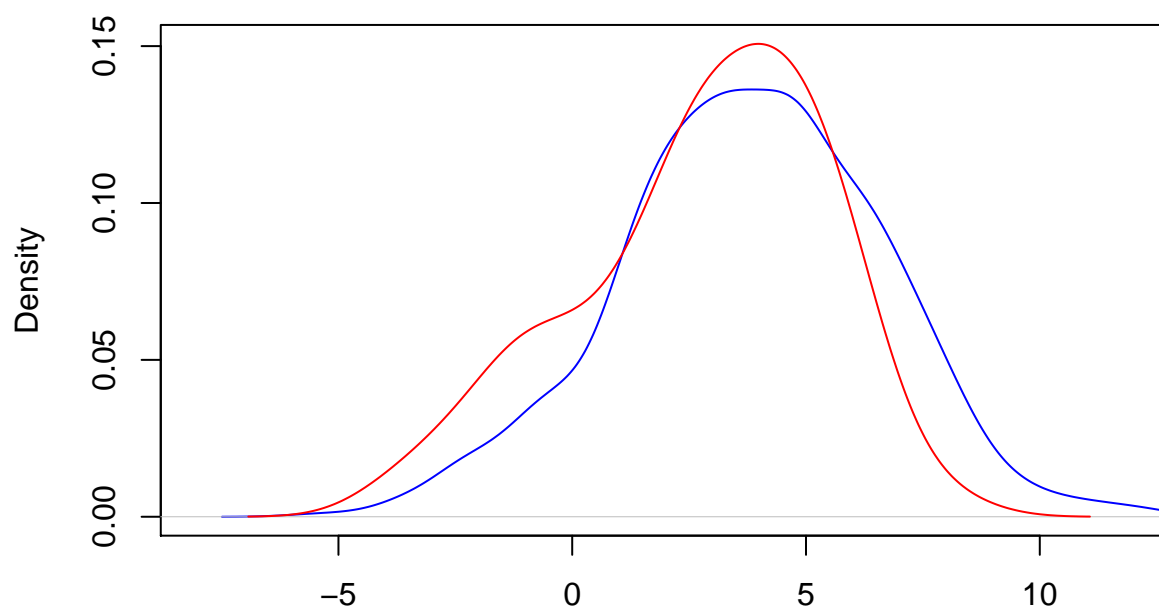


# Monte Carlo Methods HW 5 Output

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## Original density and Predictive Density



N = 1000 Bandwidth = 0.6389

```
#generated the data
set.seed(90)
N = 100
K <- rep(0,N)
for(h in 1:N){
  if(runif(1)<(1/3)){
    K[h] <- rnorm(1, 0, 2)
  }else{
    K[h] <- rnorm(1, 4, 2)
  }
}

#possibly change bandwidth of density

#set aside memory to store into
t = 10000
w <- lam <- mu1 <- mu2 <- n1 <- n2 <- R <- E1 <- E2 <- A12 <- A21 <- rep(0,t)
d1 <- d2 <- F<- dis <- rep(0,N)
M <- E1 <- E2 <- A12 <- A21 <- rep(0, t)
burnin <- 0.1*t
```

```

#set initial values
lam[1] <- 2
mu1[1] <- 2
mu2[1] <- 3
w[1] <- 1
M[1] <- R[1] <- 1
n1[1] <- n2[1] <- 50
d1[1] <- d2[1] <- 0.5
e1 <- e2 <- 1

for(i in 2:t){
  # MH step for finding M's
  currentx = M[i-1]
  proposedx = (M[i-1] %% 2) + 1
  ### fix all this shit!!
  for(q in 2:N){
    e1 <- w[i-1]*dnorm(K[q-1], mu1[i-1], 1/sqrt(lam[i-1])) + (1-w[i-1])*dnorm(K[q-1], mu2[i-1], 1/sqrt(lam[i-1]))
    e2 <- dnorm(K[q-1], mu1[i-1], 1/sqrt(lam[i-1]))*e1
  }

  #A12[i] = (dnorm(rnorm(1,0,100), 0, 100)*e1)/(dnorm(rnorm(1,0,100), mu1[i-1], 100)*e2)
  #A21[i] = (dnorm(mu2[i-1], mu1[i-1], 100)*e2)/(dnorm(mu2[i-1], 0, 100)*e1) ###make sure A12 m
  A12[i] = 1
  A21[i] = 1

  if(currentx < proposedx){

    if(runif(1)<A12[i]){
      M[i] = proposedx # accept move with probabily min(1,A)

      for(q in 1:N){
        norm <- (w[1]*dnorm(K[q], mu1[1], sqrt(1/lam[1]))) + ((1-w[1])*dnorm(K[q], mu2[1], sqrt(1/lam[1])))
        d1[q] <- (w[1]*dnorm(K[q], mu1[1], sqrt(1/lam[1])))/norm
        d2[q] <- ((1-w[1])*dnorm(K[q], mu2[1], sqrt(1/lam[1])))/norm
        dis[q] <- ifelse(d1[q] >= d2[q], 1, 0)
      }

      n1[i] <- sum(dis)
      n2[i] <- N - n1[i]

      w[i] <- rbeta(1, n1[i] +0.5, n2[i] +0.5)

      DX <- cbind(dis, K)

      DX1 <- DX[DX[,1] == 1,]
      mu1[i] <- rnorm(1, ((lam[i-1]/2)*sum(DX1[,2]))/(0.01 + n1[i]*lam[i-1]/2), (1/(0.01 + n1[i]*lam[i-1]/2)))

      DX2 <- DX[DX[,1] == 0,]
      mu2[i] <- rnorm(1, ((lam[i-1]/2)*sum(DX2[,2]))/(0.01 + n2[i]*lam[i-1]/2), (1/(0.01 + n2[i]*lam[i-1]/2)))

      lam[i] <- rgamma(1, 0.5 +N/2, 0.5 + 0.5*(sum(DX1[,2] - mu1[i])^2 + sum(DX2[,2] - mu2[i])^2))
    }
  }
}

```

```

    if(runif(1)<w[i]){
      R[i] = rnorm(1, mu1[i], sqrt(1/lam[i]))
    } else {
      R[i] = rnorm(1, mu2[i], sqrt(1/lam[i]))
    }

  }else{
    M[i] = currentx # otherwise "reject" move, and stay where we are

    mu1[i] <- rnorm(1, ((lam[i-1]/2)*sum(K))/(0.01 + N*lam[i-1]/2), (1/(0.01 + N*lam[i-1]/2)))
    lam[i] <- rgamma(1, 0.5+(N/2), rate = (0.5 + 0.5* sum((K-mu1[i-1])^2)))
    R[i] <- rnorm(1, mu1[i], sqrt(lam[i]))
  }
}else{

  if(runif(1) < A21[i]){
    M[i] = proposedx # accept move with probabily min(1,A)

    mu1[i] <- rnorm(1, ((lam[i-1]/2)*sum(K))/(0.01 + N*lam[i-1]/2), (1/(0.01 + N*lam[i-1]/2)))
    lam[i] <- rgamma(1, 0.5+(N/2), rate = (0.5 + 0.5* sum((K-mu1[i-1])^2)))
    R[i] <- rnorm(1, mu1[i], sqrt(lam[i]))

  }else{
    M[i] = currentx # otherwise "reject" move, and stay where we are

    for(q in 1:N){
      norm <- (w[i-1]*dnorm(K[q], mu1[i-1], sqrt(1/lam[i-1]))) + ((1-w[i-1])*dnorm(K[q], mu2[i-1], sqrt(1/lam[i-1])))
      d1[q] <- (w[i-1]*dnorm(K[q], mu1[i-1], sqrt(1/lam[i-1])))/norm
      d2[q] <- ((1-w[i-1])*dnorm(K[q], mu2[i-1], sqrt(1/lam[i-1])))/norm
      dis[q] <- ifelse(d1[q] >= d2[q], 1, 0)
    }

    n1[i] <- sum(dis)
    n2[i] <- N - n1[i]

    w[i] <- rbeta(1, n1[i] +0.5, n2[i] +0.5)

    DX <- cbind(dis, K)

    DX1 <- DX[DX[,1] == 1,]
    mu1[i] <- rnorm(1, ((lam[i-1]/2)*sum(DX1[,2]))/(0.01 + n1[i]*lam[i-1]/2), (1/(0.01 + n1[i]*lam[i-1]/2)))

    DX2 <- DX[DX[,1] == 0,]
    mu2[i] <- rnorm(1, ((lam[i-1]/2)*sum(DX2[,2]))/(0.01 + n2[i]*lam[i-1]/2), (1/(0.01 + n2[i]*lam[i-1]/2)))

    lam[i] <- rgamma(1, 0.5 +N/2, 0.5 + 0.5*(sum(DX1[,2] - mu1[i])^2 + sum(DX2[,2] - mu2[i])^2))

    if(runif(1)<w[i]){
      R[i] = rnorm(1, mu1[i], sqrt(1/lam[i]))
    } else {
      R[i] = rnorm(1, mu2[i], sqrt(1/lam[i]))
    }
  }
}

```

```
    }  
  }  
}
```