# Hw 2/23

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### **Basic Function**

#### Genderate X

```
We are getting a X every time
getX <- function(mean, sigma)
{
   return(rmultinom(1, mean, sigma))</pre>
```

#### Calculate distribution of theta

Since theta is following Dirichlet Distribution, the marginal distribution of theta i is following a beta distribution.

```
#calcConstant <- function(alpha){
# return(exp(sum(lgamma(alpha)) - lgamma(sum(alpha))))
#}

calMarginalAlpha <- function(X, i){
    X.dim = length(X[,1])
    alpha <- rowSums(X)
    #balpha = calcConstant(alpha)
    return(c(alpha[i], sum(alpha) - alpha[i]))
}

calcDensity <- function( alpha )
{
    x <- seq(0,1,length = 200)
    return(dbeta(x, alpha[1], alpha[2]))
}</pre>
```

#### Plot function

```
plotCurrent <- function( data, n )
{
   par(mfrow = c(1,3))
   thetaValue <- seq(0,1, length = 200)
   length(data[1])
   plot(x = thetaValue, y = data[,1], type='l', main= paste("Theta 1 in run ", n ))
   plot(x = thetaValue, y = data[,2], type='l', main= paste("Theta 2 in run ", n ))
   plot(x = thetaValue, y = data[,3], type='l', main= paste("Theta 3 in run ", n ))
}</pre>
```

## other supporting function

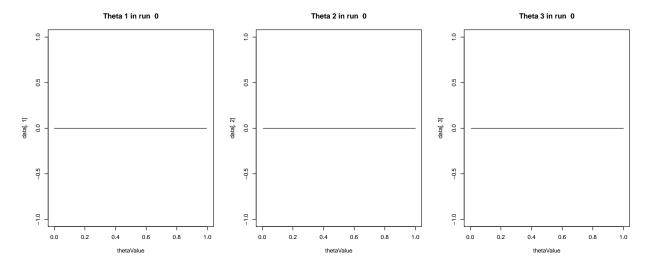
```
calcMean <- function( DensityData )</pre>
{
  thetaValue <- seq(0,1, length = 200)
  return(sum(DensityData * thetaValue / sum(DensityData)))
}
calcVar <- function( DensityData )</pre>
  mean <- calcMean(DensityData)</pre>
  thetaValue <- seq(0,1, length = 200)
  return(sum(DensityData*(thetaValue - mean)^2))
calcDensityAndPlot <- function(Data, i)</pre>
    alpha_1 <- calMarginalAlpha(Data, 1)</pre>
    alpha 2 <- calMarginalAlpha(Data, 2)</pre>
    alpha_3 <- calMarginalAlpha(Data, 3)</pre>
    DensityData_1 <- calcDensity(alpha_1)</pre>
    DensityData_2 <- calcDensity(alpha_2)</pre>
    DensityData_3 <- calcDensity(alpha_3)</pre>
    plotCurrent(cbind(DensityData_1,DensityData_2,DensityData_3), i)
}
calcDensityAndPrintMeanSn <- function(Data)</pre>
    alpha_1 <- calMarginalAlpha(Data, 1)</pre>
    alpha_2 <- calMarginalAlpha(Data, 2)</pre>
    alpha_3 <- calMarginalAlpha(Data, 3)</pre>
    DensityData_1 <- calcDensity(alpha_1)</pre>
    DensityData_2 <- calcDensity(alpha_2)</pre>
    DensityData_3 <- calcDensity(alpha_3)</pre>
      print(calcMean(DensityData_1))
      print(calcVar(DensityData_1)/10000)
      print(calcMean(DensityData_2))
      print(calcVar(DensityData_2)/10000)
      print(calcMean(DensityData_3))
      print(calcVar(DensityData_3)/10000)
```

# **Start Simulation**

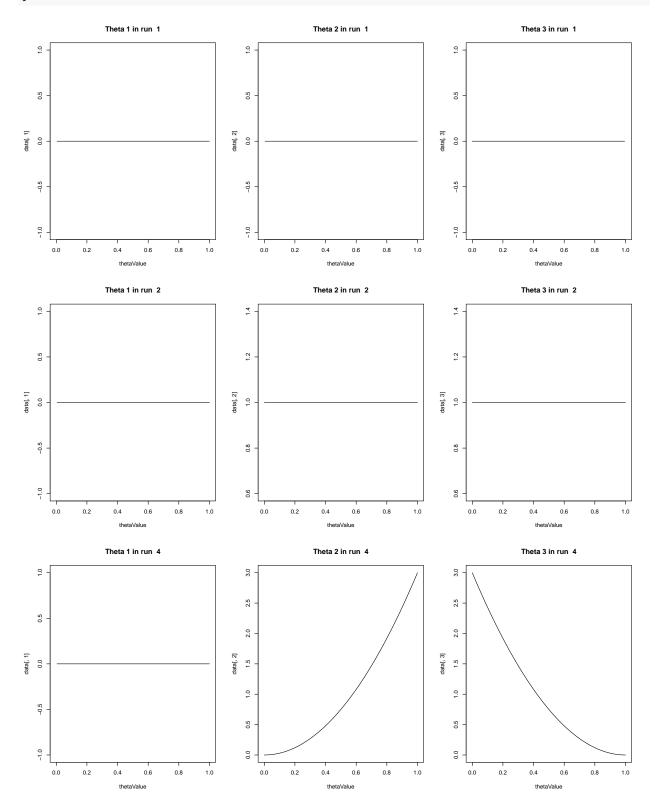
# Start with theta $\sim Dir(1)$

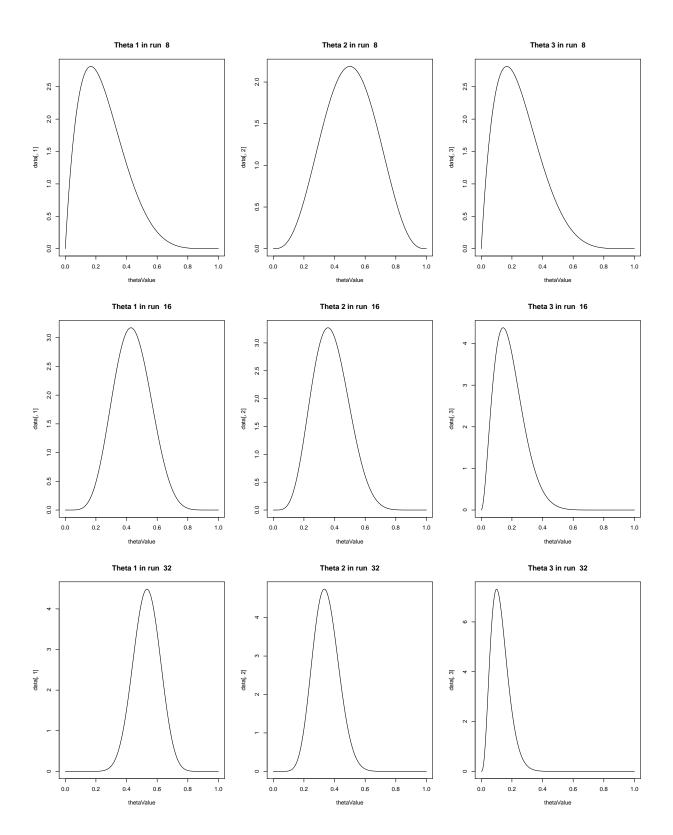
```
D <- numeric(length=3)
D <- matrix(D, nrow = 3, ncol = 1)
D[1,1] = 1

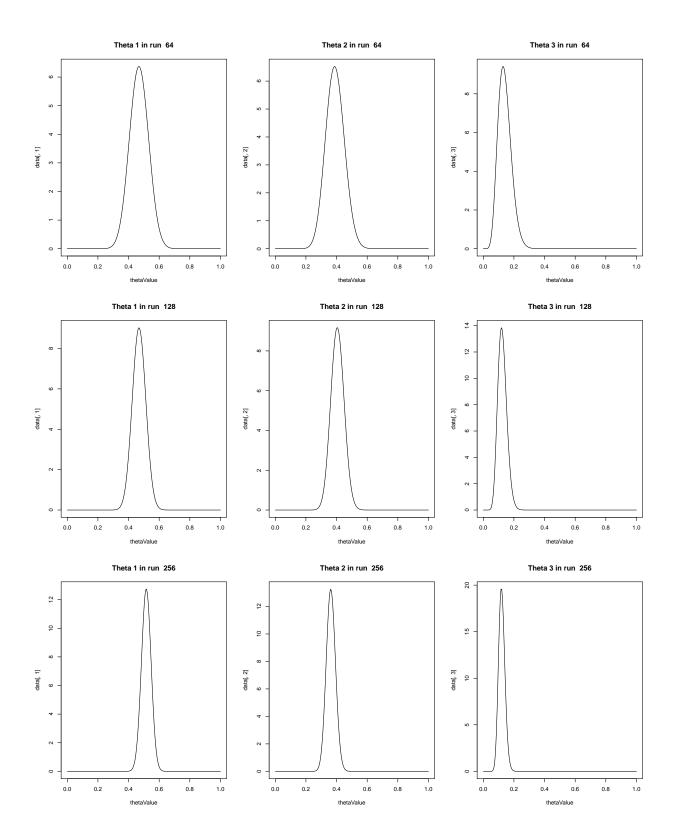
calcDensityAndPlot(D, 0)</pre>
```

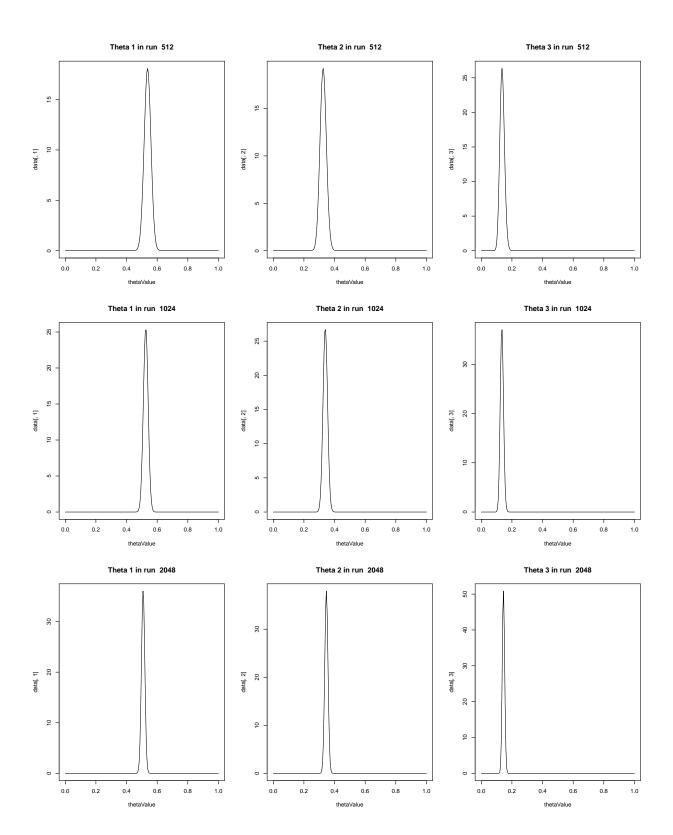


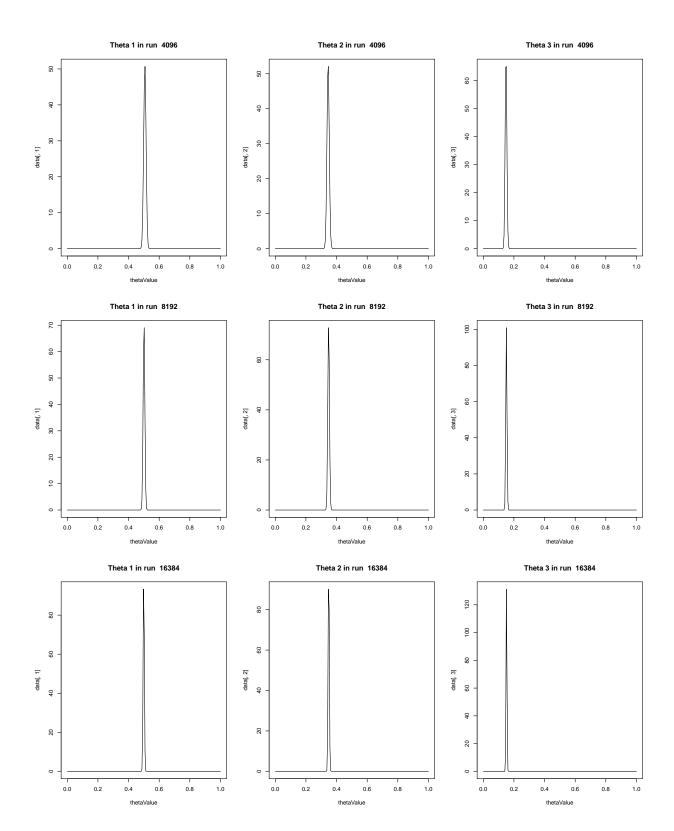
# Sample X1 and plot the marginal distribution

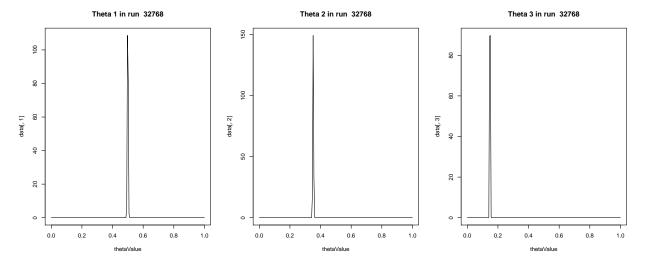




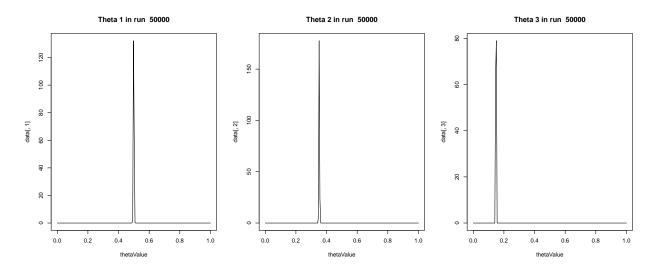








### calcDensityAndPlot(Data, 50000)



### Calculate Mean and Sn

### ${\tt calcDensityAndPrintMeanSn(Data)}$

```
## [1] 0.4990075
```

## [1] 1.138926e-07

## [1] 0.3522086

## [1] 6.894541e-08

## [1] 0.1484818

## [1] 9.022986e-08

# Conclusion

By simulation, we find the marginal posterior distribution of thetas.

It is a normal distribution, which is following Central Limit Theorem.