

## STAT 406 Lab 4, 10/06/2015

### 1. Generate Random Variable

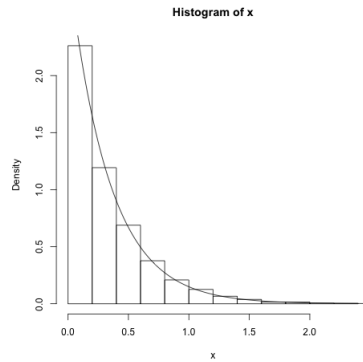
We learned from lecture that random variables from a particular distribution can be generated from uniform random variables by inverting their cumulative distribution function (cdf). That is, if you create samples  $U_1, \dots, U_n$  from  $Uniform(0, 1)$  distribution and compute  $F^{-1}(U_1), \dots, F^{-1}(U_n)$ , then what you get is samples from a distribution with cdf  $F$ . Using this property, we'll generate exponential random variables and gamma random variables.

(a) The  $Exponential(\lambda)$  distribution has cdf:

$$F(x) = 1 - e^{-\lambda x}, \lambda > 0$$

Using `runif` function, generate 100 samples from  $Exponential(3)$  distribution using the inversion method. Graph the density histogram for the sample with the true density superimposed for comparison.

```
n = 100
lambda = 3
x = -(1/lambda)*log(runif(n))
hist(x, prob = TRUE)
y = seq(0,10,length = 1000)
lines(y,dexp(y,3))
```



(b)  $Gamma(k, \theta)$ ,  $k > 0$ ,  $\theta > 0$  distribution has density:

$$f(x) = \frac{\theta^k}{(k-1)!} x^{k-1} e^{-\theta x}$$

where  $k$  and  $\theta$  are shape and rate parameters. What are  $k$  and  $\theta$  for  $Exponential(\lambda)$  distribution in terms of Gamma distribution?

Gamma distribution also has the property that if  $X \sim Gamma(a, \theta)$  and  $Y \sim Gamma(b, \theta)$ , and  $X$  and  $Y$  are independent, then

$$X + Y \sim Gamma(a + b, \theta).$$

Using this property, create 10 samples from  $Gamma(5, 3)$  distribution.

```
n = 10
k = 5
lambda = 3
x = matrix(-(1/lambda)*log(runif(n*k)), ncol=k)
g = apply(x, 1, sum)
```

(c) Take an  $n$ -sample from the  $Gamma(5, 3)$  using your algorithm from part (b), with  $n = 10, 20, 50, 100, 200, 1000$ .

```

n = c(10, 20, 50, 100, 200, 1000)
g = list()
k = 5
lambda = 3
g = lapply(n, function(i) matrix(-(1/lambda)*log(runif(i*k)), ncol=k))
class(g) #to see what is the data structure used for g
sapply(g, dim) #check the dimension of each matrix in g
## the following will get the gamma random numbers of different sizes n. ##
gamma = lapply(g, function(M) matrix(-(1/lambda)*log(runif(i*k)), ncol=k))
class(gamma) #to see what is the data structure
sapply(gamma, length) #check the dimension of each vector in g

```

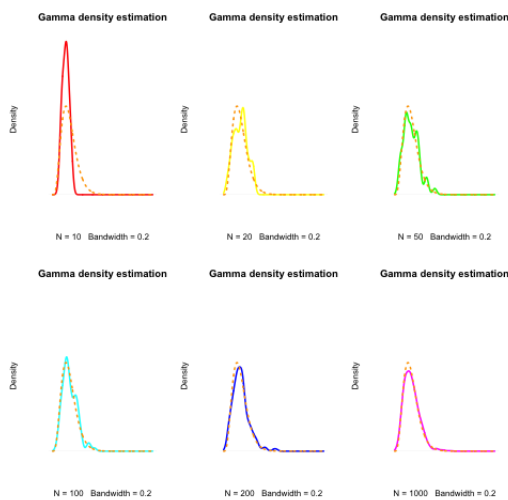
(d) Now, use R function **density** to obtain an estimate of the underlying density. Show how the estimate gets better as n increases, and graph the estimated densities and the true density superimposed for comparison.

```

#continue with gamma from the output of part (c)
grid = 1000
y = seq(0,10,length = grid)
plot(0,xlim=c(0,10),ylim=c(0,1),xlab="x",ylab="Density",main="Gamma density estimation", type='n')
par(mfrow=c(2,3))
for(i in 1:6){
  di = density(gamma[[i]],from=0,to=10, n = grid, bw=0.2)
  plot(di, xlim=c(0,10), ylim=c(0,1), axes=F, col=rainbow(6)[i], lwd = 2, main="Gamma density estimation")
  lines(y,dgamma(y,5,3), col = 'orange', lty = 3, lwd = 2)
}

names(di) #check to see what are in the density estimator output di, notice di$x and di$y
##multiple plot in R : http://cran.r-project.org/doc/contrib/Lemon-kickstart/kr\_addat.html

```



## 2. More of LLN - Empirical Distribution Functions

Write a function which calculates the empirical distribution function of  $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} N(0, 1)$ . The empirical distribution function is defined as

$$\Phi_m(x) = \frac{1}{n} \sum_{i=1}^n I(X_i \leq x)$$

where  $I(X \leq x)$  is an indicator function, taking value 1 if  $X \leq x$  and 0 otherwise. Plot this function for  $x \in [-4, 4]$ . Let  $m = 5, 50, 500$  respectively and plot everything on the same figure, with the true Normal cdf superimposed for comparison.

```

Phi.m = function(m,xseq){
  # xseq can be either a scalar or a vector
  X = rnorm(m)
  Phi.m.x = sapply(xseq,function(x){mean(X<x)})
  return(Phi.m.x)
}

m = c(5,50,500)
xseq = seq(-4,4,by=0.01)
plot(0,xlim=c(-4,4),ylim=c(0,1),xlab="x",ylab="Phi.m(x)",main="Empirical CDF", type='n')
for (i in 1:length(m)){
  par(new=T)
  plot(xseq,Phi.m(m[i],xseq),col=i,axes=F,type="l",xlim=c(-4,4),ylim=c(0,1),xlab="",ylab="")
}
lines(xseq,pnorm(xseq,mean=0,sd=1),col="blue",lty=3,lwd=2)
legend("topleft",legend=paste(m,'points'),col=1:length(m),lty=1)

```

