

# Stat 361 - Recitation 5

## Generating Random Variates

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### Question 1:

Generate a random sample of size  $n = 10^5$  from Beta(3,9). Try to use only *runif(1)* function to generate random variables.

Density function of Beta( $\alpha, \beta$ ) is

$$f_X(x : \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} \quad x \in (0, 1); \alpha, \beta > 0$$

If both  $\alpha$  and  $\beta$  are integers, we can generate random numbers from the beta distribution with parameter  $\alpha$  and  $\beta$  by using the following algorithm:

- Generate  $\alpha + \beta - 1$  Uniform Random Numbers;  $U_1, U_2, \dots, U_{\alpha+\beta-1}$ .
- Deliver  $X = U_{(\alpha)}$ , which is the  $\alpha^{th}$  order statistic.

Also, show that your sample has the properties of Beta(3,9).

$$E(X) = \frac{\alpha}{\alpha + \beta}, \quad Var(X) = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$$

```

n <- 10^5
alpha <- 3
beta <- 9
X <- numeric(n)

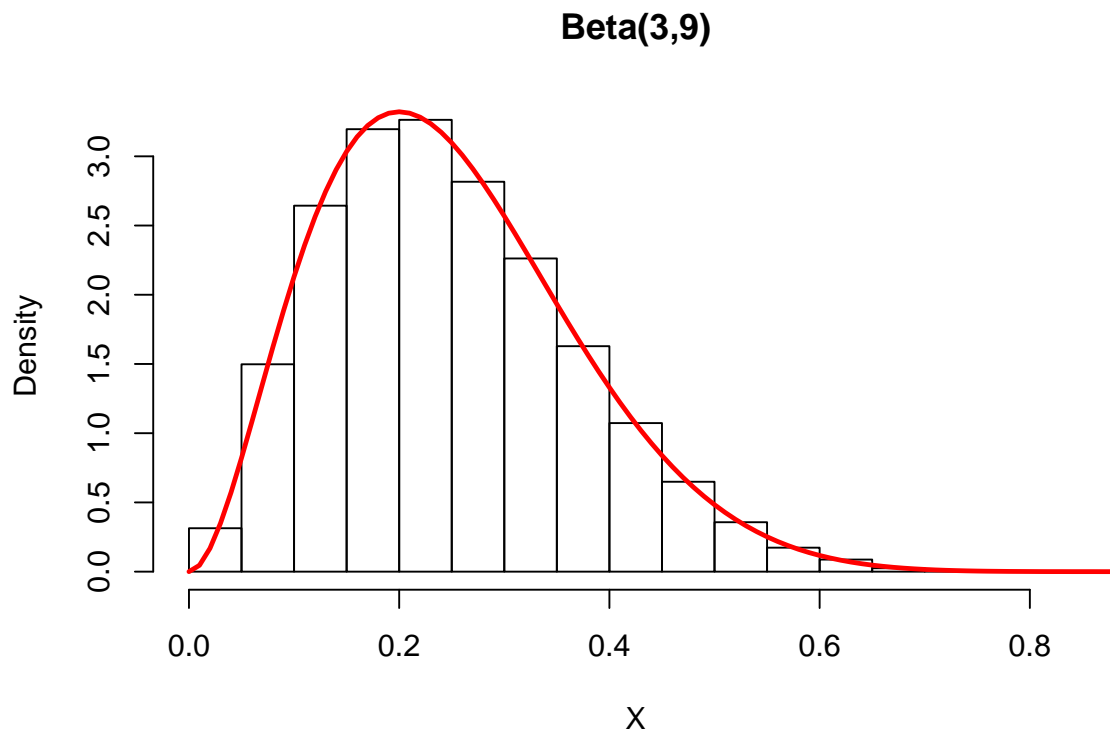
for(i in 1:n){

  u <- runif(alpha + beta - 1)
  sortu <- sort(u)
  X[i] <- sortu[alpha]

}

hist(X, prob = TRUE, main = "Beta(3,9)")
y=seq(0,1,0.01)
lines(y,dbeta(y,alpha,beta),col="red", lwd = 2.5)

```



```

theoretical_mean <- alpha / (alpha + beta)
theoretical_var <- (alpha*beta) / ((alpha+beta)^2*(alpha+beta+1))
est_mean <- mean(X)
est_var <- var(X)

```

```
control <- matrix(round(c(theoretical_mean,theoretical_var,
                        est_mean,est_var),5), nrow = 2, byrow = T)

colnames(control) <- c("Mean","Variance")
rownames(control) <- c("Theoritical", "Estimated")

control

##              Mean Variance
## Theoritical 0.25000 0.01442
## Estimated   0.24988 0.01449
```

## Question 2:

Generate a random sample of size  $n = 10^5$  from  $\text{Gamma}(5,8)$  by using relation between Exponential and Gamma distributions. Try to use only *runif(1)* function to generate random variables.

(*Hint:* Summation of  $t$  random variables from  $\text{Exponential}(\lambda)$  is distributed as  $\text{Gamma}(t,\lambda)$ )

Also, show that your sample has the properties of  $\text{Gamma}(5,8)$ .

$$E(X) = \frac{t}{\lambda}, \quad \text{Var}(X) = \frac{t}{\lambda^2}$$

## Generating Random Variables from Gamma Distribution.

Let  $X$  is a random variable from  $\text{Gamma Distribution}(t,\lambda)$  and its PDF is

$$f(x; \alpha, \beta) = \frac{1}{\Gamma(t)\lambda^t} x^{t-1} e^{-\frac{x}{\lambda}} \quad x > 0; \quad t, \lambda > 0$$

Additionally, CDF of  $X$  is given as

$$F_x(x) = \frac{1}{\Gamma(t)\lambda^t} \int_0^x x^{t-1} e^{-\frac{x}{\lambda}} dx \quad x > 0$$

The inverse transform method cannot be applied in this case since there is not a closed form of solution for its inverse.

We know that the sum of  $t$  independent  $\text{Exponential}(\lambda)$  is distributed as  $\text{Gamma}(t,\lambda)$ . This leads the following transformation based on  $t$  uniform random numbers.

**Remember** from Recitation 3 we had shown that the inverse transform of Exponential distribution is

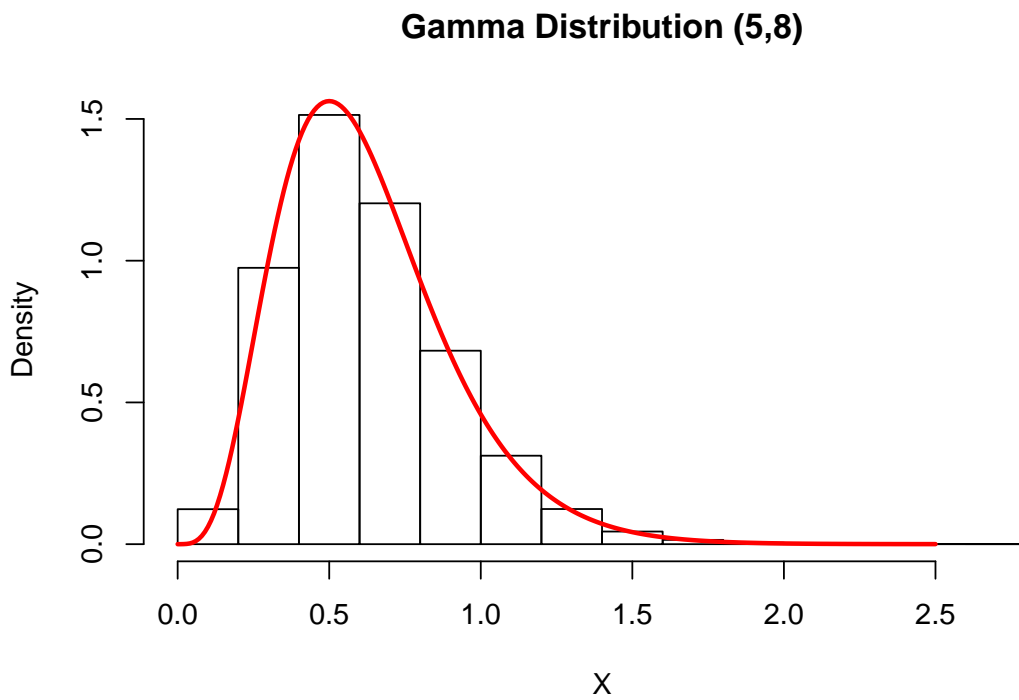
$$x = -\frac{1}{\lambda} \log(U)$$

Then, to generate a random sample from Gamma Distribution we can use inverse transform method for exponential distribution;

$$X = -\frac{1}{\lambda} \log(U_1) - \dots - \frac{1}{\lambda} \log(U_t)$$

```
set.seed(361)
n <- 10^5
t <- 5
lambda <- 8
U <- matrix(runif(n*t), nrow = t, ncol = n)
logU <- -log(U) / lambda # inverse transform method
X <- apply(logU, 2, sum) # col sums of matrix logU

hist(X, prob=TRUE, main = "Gamma Distribution (5,8)")
y = seq(0, 2.5, 0.01)
lines(y, dgamma(y, t, lambda), col="red", lwd = 2.5)
```



```

theoretical_mean <- t/lambda
theoretical_var  <- t/(lambda^2)
est_mean <- mean(X)
est_var <- var(X)

control <- matrix(round(c(theoretical_mean,theoretical_var,
                          est_mean,est_var),5), nrow = 2, byrow = T)

colnames(control) <- c("Mean","Variance")
rownames(control) <- c("Theoritical", "Estimated")

control

##              Mean Variance
## Theoritical 0.62500 0.07812
## Estimated   0.62576 0.07857

```

### Question 3:

Generate random sample of size  $10^5$  from  $\chi^2$  distribution with 7 degrees of freedom by using the relation between Chi-Square and standard normal distribution. Check the histogram and compare estimated mean and variance with the theoritical expectation and variance of  $\chi^2(v)$  distribution ( $v=7$ ).

$$E(X) = v \quad Var(X) = 2v$$

We can use the fact that the chi-square distribution with  $V$  degrees of freedom is the sum of  $V$  squared independent standard normal;

$$X = \sum_{i=1}^v Z_i^2$$

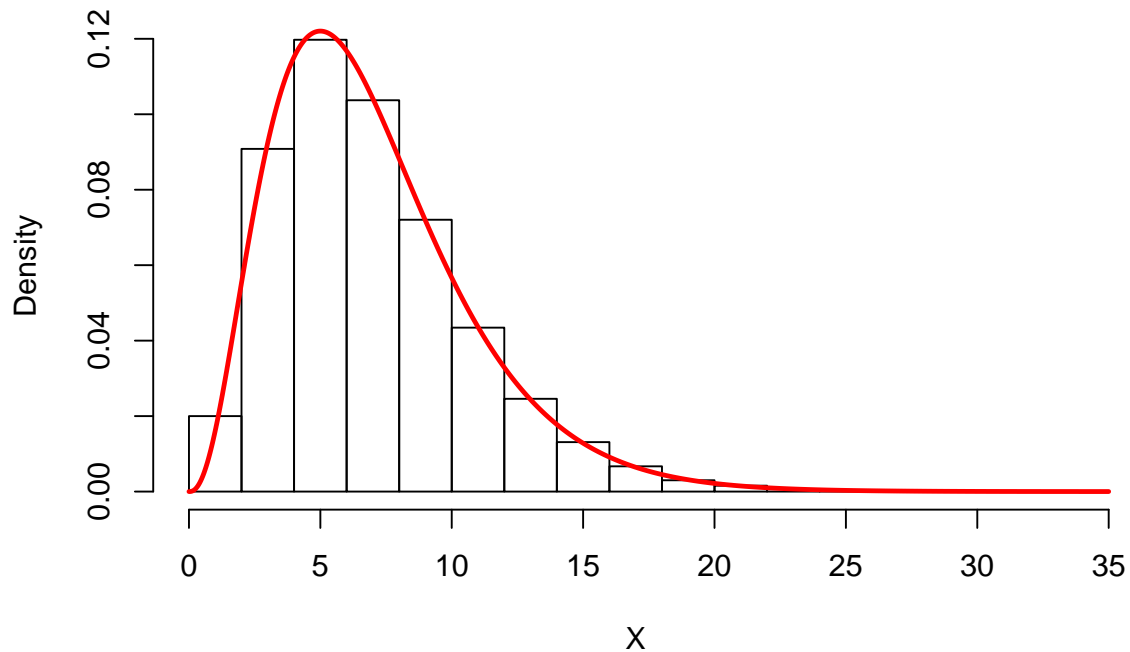
```

set.seed(361)
n <- 10^5
v <- 7
Z <- matrix(rnorm(v*n,0,1),nrow = v, ncol = n)
SquaredZ <- Z^2
X <- colSums(SquaredZ)

hist(X, prob=TRUE, main = "Chi-Square Distribution")
y = seq(0,35,0.01)
lines(y,dchisq(y,v),col="red",lwd = 2.5)

```

## Chi-Square Distribution



```
theoretical_mean <- v
theoretical_var  <- 2*v
est_mean <- mean(X)
est_var <- var(X)

control <- matrix(round(c(theoretical_mean,theoretical_var,
                          est_mean,est_var),5), nrow = 2, byrow = T)

colnames(control) <- c("Mean","Variance")
rownames(control) <- c("Theoritical", "Estimated")

control
```

```
##           Mean Variance
## Theoritical 7.00000 14.0000
## Estimated   6.98805 14.0141
```

### Question 4:

Generate random sample of size  $10^5$  from Student-T distribution with 40 degrees of freedom by using the following transformation method. Compare the histogram with the Student-T density curve.

Hint: If  $Z \sim N(0, 1)$  and  $V \sim \chi^2(n)$  are independent, then  $\frac{Z}{\sqrt{V/n}}$  has the Student-T distribution with  $n$  degrees of freedom. Generate  $\chi^2$  random variables from standard normal distribution.

```
set.seed(361)
n <- 10^5
v <- 40

Z1 <- rnorm(n)
Z2 <- matrix(rnorm(n*v), n, v)
Z2_sqr <- Z2^2
V <- rowSums(Z2_sqr)

t <- Z1 / (sqrt(V / v))
length(t)

## [1] 100000

hist(t, prob = TRUE, main = "Student-T Distribution")
y <- seq(-4, 4, 0.01)
lines(y, dt(y, v), col = "Red", lwd = 2.5)
```

