

# Statistical Computing HW 4

*Mario Ibanez*

*January 30, 2016*

## Problem 4.3)

Give an efficient algorithm to simulate the value of a random variable  $X$  such that

$$P\{X = 1\} = 0.3$$

$$P\{X = 2\} = 0.2$$

$$P\{X = 3\} = 0.35$$

$$P\{X = 4\} = 0.15$$

## Answer

### Derivation

The CDF of the distribution is:

$$F(x) = \begin{cases} 0 & x < 1 \\ 0.3 & 1 \leq x < 2 \\ 0.5 & 2 \leq x < 3 \\ 0.85 & 3 \leq x < 4 \\ 1 & 4 \leq x \end{cases}$$

The can be used along with a standard uniform random variable to generate values of the random variable  $X$ .

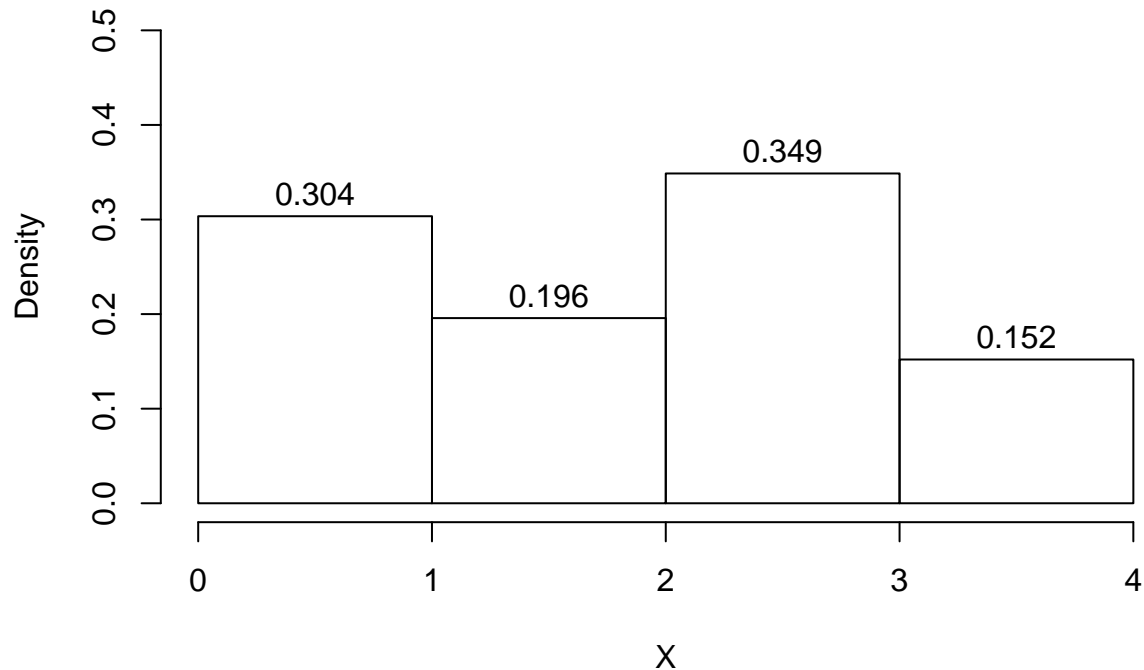
### Algorithm

1. Generate a number  $u$  from the standard uniform distribution
2. Set  $x = F^{-1}(u) + 1$
3. Repeat  $n$  times

### Program

Below is a histogram using the derivation and algorithm. 10,000 values were generated. The distribution is quite good.

### Distribution of 10,000 generated X values



#### Problem 4.14a)

Let  $X$  be a binomial random variable with parameters  $n$  and  $p$ . Suppose that we want to generate a random variable  $Y$  whose probability mass function is the same as the conditional mass function of  $X$  given that  $X \geq k$ , for some  $k \leq n$ . Let  $\alpha = P\{X \geq k\}$  and suppose that the value of  $\alpha$  has been computed. Give the inverse transform method for generating  $Y$ .

**Answer**

**Derivation**

**Algorithm**

**Program**

**Analytical Result**

#### Problem 4.15)

Give a method for simulating  $X$ , having the probability mass function  $p_j$ ,  $j = 5, 6, 7, \dots, 14$ , where

$$p_j = \begin{cases} 0.11 & \text{when } j \text{ is odd and } 5 \leq j \leq 13 \\ 0.09 & \text{when } j \text{ is even and } 6 \leq j \leq 14 \end{cases}$$

Use the text's random number sequence to generate  $X$ .

## Answer

## Derivation

## Algorithm

## Program

## Analytical Result

## Problem 5.10)

A casualty insurance company has 1000 policyholders, each of whom will independently present a claim in the next month with probability 0.05. Assuming that the amounts of the claims made are independent exponential random variables with mean \$800, use simulation to estimate the probability that the sum of these claims exceeds \$50,000.

## Answer

## Derivation

In order to generate random values from an exponential distribution with mean 800, we'll use the fact that the CDF of this distribution is

$$F(x) = 1 - e^{-x/800}$$

Given that  $F(x)$  has a standard uniform distribution, then:

$$\begin{aligned}u &= 1 - e^{-x/800} \\e^{-x/800} &= 1 - u \\ \frac{-x}{800} &= \ln(1 - u) \\ x &= (-800)\ln(u)\end{aligned}$$

Note that the random variables  $U$  and  $1 - U$  have the same distribution if  $U$  is standard uniform.

Also, since each policyholder has a 5% chance of making a claim, then the number of claims made out of 1000 has a binomial distribution with  $n = 1000$  and  $p = 0.05$ . The random variable we are interested in is then

$$S = X_1 + X_2 + X_3 + \dots + X_N$$

where the  $X_i$  are *iid* exponential with mean 800 and  $N$  is binomial with  $n = 1000$  and  $p = 0.05$ . For simplicity in the program, values from the binomial distribution will be done using an R function, but the exponential values will be generated using values from a uniform distribution according to the derivation above.

## Algorithm

- For each trial:
  - Generate a value  $n$  from Binomial(1000, 0.05)
  - Generate  $n$  values  $u_1, u_2, \dots, u_n$  from  $U(0,1)$

- Calculate each of the  $n$  claim amounts  $x$  by evaluating  $x_i = (-800)\ln(u_i)$
- Find the sum of the  $n$  claims for that trial
- After the sum of each trial is calculated, then determine how many are greater than \$50,000

The R code appears different than this algorithm, but in practice this is what is happening. The R language benefits from avoiding explicitly writing *for* loops, and instead using other methods like *apply*, *lapply*, and *sapply*.

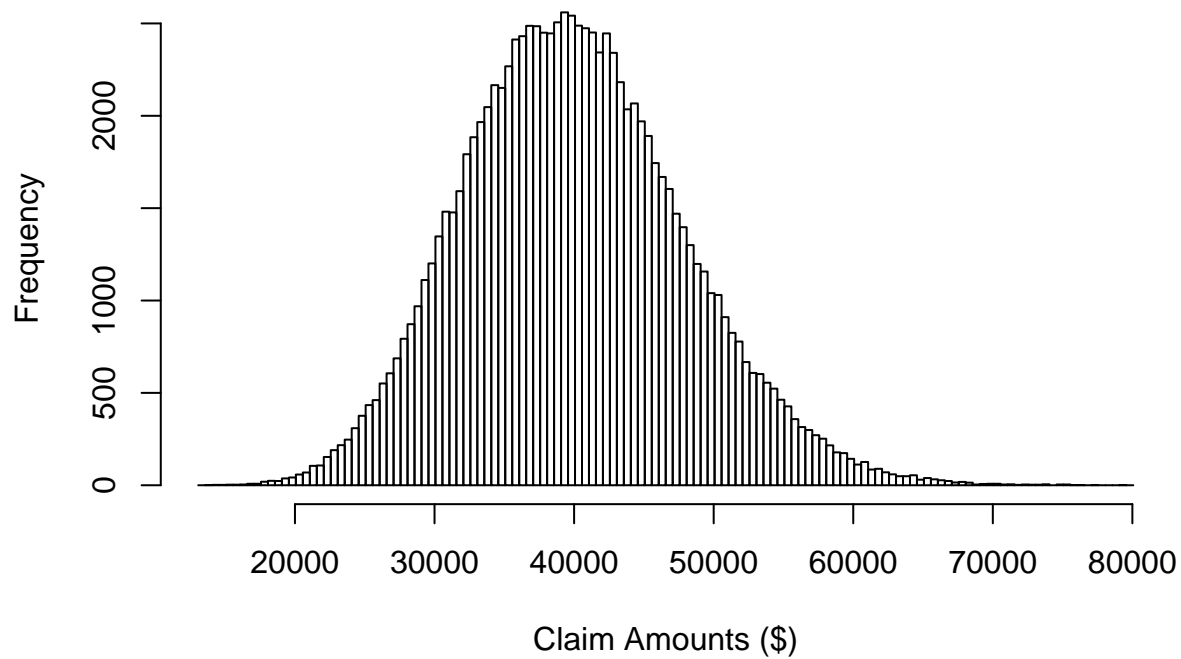
## Program

The program uses the `generate_claims()` function to generate  $n$  claims from an exponential distribution, where  $n$  has a binomial distribution. 100,000 trials are run, and then the sum of the claims from each trial is calculated.

```
# Set seed, set N = number of simulations
set.seed(1234)
N = 10^5
# Returns variable number claims from exponential distribution
generate_claims <- function(vector, holders = 1000, chance = 0.05, beta = 800){
  # Number of claims that will occur is binomial(holders, chance)
  num_of_claims <- rbinom(1, holders, chance)
  # Generate claim amounts from exponential distribution for each claimant
  claims <- -beta * log(runif(num_of_claims))
  # Return vector of claims
  return(claims)
}
# Initialize claims list
claims <- rep(list(rep(0,100)), N)
# Simulate N times
claims <- lapply(claims, generate_claims)
# Find the sum for each trial
distribution <- sapply(claims, sum)
```

Below is a histogram of the results. The distribution is a sum of exponential random variables and thus is a continuous distribution. It appears to be slightly right tailed with a mean around \$40,000.

## Distribution of Claim Amounts



The following R command returns the estimated probability that the sum of the claims is above \$50,000:

```
mean(distribution>50000)
```

```
## [1] 0.10661
```

Out of 100,000 trials, approximately 0.10661 trials resulted in an outcome where the total value of the claims exceeded \$50,000.

### Analytical Result

*find out how to do this analytically*