



# STATISTICS FOR DATA SCIENCE

## Generation of Random Variates

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Computer Science and Engineering

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## Generation of Random Variates

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## Topics to be covered...

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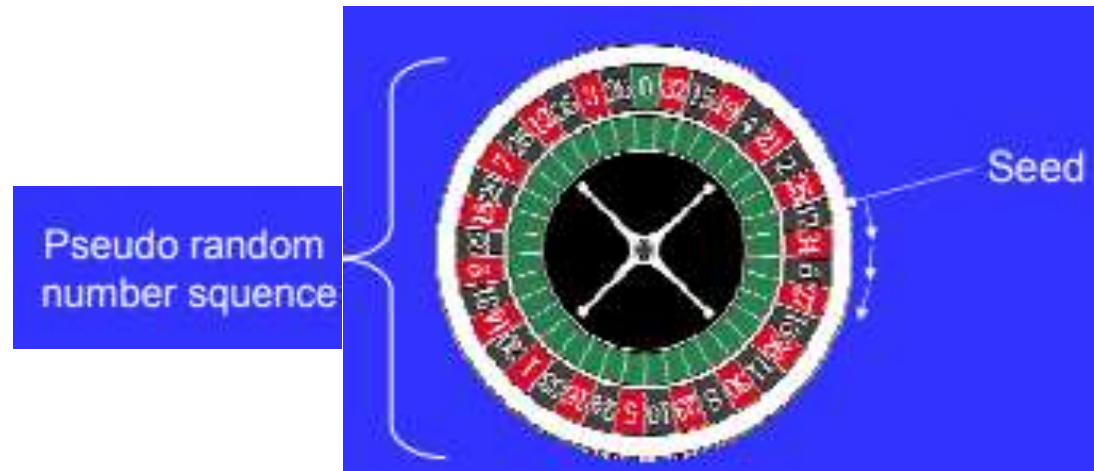
- **Random Numbers**
- **Random Variate Generator**
- **Random Variates**
- **Techniques for Generating Random Variates**



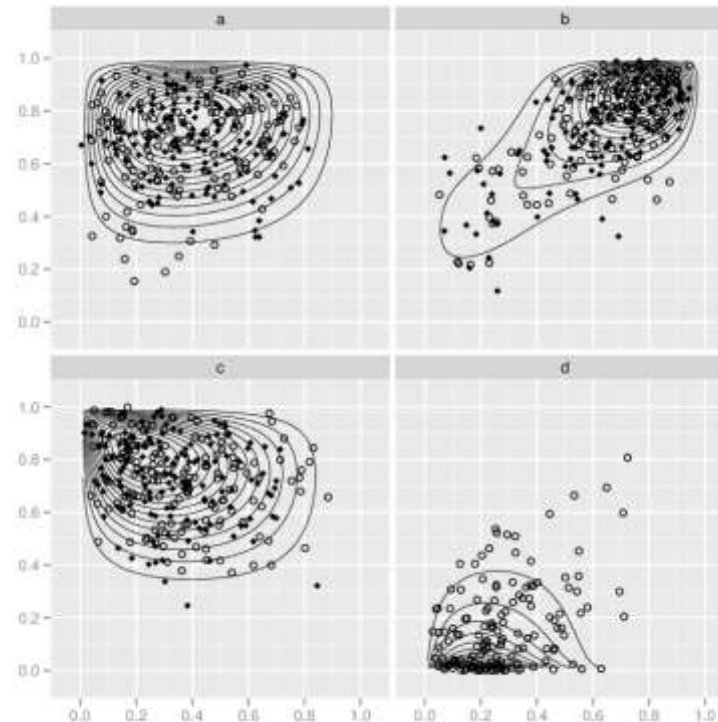
- Random numbers are very important for a simulation.
- Random number generator output: Sequence of independent and identically (uniformly) distributed random numbers between 0 and 1.
- These random numbers are transformed into required probability distributions.



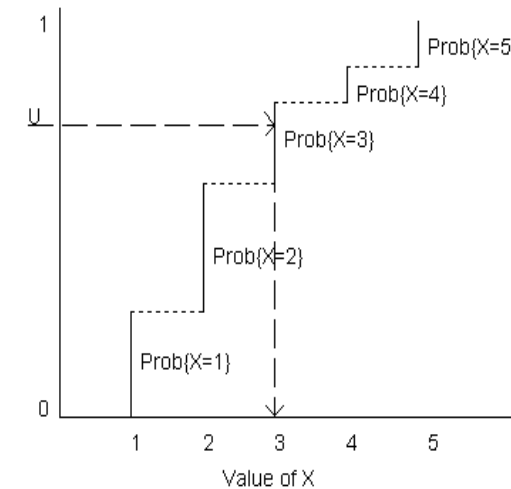
- Computer-based generators use random number seeds for setting the starting point of the random number sequence.
- These seeds are often initialized using a computer's real time clock in order to have some external noise.



It is assumed that a distribution is completely specified and we wish to generate samples from this distribution as input to a simulation model.



- Process of producing observations that have the distribution of the given random variables.
- This is to develop simulation models for the purpose of analysis and decision making.
- It rely on generating uniformly distributed random number on the interval (0,1).
- Random variate generators use as starting point random numbers distributed  $U[0,1]$ .



- A computational or physical device designed to generate a sequence of numbers that lack any pattern (i.e. appear random).
- Computer-based generators are simple deterministic programs trying to fool the user by producing a deterministic sequence that looks random (pseudo random numbers).
- They should meet some statistical tests for randomness intended to ensure that they do not have any easily discernible patterns.



A **random variate** is a variable generated from uniformly distributed pseudorandom numbers.

Depending on how they are generated, a **random variate** can be uniformly or non-uniformly distributed.

**Random variates** are frequently used as the input to simulation models.

**Examples:** Inter-arrival time and service time.

**RV Generators** – Techniques used to generate random variates.

- Inverse transform technique
- Direct transformation for the Normal Distribution
- Convolution Method
- Acceptance and Rejection Technique

1. Compute CDF of the desired random variable X.
2. Set  $F(X)=R$  on the range of X.
3. Solve the equation  $F(X)=R$  for X in terms of R.
4. Generate uniform random numbers  $R_1, R_2, R_3, \dots$  and compute the desired random variate by

$$X_i = F^{-1}(R_i)$$

Inverse transform method – Uniform Distribution Example:

Step 1 – compute *cdf* of the desired random variable  $X$

$$F(x) = \begin{cases} 0, & x < a \\ \frac{x-a}{b-a}, & a \leq x < b \\ 1, & x \geq b \end{cases}$$

Step 2 – Set  $F(X) = R$  where  $R$  is a random number  $\sim U[0,1)$

$$F(x) = R = \frac{x-a}{b-a}$$

Step 3 – Solve  $F(X) = R$  for  $X$  in terms of  $R$ .  $X = F^{-1}(R)$ .

$$R(b-a) = X-a, \quad X = R(b-a) + a$$

Step 4 – Generate random numbers  $R_i$  and compute desired random variates:

$$X_i = R_i(b-a) + a$$

# STATISTICS FOR DATA SCIENCE

## Generation of Bernoulli and Binomial Random Variates

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## Generation of Bernoulli and Binomial Random Variates

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# STATISTICS FOR DATA SCIENCE

## Generation of Poisson Random Variate

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# STATISTICS FOR DATA SCIENCE

## Generation of Normal Random Variate

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## Generation of Normal Random Variate

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Examples of other distributions for which inverse CDF works are:

- Uniform distribution
- Weibull distribution
- Triangular distribution

All discrete distributions can be generated via inverse transform technique.

### **Method:**

Numerically, table-lookup procedure, algebraically, or a formula

### **Examples of application:**

- Empirical
- Discrete uniform
- Geometric

## Inverse-transform Technique: Continuous Distributions

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A number of continuous distributions do not have a closed form expression for their CDF, e.g. Normal, Gamma and Beta.

The presented method does not work for these distributions.

### **Solution**

- Approximate the CDF or numerically integrate the CDF

### **Problem**

- Computationally slow

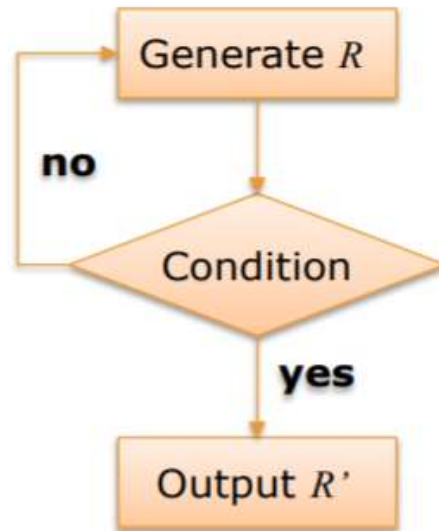
- Useful particularly when inverse CDF does not exist in closed form
- **Illustration:** To generate random variates,  $X \sim U(1/4, 1)$

Procedure:

Step 1. Generate  $R \sim U(0, 1)$

Step 2. If  $R \geq 1/4$ , accept  $X=R$ .

Step 3. If  $R < 1/4$ , reject  $R$ , return to Step 1



$R$  does not have the desired distribution, but  $R$  conditioned ( $R'$ ) on the event  $\{R \geq \frac{1}{4}\}$  does.

- Efficiency: Depends heavily on the ability to minimize the number of rejections.

Procedure of generating a Poisson random variate  $N$  is as follows

1. Set  $n=0, P=1$
2. Generate a random number  $R_{n+1}$ , and replace  $P$  by  $P \times R_{n+1}$
3. If  $P < \exp(-\alpha)$ , then accept  $N=n$ 
  - Otherwise, reject the current  $n$ , increase  $n$  by one, and return to step 2.



- Example: Generate three Poisson variates with mean  $\alpha=0.2$ 
  - $\exp(-0.2) = 0.8187$
- Variate 1
  - Step 1: Set  $n = 0, P = 1$
  - Step 2:  $R1 = 0.4357, P = 1 \times 0.4357$
  - Step 3: Since  $P = 0.4357 < \exp(-0.2)$ , **accept**  $N = 0$
- Variate 2
  - Step 1: Set  $n = 0, P = 1$
  - Step 2:  $R1 = 0.4146, P = 1 \times 0.4146$
  - Step 3: Since  $P = 0.4146 < \exp(-0.2)$ , **accept**  $N = 0$
- Variate 3
  - Step 1: Set  $n = 0, P = 1$
  - Step 2:  $R1 = 0.8353, P = 1 \times 0.8353$
  - Step 3: Since  $P = 0.8353 > \exp(-0.2)$ , reject  $n = 0$  and return to Step 2 with  $n = 1$
  - Step 2:  $R2 = 0.9952, P = 0.8353 \times 0.9952 = 0.8313$
  - Step 3: Since  $P = 0.8313 > \exp(-0.2)$ , reject  $n = 1$  and return to Step 2 with  $n = 2$
  - Step 2:  $R3 = 0.8004, P = 0.8313 \times 0.8004 = 0.6654$
  - Step 3: Since  $P = 0.6654 < \exp(-0.2)$ , **accept**  $N = 2$

## Acceptance and Rejection Technique – Poisson Distribution

- It took five random numbers to generate three Poisson variates
- In long run, the generation of Poisson variates requires some overhead!

| $N$ | $R_{n+1}$ | $P$    | Accept/Reject          |        | Result |
|-----|-----------|--------|------------------------|--------|--------|
| 0   | 0.4357    | 0.4357 | $P < \exp(-\alpha)$    | Accept | $N=0$  |
| 0   | 0.4146    | 0.4146 | $P < \exp(-\alpha)$    | Accept | $N=0$  |
| 0   | 0.8353    | 0.8353 | $P \geq \exp(-\alpha)$ | Reject |        |
| 1   | 0.9952    | 0.8313 | $P \geq \exp(-\alpha)$ | Reject |        |
| 2   | 0.8004    | 0.6654 | $P < \exp(-\alpha)$    | Accept | $N=2$  |

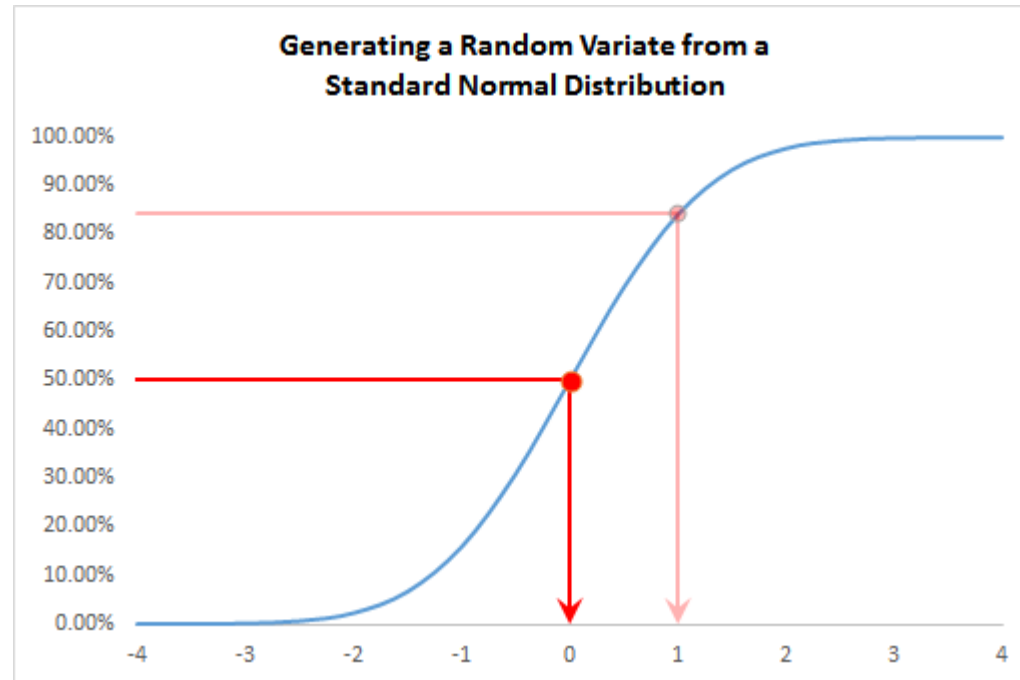
### Approach for $N(0,1)$

- PDF

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

- CDF, No closed form available

$$F(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

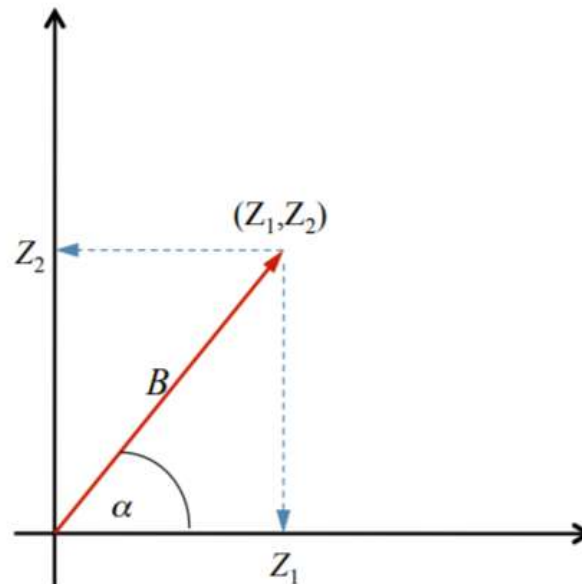


### Approach for $N(0,1)$

- Consider two standard normal random variables,  $Z_1$  and  $Z_2$ , plotted as a point in the plane:

- In polar coordinates:

- $Z_1 = B \cos(\alpha)$
- $Z_2 = B \sin(\alpha)$



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## Direct Transformation

- Approach for  $N(\mu, \sigma^2)$ :
  - Generate  $Z_i \sim N(0,1)$

$$X_i = \mu + \sigma Z_i$$

- Approach for Lognormal( $\mu, \sigma^2$ ):
  - Generate  $X \sim N(\mu, \sigma^2)$

$$Y_i = e^{X_i}$$

Let  $R_1 = 0.1758$  and  $R_2 = 0.1489$

- Two standard normal random variates are generated as follows:

$$Z_1 = \sqrt{-2 \ln(0.1758)} \cos(2\pi 0.1489) = 1.11$$

$$Z_2 = \sqrt{-2 \ln(0.1758)} \sin(2\pi 0.1489) = 1.50$$

- To obtain normal variates  $X_i$  with mean  $\mu=10$  and variance  $\sigma^2 = 4$

$$X_1 = 10 + 2 \cdot 1.11 = 12.22$$

$$X_2 = 10 + 2 \cdot 1.50 = 13.00$$

**Do It Yourself !!!!**

Implement Random Variate Generation for **Poisson Distribution**.

Implement Random Variate Generation for **Normal Distribution**.



**THANK YOU**

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