

# Statistics

FOR

DATA SCIENCE

## UNIT-2

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# random NUMBER GENERATION

Inverse transform technique

Direct transformation for the Normal Distribution

Convolution Method

Acceptance / Rejection Method

- All assume that uniformly distributed numbers in  $[0,1]$  exists

## Inverse transform technique

- cdf is  $F(x)$  where  $x$  is the random variable
- Set  $F(x) = R$  where  $R$  is a uniformly distributed random variable in  $[0,1]$
- Solve  $F(x) = R$  for  $x$  in terms of  $R$  (in range of  $x$ )
- Generate uniform random numbers  $R_1, R_2, R_3 \dots$  and compute  $X_i$  by

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

Q25. Generate RVs for given cdf

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

- Range of  $x$ :  $[a, b]$

$$R = \frac{x-a}{b-a}$$

$$R(b-a) + a = X \quad \text{where } R \in [0, 1]$$

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

### Generation of Bernoulli RVs

$$P(X=1) = p$$

$$P(X=0) = 1-p$$

- A Bernoulli RV can only take up 2 values (0 and 1) and the probability of getting  $X=1$  is  $p$
- Generate  $U$  from  $U(0,1)$
- If  $U \leq p$ ,  $X=1$ ; else  $X=0$

## Generation of Binomial RVs

$$P(X=i) = \binom{n}{i} p^i (1-p)^{n-i}$$

$$P(X=i) = {}^nC_i p^i (1-p)^{n-i}$$

for  $i$  successes in  $n$  independent Bernoulli trials

- Generate  $n$  Bernoulli( $p$ ) RVs (using the technique shown above)  $Y_1, Y_2, Y_3, \dots, Y_n$
- Set  $X = Y_1 + Y_2 + \dots + Y_n$  (sum of  $n$  Bernoulli RVs)
- If  $Y_i = 1$ , number of successes out of  $n$  increases
- Total no. of successes =  $\sum_{i=1}^n Y_i$

## Generation of Poisson RVs

$$X \sim \text{Poisson}(\lambda)$$

$$P(X=i) = \frac{\lambda^i}{i!} e^{-\lambda}$$

$i$  number of events in a unit time interval

### Method 1

- Generate exponential inter-event times  $Y_1, Y_2 \dots$  with mean 1
- Let  $I$  be the smallest index such that

$$\sum_{i=1}^{I+1} Y_i > \lambda$$

- Set  $X = I$

## Method 2

- Generate  $U(0,1)$  RVs  $U_1, U_2 \dots$
- Let  $N$  be the smallest index such that

$$\prod_{i=1}^{N+1} U_i < e^{-\lambda}$$

## Steps

1. Set  $i=0, P=1$
2. Generate  $U_{i+1}$  from  $U(0,1)$  and replace  $P$  with  $P \cdot U_{i+1}$
3. If  $P < e^{-\lambda}$ , accept  $N=i$  and go to step 1.

Else, reject  $N=i$  and increment  $i$  and return to step 2.

Upon completion of step 3,  $P = \prod_{i=1}^{N+1} U_i$

- If  $N=n$ , then  $n+1$  RVs are requested. So, the average number is given by  $E(N+1) = \lambda + 1$

Q26. Generate 3 Poisson variates with mean  $\lambda=0.2$  for the random numbers  $R=0.4357, 0.4146, 0.8353, 0.9952, 0.8004$

$$e^{-\lambda} = e^{-0.2} = 0.8187$$

1. For  $X_1$

$$i=0, P=1, R=0.4357$$

$$(a) P = 0.4357 \times 1 = 0.4357 < 0.8187 \quad \checkmark$$

(b) Accept  $N=0$

$$(c) X_1 = 0$$

2. For  $X_2$

$$i=0, P=1, R=0.4146$$

$$(a) P = 1 \times 0.4146 = 0.4146 < 0.8187 \quad \checkmark$$

(b) Accept  $N=0$

$$(c) X_2 = 0$$

3. For  $X_3$

$$i=0, P=1, R_1=0.8353, R_2=0.9952, R_3=0.8004$$

$$(a) P = 1 \times 0.8353 = 0.8353 > 0.8187 \quad \times \quad i=0$$

$$(b) P = 0.8353 \times 0.9952 = 0.8313 > 0.8187 \quad \times \quad i=1$$

$$(c) P = 0.8313 \times 0.8004 = 0.6654 < 0.8187 \quad \checkmark \quad i=2$$

(d) Accept  $N=2$

$$(e) X_3 = 2$$

Poisson numbers : 0, 0, 2

## Generation of Normal RVs

### Acceptance / Rejection Technique

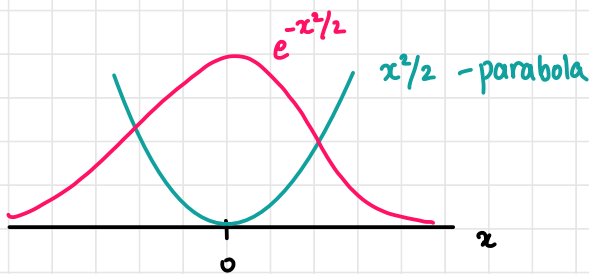
- If  $X \sim N(0, 1^2)$ , the pdf of  $X$  is

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

reason for  $1/\sqrt{2\pi}$  :

$$\int_{-\infty}^{\infty} e^{-x^2/2} dx = \sqrt{2\pi}$$

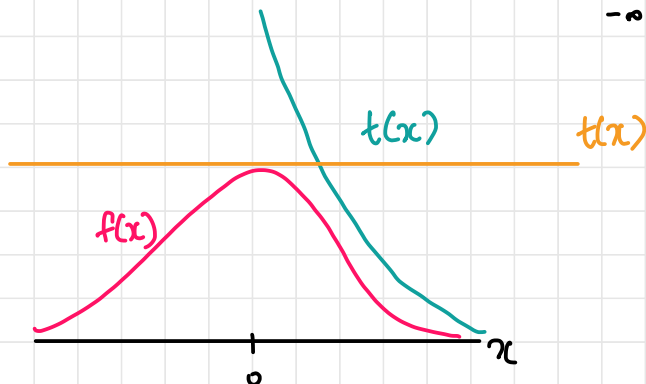
and  $\int_{-\infty}^{\infty} f(x) dx = 1$



- Let a function  $t(x)$  majorise pdf  $f(x) \Rightarrow t$  is NOT a density

$$t(x) \geq f(x) \quad \forall x$$

$$\left( \int_{-\infty}^{\infty} t(x) dx \geq 1 \right)$$



- total area

$$c = \int_{-\infty}^{\infty} t(x) dx \geq \int_{-\infty}^{\infty} f(x) dx = 1$$

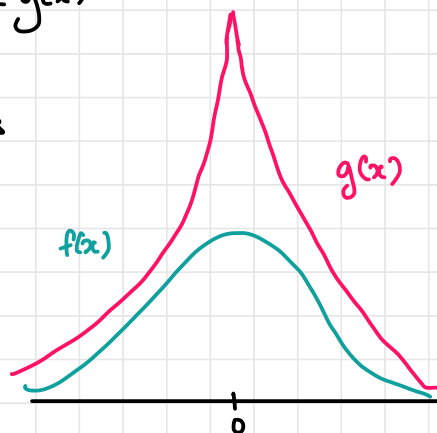
- However,  $\frac{t(x)}{c}$  is a density

$$\text{Let } g(x) = \frac{t(x)}{c} \Rightarrow t(x) = c g(x)$$

- If  $X \sim N(0, 1^2)$ , the pdf of  $|X|$  is

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

$$\text{Let } g(x) = \sqrt{\frac{2e}{\pi}} e^{-x}$$



### Steps

1. Generate exponential  $Y$  with mean 1
2. Generate  $U$  from  $U(0, 1)$
3. If  $U \leq e^{-(Y-1)^2/2}$ , accept  $Y$

$$\frac{f(Y)}{g(Y)} \leq c$$

Else, go to step 1

4. Return  $X = Y$  or  $X = -Y$  with probability 0.5



## Box-Muller Method

- Box-Muller Transform transforms from a two-dimensional uniform distribution to a two-dimensional bivariate normal distribution (complex normal distribution)
- If  $U_1$  and  $U_2$  are independent RVs from  $U(0,1)$

$$\begin{aligned} z_1 &= \sqrt{-2 \ln U_1} \cos(2\pi U_2) \\ z_2 &= \sqrt{-2 \ln U_1} \sin(2\pi U_2) \end{aligned} \quad \leftarrow \text{for single-variable normal distribution}$$

No need to remember this:

- By solving for  $U_1$  and  $U_2$

$$z_1^2 + z_2^2 = -2 \ln U_1$$

$$\frac{z_2}{z_1} = \tan(2\pi U_2)$$

$$-\ln U_1 = \frac{z_1^2 + z_2^2}{2}$$

$$2\pi U_2 = \tan^{-1}\left(\frac{z_2}{z_1}\right)$$

$$U_1 = e^{-\frac{(z_1^2 + z_2^2)}{2}}$$

$$U_2 = \frac{1}{2\pi} \tan^{-1}\left(\frac{z_2}{z_1}\right)$$

- Taking the Jacobian  $\frac{\partial(U_1, U_2)}{\partial(z_1, z_2)}$

$$\begin{vmatrix} \frac{\partial U_1}{\partial z_1} & \frac{\partial U_1}{\partial z_2} \\ \frac{\partial U_2}{\partial z_1} & \frac{\partial U_2}{\partial z_2} \end{vmatrix} = - \left( \frac{1}{\sqrt{2\pi}} e^{-z_1^2/2} \right) \left( \frac{1}{\sqrt{2\pi}} e^{-z_2^2/2} \right)$$

Q27. Suppose the height of adult males in a certain area is normally distributed with a mean of 168 cm and a standard deviation of 8 cm. Simulate the height of 4 adults.

$$X \sim N(168, 8^2)$$

$$\begin{aligned}\mu &= 168 \\ \sigma &= 8 \\ n &= 4\end{aligned}$$

$$Z = \frac{X - \mu}{\sigma}$$

- We first generate random  $Z$  values

- $Z = \sqrt{-2 \ln(U_1)} \cos(2\pi U_2)$

$X_1$

$$\text{Let } U_1 = 0.2432, U_2 = 0.5214$$

$$Z_1 = \sqrt{-2 \ln(0.2432)} \cos(2\pi \times 0.5214)$$

$$Z_1 = -1.666$$

$$X_1 = Z_1 \times \sigma + \mu = -1.666 \times 8 + 168$$

$$X_1 = 164.67 \text{ cm}$$

$X_2$

$$\text{Let } U_1 = 0.8921, U_2 = 0.6232$$

$$Z_2 = \sqrt{-2 \ln(0.8921)} \cos(2\pi \times 0.6232)$$

$$Z_2 = -0.3417$$

$$X_2 = 165.27 \text{ cm}$$

$x_3$

$$\text{Let } U_1 = 0.4421, U_2 = 0.0012$$

$$z_3 = \sqrt{-2 \ln(0.4421)} \cos(2\pi \times 0.0012)$$

$$z_3 = 1.2776$$

$$x_3 = 178.22 \text{ cm}$$

$x_4$

$$\text{Let } U_1 = 0.9921, U_2 = 0.7324$$

$$z_4 = \sqrt{-2 \ln(0.9921)} \cos(2\pi \times 0.7324)$$

$$z_4 = -0.0139$$

$$x_4 = 167.89 \text{ cm}$$

Q25. Suppose the no. of shipments,  $x$ , on the loading dock of a company is either 0, 1 or 2. Generate RVs given  $U = 0.23, 0.52, 0.81, 0.34$

$x$	$P(x)$	$F(x)$
0	0.5	0.5
1	0.3	0.8
2	0.2	1.0

(discrete RVs - inverse)

$$\text{Let } U \sim U(0,1)$$

$$x = \begin{cases} 0 & U \leq 0.5 \\ 1 & 0.5 < U \leq 0.8 \\ 2 & U > 0.8 \end{cases}$$

$$U_1 = 0.23 \Rightarrow U < 0.5 \Rightarrow x_1 = 0$$

$$U_2 = 0.52 \Rightarrow 0.5 < U \leq 0.8 \Rightarrow x_2 = 1$$

$$U_3 = 0.81 \Rightarrow U > 0.8 \Rightarrow x_3 = 2$$

$$U_4 = 0.34 \Rightarrow U < 0.5 \Rightarrow x_4 = 0$$