Summary on Discrete Random Variables

1. Bernoulli (p)

$$X=\{0, 1\}, 1-> success, 0-> failure$$

 $Pr(X=1) = p, Pr(X=0)=1-p$

- 2. Discrete uniform (a, b)
- 3. Geometric (p) $P = P_1 L / \neq 0$
- 4. Binomial (n, p)
 - f(x): probability of x successes in n independent Bernoulli trials.

Summary on Discrete Random Variables- cont.

5. Poisson (μ) – limiting case of binomial when

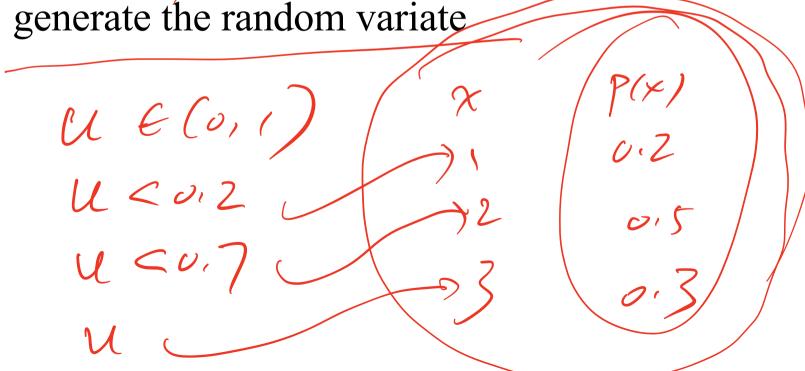
$$n \to \infty, p \to 0$$
 and $np \to \lambda$

$$f(n) = \frac{\lambda^n e^{-\lambda}}{n!}, n = 0, 1, 2, \dots$$

Model arrival process with a large population. If interarrival time has exponential distribution, arrival time follow Poisson distribution.

Summary on Discrete Random Variables- cont.

- 6. Empirical Distribution
 - based on observed data
 - use the Inverse-Transform method to



Random Variae Generation

Special methods for some common distributions

Bernoulli Trial

Parameter: Pr(X=1) = p

- (a) Generate u
- (b) If u<p then variate =1 else variate =0

Binomial

- (a) Perform algorithm for Bernoulli trial n times
- (b) Variate = number of successes

Geometric

(a)Perform a series of Bernoulli trials until the first failure is encountered

(b) Variate = number of successes before the first failure

Discrete Uniform

- (a) Generate u
- (b) Variate = a + Integer (u*(b-a+1))

Empirical Distribution

- Based on observed data
- Inverse Transformation Method

• Example:

$$P(1) = 0.1, P(2) = 0.5$$

$$P(3) = 0.2, P(4)=0.15, P(5)=0.05$$

Inverse Transformation Method

• Suppose u is a random number uniformly distributed between 0 and 1, the following table can be used to select the variate:

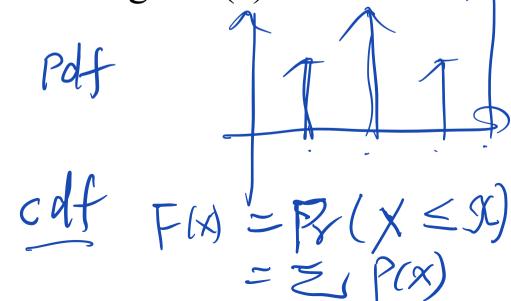
u	SelectedVariate
0–0.1	1
0.1–0.6	2
0.6-0.8	3
0.8-0.95	4
0.95-1.0	5



Algorithm

P(Xi)

- (a) Generate u
- (b) i=1
- (c) u=u-p(i)
- (d) If u<0 then random variate is given by i Else increment i and go to (c)



Continuous Random Variables

- A random variable X is continuous if its values are over a continuous interval.
- Properties of pdf:

$$(a)f(x) >= 0$$

$$(b)\Pr(a \le x \le b) = \int_{a}^{b} f(x) dx$$

(c) By definition:
$$\int_{a}^{\infty} f(x) dx = 1$$

Pat
$$f(x)$$

Soft $f(x)$ $dx = 1$

Pat $f(x)$

Pat :

 $f(x) \ge 0$

$$\int_{a}^{b} f(x) dx = 1$$

$$\int_{a}^{b} f(x) dx = 1$$

Pr $(a_1 \le x \le b_1) = \int_{a_1}^{b_1} f(x) dx$

Pat :

$$\int_{a}^{b} f(x) = \int_{a_1}^{b_1} f(x) dx$$

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$$\int_{a_1}^{b_1} f(x) dx = \int_$$

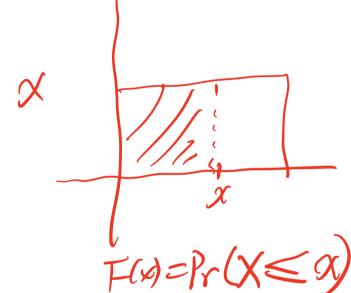
$$\left(\int_0^{\alpha} f(t)dt\right)_{\alpha} = f(\alpha)$$

Continuous Random Variables – cont.

• Properties of cdf:

$$F(x) = \Pr(X \le x) = \int_{0}^{x} f(t) dt$$

pdf and cdf: $f(x) = F(x) = \frac{dF}{dx}$



Mean:
$$\mu = \int_{0}^{\infty} x f(x) dx$$

Standard deviation: $\sigma = \sqrt{\int_{0}^{\infty} x^2 f(x) dx - \mu^2} = \sqrt{\int_{0}^{\infty} (x - \mu)^2 f(x) dx}$

Variance: σ^2

$$6^2 = \int_0^\infty (x - \mu)^2 f(x) dx$$

$$\mu = \sum_{0}^{\infty} f(x_{0}) x_{0}$$

$$\mu = \int_{0}^{2} f(x_{0}) x_{0}$$

$$\mu = \int_{0}^{2} f(x_{0}) dx$$

$$= \sum_{0}^{2} f(x_{0}) - \mu^{2}$$

$$= \sum_{0}^{2} f(x_{0}) - \mu^{2}$$

$$= \sum_{0}^{2} f(x_{0}) dx - \mu^{2}$$

$$= \int_{0}^{\infty} x^{2} f(x_{0}) dx - \mu^{2}$$

Continuous Uniform (0, 1)

$$P_{Y}(X \le \frac{1}{2}) = \frac{1}{2}$$

$$P_{Y}(X \le \frac{1}{4}) = \frac{1}{5}$$

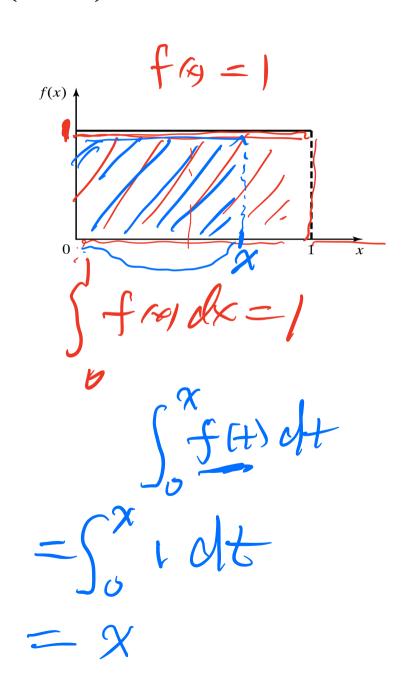
$$f(x) = \begin{cases} 1, & 0 \le x \le 1 \\ 0, & \text{otherwise} \end{cases}$$

$$E(R) = \int_{0}^{1} x \, dx = \frac{x^{2}}{2} \Big|_{0}^{1} = \frac{1}{2}$$

$$cdf: F(x) = \Pr(X \le x)$$

$$= (f(t))dt = \int_0^x dt$$

$$(F(x)) = X$$

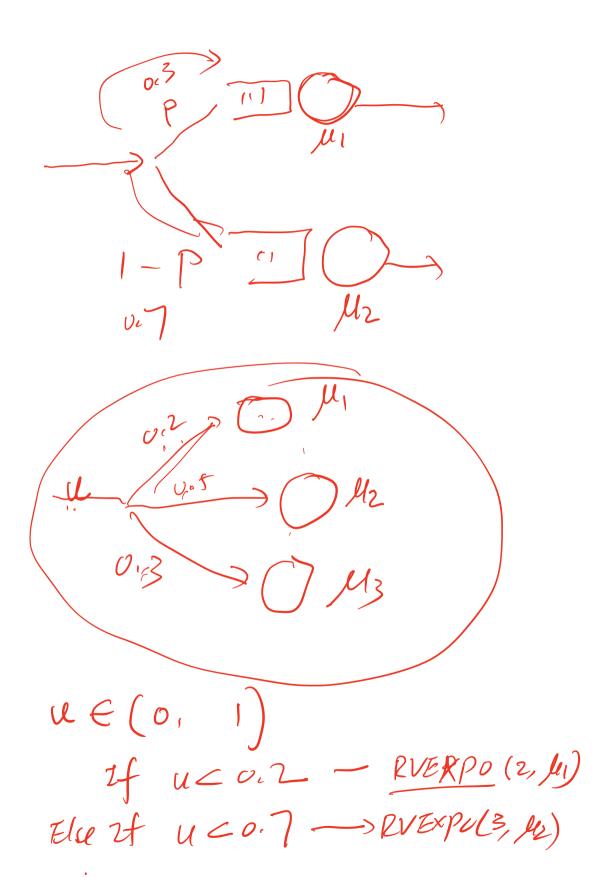


 $\mu = \iint_{0}^{1} x \, dx = \frac{1}{2}.$

Examples of Continuous Random Variables

- Continuous Uniform (a, b)
- Exponential (μ) $\alpha = -\mu \ln(1 \mu)$
- Erlang (b, n) Combination of exponential random variable, sum of n exponentially distributed random variables each with parameter b:

• Hyper exponential – weighted sum of exponentials



Else ~ prespoliss)

Variate generation for Erlang Distribution

- (a) Generate k exponential variates, each with parameter μ
- (b) Variate = sum of these k variates

20-20-20

Variate generation for Hyper-exponential

- (a)Generate a random number for Bernoulli trial
- (b)If less than pgenerate an exponential variate withparameter μ1Else

generate an exponential variate with parameter μ_2

The Inverse-Transform Technique for Continuous Distributions

Algorithm

Step 1: Compute the *cdf* for F(x)

Step 2: Set $F(x)=\hat{u}$ on the range of x

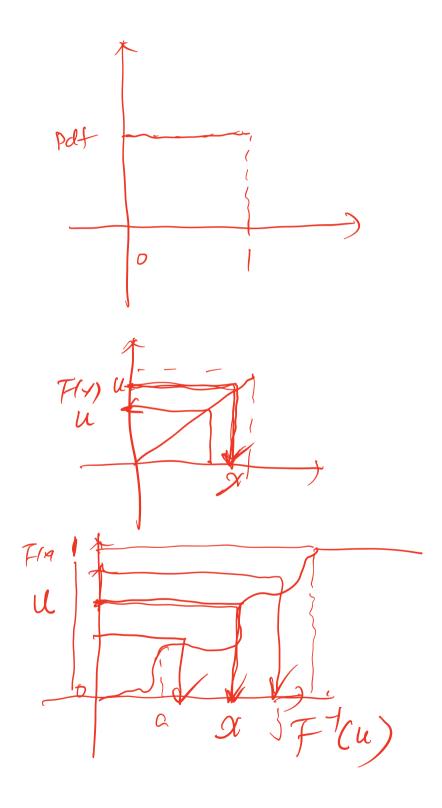
u is uniform distribution over (0,1)

Step 3: Solve the equation F(x)=u in term of u,

$$x=F^{-1}(u) \tag{1}$$

// (1) is called the random-variate generator

Step 4: Generate
$$u_1, u_2, ..., x_1 = F^{-1}(u_1), x_2 = F^{-1}(u_2), x_3 = F^{-1}(u_3), ...$$



Pot
$$\int_{0}^{x} f(x) dt$$

$$= \int_{0}^{2} \frac{1}{3} dt$$

FIN=PI(X=X) 1=x<2 $=\frac{1}{3}+(\chi-1)-\frac{2}{3}$ $=\frac{1}{3}+\frac{2}{3}x-\frac{2}{3}$ -) Thy=3x-1

$$F(M) = \int_{3}^{3} x \int_{6}^{6} x \leq 1$$

$$\int_{3}^{2} x - \frac{1}{3} = 1 \leq x \leq 2$$

$$\int_{3}^{2} x = u \implies x = 3u$$

$$\int_{3}^{2} x = u \implies x = 3u$$