### Discrete-Event Simulation

# Exponential and Geometric Variates

### Geometric Distribution

 A coin has p as its probability of a head. Toss it until the first tail occurs. If X is the number of heads,

$$\mathcal{X} = \{x \mid x = 0, 1, 2, ...\}$$

The pdf is

$$P_{x}(x=x) - f(x) = p^{x}(1-p), x = 0, 1, 2, ...$$

X is Geometric(p) and the set of possible values is infinite

$$\sum_{x=0}^{\infty} p^{x} (1-p) = (1-p)(1+p+p^{2}+p^{3}+...) = 1$$

 $\Sigma f(x) = 1 ?$ 

$$P_{Y}(X=1)=P, P_{Y}(X=0)=I-P^{2}$$

$$H \rightarrow P, T \rightarrow X \downarrow R$$
 $T \rightarrow 0, I-P$ 
 $H \downarrow T \rightarrow 1: P(I-P)$ 
 $H \downarrow H \downarrow T \rightarrow 2: PP(I-P)$ 
 $H \downarrow H \downarrow H \downarrow T \rightarrow 3: P^{3}(I-P)$ 
 $(I-P) + P(I-P) + P^{3}(I-P) + P^{3}(I-P) + \cdots$ 
 $= X = 0$ 
 $= (I-P) \cdot [I+P+P^{3}+P^{3}+P^{4}+\cdots P^{3}+P^{4}+\cdots]$ 
 $= (I-P) \cdot [I+P+P^{3}+P^{3}+P^{4}+\cdots P^{3}+P^{4}+\cdots]$ 
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### **Cumulative Distribution Function**

• The cumulative distribution function (cdf) of the discrete random variable X is the real-valued function  $F(\cdot)$  for each  $x \in X$ :

$$F(x) = \Pr(X \le x) = \sum_{t \le x} f(t)$$

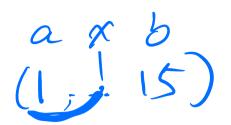
• If X is Equilibely(1, 6),

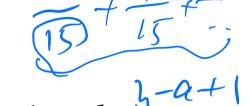
$\mathcal{X}$	1	2	3	4	5	6
f(x) = Pr(X=x)	1/6	1/6	1/6	1/6	1/6	1/6
$F(x)=Pr(X \leq x)$	1/6	2/6	3/6	4/6	5/6	1

$$F(2) = Pr(X \leq 2) = P_r(X \leq 1) + R(X = 2)^3$$

$$=F(1)+f(2)$$

# Examples - cdf (1) (5)





• If X is Equilikely(a, b) then the cdf is
$$F(x) = \sum_{t=a}^{x} \frac{1}{b-a+1} = \frac{x-a+1}{b-a+1}, \quad x = a, a+1, ..., b$$

If X is Geometric(p) then the cdf is

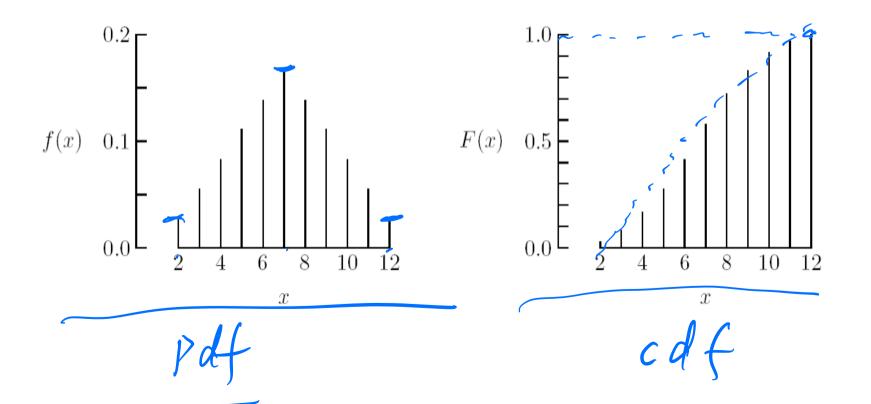
$$F(x) = \sum_{t=0}^{x} p^{t} (1-p) = (1-p)(1+p+...+p^{x}) = 1-p^{x+1},$$

$$= 1 - P^{x+1}x = 0, 1, 2, ...$$

$$= 1 - P^{x+1}x = 1 - P^{x+1}x =$$

## pdf and cdf – sum of two dice

 No simple equation for F(·) for sum of two dice.



# Relationship Between cdfs and pdfs

• A cdf can be generated from its corresponding pdf by recursion. For example,  $X = \{x \mid x = a, a + 1, ..., b\}$ :

F(a) = f (a),  
F(x) = F(x - 1) + f (x), 
$$x = a + 1, a + 2, ..., b$$
.

A pdf can be generated from its corresponding cdf by subtraction.

f (a) = F(a)  
f (x) = F(x) - F(x - 1), 
$$x = a + 1, a + 2, ..., b$$
.

 A discrete random variable can be defined by specifying either its pdf or its cdf.

### Other cdf Properties

A cdf is strictly monotone increasing:

if 
$$x_1 < x_2 \implies F(x_1) < F(x_2)$$

- The cdf values are bounded between 0.0 and 1.0.
- Monotonicity of F(·) is the basis to generate discrete random variates.

### Mean and Variance

- The mean is the measure of the central tendency of a random variable where the variance is the spread of possible values around the mean.
- The mean of a random variable is also known as the expected value:

$$E[X] = \sum_{x} xf(x) = \mu$$

$$\mu = z \alpha f(x)$$

#### Mean and Standard Deviation

The mean µ of the discrete random variable X is

$$\mu = \sum_{x} x f(x)$$

• The corresponding standard deviation  $\sigma$  is

$$\sigma = \sqrt{\sum_{x} (x - \mu)^2 f(x)} \quad \text{or} \quad \sigma = \sqrt{\left(\sum_{x} x^2 f(x)\right) - \mu^2}$$

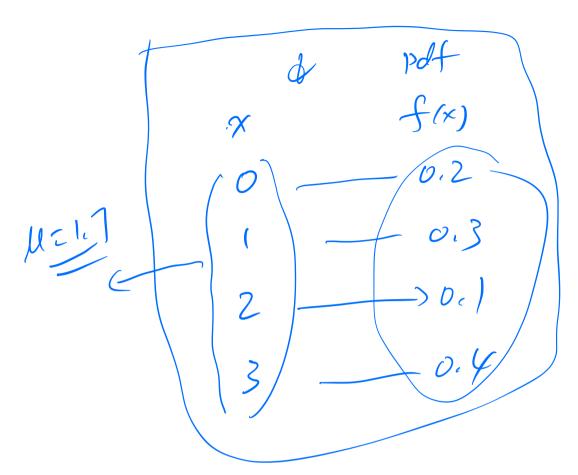
• The variance is  $\sigma^{2}$ .

$$5^2 = \sum_{x} (x - \mu)^2 f(x)$$
Mean.

## Example

• If X is Geometric (p) random variable, what are the mean, variance and standard deviation?

P(1-P)
P2(1-P)



Mean :  $\mu = 0 \times 0.2 + 1 \times 0.3 + 2 \times 0.1$   $+ 3 \times 0.4$  = 0.3 + 0.2 + 1.2 = 0.5 + 1.2 = 1.7  $= (0 - 1.7)^2 \cdot 0.2 + (1 - 1.7)^2 \cdot 0.3$   $+ (2 - 1.7)^2 \cdot 0.1 + (3 - 1.7)^2 \cdot 0.4$ = 6 = 0