

*CDA6530: Performance Models of Computers and Networks*

***Chapter 8: Using Simulation for  
Statistical Analysis and Verification***

# Objective

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- ❑ We have learned how to generate random variables in simulation, but
- ❑ How can we use them?
- ❑ What is the purpose for such simulation?

# Example: Generate discrete R.V.

- A loaded dice, r.v.  $X$ : number shown up
  - $P(X=1) = 0.05$ ,  $P(X=2) = 0.1$ ,  $P(X=3) = 0.15$ ,  
 $P(X=4) = 0.18$ ,  $P(X=5) = 0.22$ ,  $P(X=6) = 0.3$
- Q1: Simulate to generate 100 samples

$$X = \begin{cases} x_0 & \text{if } U < p_0 \\ x_1 & \text{if } p_0 \leq U < p_0 + p_1 \\ \vdots & \\ x_j & \text{if } \sum_{i=0}^{j-1} p_i \leq U < \sum_{i=0}^j p_i \\ \vdots & \end{cases} \quad X = \begin{cases} 1 & \text{if } U < 0.05 \\ 2 & \text{if } 0.05 \leq U < 0.15 \\ 3 & \text{if } 0.15 \leq U < 0.3 \\ 4 & \text{if } 0.3 \leq U < 0.48 \\ 5 & \text{if } 0.48 \leq U < 0.7 \\ 6 & \text{if } 0.7 \leq U \end{cases}$$

Code in Matlab

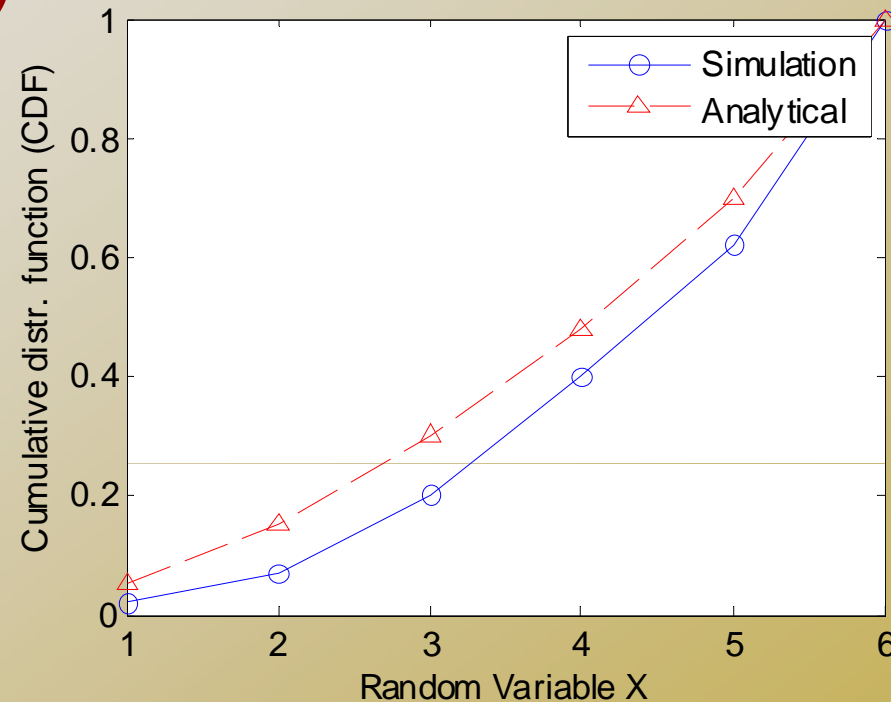
# ***Draw CDF (cumulative distr. function)***

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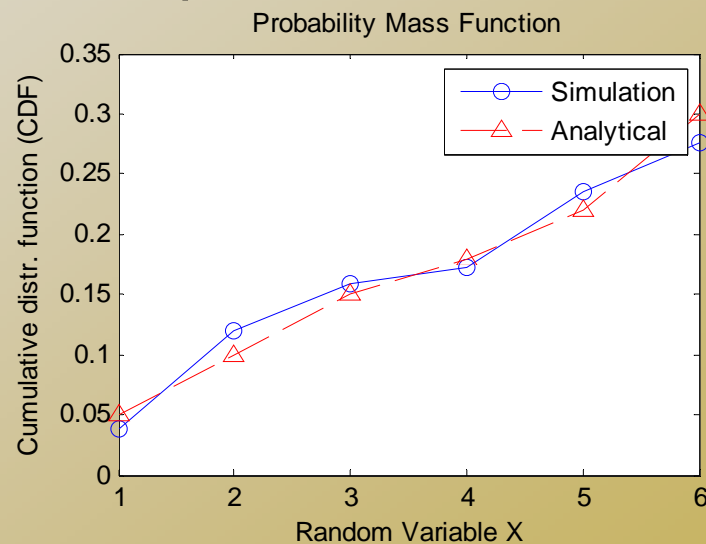
- - Remember  $F(x) = P(X \leq x)$
  - For our question, r.v.  $X$  has 6 different sample values, thus we could derive from simulation:  
 $F(1), F(2), F(3), F(4), F(5), F(6)$
  - $F(6) = 1$ , so we need to derive the other five
  - $F(x) = m/n$  where
    - $n$ : # of sample values we generate
    - $m$ : # of sample r.v. values  $\leq x$

- ❑ The CDF curve for our example in one simulation run (compared with analytical results)



# Draw PMF (Probability Mass Function)

- Pmf:  $P(X=1) = 0.05$ ,  $P(X=2) = 0.1$ ,  $P(X=3) = 0.15$ ,  
 $P(X=4) = 0.18$ ,  $P(X=5) = 0.22$ ,  $P(X=6) = 0.3$
- $PMF(x) = m/n$ 
  - $n$ : # of sample values we generate
  - $m$ : # of sample r.v. values =  $x$



# Continuous R.V.

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- Use inverse transform method:
  - One value of  $U \rightarrow$  one r.v. sample
- Normal distr. use the polar method to generate
- How to draw CDF?
  - Problem: r.v.  $x$  could be any value
  - Solve: determine  $x_i$  points to draw with fixed interval ( $i=1,2,\dots$ )
  - $F(x_i) = P(X \leq x_i) = m/n$ 
    - $n$ : # of samples generated
    - $m$ : # of sample values  $\leq x_i$

# Analytical Results

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- ❑ Use the formula of the distribution to directly calculate  $F(x_i)$
- ❑ How to calculate integration in Matlab?
  - ❑ Use function `quad()`

```
Q = quad(@myfun,0,2);
```

```
%where myfun.m is the M-file function:
```

```
function y = myfun(x)
```

```
    y = 1./(x.^3-2*x-5);
```



# Markov Chain Simulation

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- **Discrete-time Markov Chain**
  - Simulate N steps
  - For each step, use random number U to determine which state to jump to
    - Similar to discrete r.v. generation
  - $\pi(i) = m_i/N$ 
    - N: # of simulated steps
    - $m_i$ : number of steps when the system stays in state i.

# Markov Chain Simulation

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- ❑ **Continuous-time Markov Chain**
  - ❑ Method #1:
    - ❑ Determine how long current state lasts
      - ❑ Generate exponential distr. r.v.  $X$  for the staying time
    - ❑ Determine which state the system jumps to
      - ❑ Similar to discrete-time Markov Chain jump simulation
  - ❑ Method #2:
    - ❑ Determine the jumping out time for each jump out link (everyone is expo. Distr.)
    - ❑ The staying time is the shortest jumping out time
    - ❑ The outgoing state corresponds to this shortest jumping out time
  - ❑ Method #2 is more intuitive reasonable

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- $\pi(i) = \sum t_k(i) / T$ 
    - T: overall simulated time
    - $t_k(i)$ : the time when the system is in state i for the k-th time