

CSCE 619-600

Networks and Distributed Processing

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## Review of Probability III

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# Agenda

- Expectations
  - Of random variables
  - Of functions
- Examples

# Expectations

- Expected value of a random variable
  - Also called the **mean**
  - Intuitive definition: if we generate infinitely many samples of  $X$ , their average will be  $E[X]$
- Definition: expectation  $E[X]$  is given by:

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

- For a discrete variable with outcomes  $x_1, x_2, \dots$

$$E[X] = \sum_{i=1}^{\infty} x_i P(X = x_i)$$

## Expectations 2

- For **non-negative** variables, there is a very useful alternative formula for  $E[X]$ 
  - Recall integration by parts

$$\int_0^{\infty} u dv = uv \Big|_0^{\infty} - \int_0^{\infty} v du$$

- From which

$$\begin{aligned} \int_0^{\infty} x f(x) dx &= \int_0^{\infty} x dF(x) = - \int_0^{\infty} x dF^c(x) \\ &= -x F^c(x) \Big|_0^{\infty} + \int_0^{\infty} F^c(x) dx \\ &= \int_0^{\infty} F^c(x) dx \end{aligned}$$

# Expectations 3

- Thus, for non-negative continuous variables:

$$E[X] = \int_0^{\infty} F^c(x) dx$$

– Integration of the tail yields the mean

- Similarly, for non-negative integer-valued variables:

$$E[X] = \sum_{i=0}^{\infty} P(X > i)$$

- Example: compute the mean of exponential  $X$

$$E[X] = \int_0^{\infty} e^{-\lambda x} dx = \frac{e^{-\lambda x}}{-\lambda} \Big|_0^{\infty} = \frac{1}{\lambda}$$

rate

## Expectations 4

- Example: compute the expectation of the Pareto distribution, i.e.,  $F(x) = 1 - (1+x/\beta)^{-\alpha}$  for  $x \in [0, \infty)$

$$E[X] = \int_0^{\infty} (1+x/\beta)^{-\alpha} dx = \frac{\beta}{1-\alpha} (1+x/\beta)^{1-\alpha} \Big|_0^{\infty}$$

- which reduces to

$$E[X] = \frac{\beta}{\alpha - 1}$$

- Also notice that if  $\alpha < 1$ , then  $E[X] = \infty$
- Example: compute the mean wait time  $E[W_t]$

## Expectations 5

$$E[XY] = \int \int xyf(x, y)dydx$$

- Expectations are linear:

$$E[cX] = cE[X], \quad E[X + Y] = E[X] + E[Y]$$

- For independent  $X, Y$ :  $E[XY] = E[X]E[Y]$

- Definition: for arbitrary functions  $u(\cdot)$

$$E[u(X)] = \int_{-\infty}^{\infty} u(x)f(x)dx$$

- Example: compute  $E[e^X]$  for uniform  $X$  in  $[0, s]$

$$E[e^X] = \frac{1}{s} \int_0^s e^x dx = \frac{e^s - 1}{s}$$

# Expectations 6

- What the last example means is that if you generate a sequence of random numbers  $x_1, \dots, x_n$  using the distribution of  $X$ , then the mean

$$\lim_{n \rightarrow \infty} \frac{e^{x_1} + \dots + e^{x_n}}{n} = \frac{e^s - 1}{s}$$

- Simulation:  $s = 5$ 
  - I generated 10,000 random numbers in  $[0,5]$
  - Then averaged  $e^{x_i}$  and obtained 29.35
  - The analytical result predicts 29.48
  - Notice that this does not equal  $e^{E[X]} = 12.18$



# Examples

- Note: unless specified otherwise, all variables in various problems are **independent**
- Compute  $P(3X < Y)$  for  $X$  uniform in  $[-2, 3]$  and  $Y$  exponential with mean  $1/\lambda$
- Assume  $X$  is discrete with PMF  $p_j = P(X = j)$ , where  $p_1 = 0.3$ ,  $p_{34} = 0.5$ , and  $p_{90} = 0.2$ 
  - Compute the distribution of  $Y = e^X$
- Compute  $P(X > 2 \mid X < 3)$  for exponential  $X$  with  $\lambda$
- How about  $P(X + Y < 10 \mid X < 3)$  for  $X$  uniform in  $[0, 5]$  and  $Y$  exponential with rate  $\lambda$
- Compute  $E[3X^2Y]$  for  $X$  uniform in  $[0, 1]$  and  $Y$  Pareto