

# CSE 312 24SP

Section 8: Tail Bounds, Joint Distributions, and the  
Law of Total Expectation

# Review

	Discrete	Continuous
Joint PMF/PDF	$P_{X,Y}(x, y) = P(X = x, Y = y)$	$f_{X,Y}(x, y) \neq P(X = x, Y = y)$
Joint support $\Omega_{X,Y}$	$\{(x, y) \in \Omega_X \times \Omega_Y : p_{X,Y}(x, y) > 0\}$	$\{(x, y) \in \Omega_X \times \Omega_Y : f_{X,Y}(x, y) > 0\}$
Joint CDF	$F_{X,Y}(x, y) = \sum_{t \leq x, s \leq y} p_{X,Y}(t, s)$	$F_{X,Y}(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(t, s) ds dt$
Normalization	$\sum_{x,y} p_{X,Y}(x, y) = 1$	$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx dy = 1$
Marginal PMF/PDF	$p_X(x) = \sum_y p_{(X,Y)}(x, y)$	$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy$
Expectation	$E[g(X, Y)] = \sum_{x,y} g(x, y) p_{X,Y}(x, y)$	$\begin{aligned} E[g(X, Y)] \\ = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X,Y}(x, y) dx dy \end{aligned}$
Requirements for Indep.	$\begin{aligned} \forall x \forall y p_{X,Y}(x, y) &= p_X(x) p_Y(y) \\ \Omega_{X,Y} &= \Omega_X \times \Omega_Y \end{aligned}$	$\begin{aligned} \forall x \forall y f_{X,Y}(x, y) &= f_X(x) f_Y(y) \\ \Omega_{X,Y} &= \Omega_X \times \Omega_Y \end{aligned}$

# Review (cont.)

- **Law of Total Probability (r.v. version):** If  $X$  is a discrete random variable, then

$$P(A) = \sum_{x \in \Omega_X} P(A|X = x)p_X(x)$$

( $X$  is discrete.)

- **Law of Total Expectation (event version):** Let  $X$  be a discrete random variable, and let events  $A_1, \dots, A_n$  partition the sample space. Then,

$$E[X] = \sum_{i=1}^n E[X|A_i]P(A_i)$$

- **Law of Total Expectation (r.v. version):** Suppose  $X$  and  $Y$  are random variables. Then,

$$E[X] = \sum_y E[X|Y = y]p_Y(y)$$

( $Y$  is discrete.)

# Review (cont.)

- **Conditional Distributions (discrete case):**  $p_{X|Y}(x|y) = \frac{p_{X,Y}(x,y)}{p_Y(y)}$ . The continuous version is the same, but with a pdf instead of a pmf.
- **Conditional Expectation (discrete case):**  $E[X|Y = y] = \sum_x xp_{X|Y}(x|y)$ . The continuous version is the same, but with a pdf instead of a pmf (and integral instead of sum).
  - *Linearity of expectation still applies:*  $E[X + Y|A] = E[X|A] + E[Y|A]$
- **Covariance:** How “intertwined” are  $X$  and  $Y$ ?
  - $Cov(X, Y) = E[(X - E[X])(Y - E[Y])] = E[XY] - E[X]E[Y]$ 
    - Recall that  $Var(X) = E[(X - E[X])(X - E[X])]$
- **We have not covered the continuous versions of the Law of Total Probability and Law of Total Expectation**, but they should be intuitive given you understand the discrete versions: again, instead of pmfs, we have pdfs, and instead of sums, we have integrals.

# Review (cont.)

- **Markov's Inequality:** Let  $X$  be a nonnegative random variable and  $\alpha > 0$ . Then

$$P(X \geq \alpha) \leq \frac{E[X]}{\alpha}$$

- **Chebyshev's Inequality:** Let  $Y$  be a random variable with  $E[Y] = \mu$  and  $Var(Y) = \sigma^2$ . Then, for any  $\alpha > 0$ ,

$$P(|Y - \mu| \geq \alpha) \leq \frac{\sigma^2}{\alpha^2}$$

- **(Multiplicative) Chernoff Bound:** Suppose  $X \sim Bin(n, p)$  and  $\mu = np$ . Then for any  $0 < \delta < 1$ ,

$$P(X \geq (1 + \delta)\mu) \leq e^{-\frac{\delta^2 \mu}{3}}$$

$$P(X \leq (1 - \delta)\mu) \leq e^{-\frac{\delta^2 \mu}{2}}$$

- **Extra tidbit:** Other useful inequalities in probability include the **union bound** and **Hoeffding's Inequality** (in case you wanted to learn more). There are tons of them.

# Task 1: Kahoot!

See Task 1 in your section handout.

# Task 3: Exponential Tail Bounds

- Let  $X \sim \text{Exp}(\lambda)$  and  $k > \frac{1}{\lambda}$ .
  - (a) Use Markov's inequality to bound  $P(X \geq k)$ .
  - (b) Use Markov's inequality to bound  $P(X < k)$ .
  - (c) Use Chebyshev's inequality to bound  $P(X \geq k)$ .
  - (d) What is the exact formula for  $P(X \geq k)$ ?
  - (e) For  $\lambda k \geq 3$ , how do the bounds given in (a), (b), and (c) compare?

# Task 2(c): Tail Bounds

- Suppose  $X \sim \text{Bin}(6, 0.4)$ . Give an upper bound for  $P(X \geq 4)$  using the Chernoff bound.

# Task 4: Robbie is Late!

Suppose the probability Robbie is late to lecture on a given day is at most 0.01. Do not make any independence assumptions.

- (a) Use a union bound to bound the probability that Robbie is late at least once over a 30-lecture quarter.
- (b) Use a union bound to bound the probability that Robbie is **never** late over a 30-lecture quarter.
- (c) Use a union bound to bound the probability that Robbie is late at least once over a 120-lecture quarter.

# Task 10: Lemonade Stand

- Suppose I run a lemonade stand that costs me \$100 a day to operate. I sell a single lemonade for \$20. Every person who walks by my stand either buys a drink or doesn't (no one buys more than one). If it is raining,  $n_1$  people walk by my stand, and each buys a drink independently with probability  $p_1$ . If it isn't raining,  $n_2$  people walk by my stand, and each buys a drink independent with probability  $p_2$ . It rains each day with probability  $p_{rain}$  independently of every other day. Let  $X$  be my profit over the next week. In terms of  $n_1, n_2, p_1, p_2, p_{rain}$ , what is  $E[X]$ ?