

Probability

1. Union Bound

a. $\Pr(A \cup B \cup C) \leq \Pr(A) + \Pr(B) + \Pr(C)$

b.
$$\Pr(A_1 \cup \dots \cup A_n) \leq \sum_{i=1}^n \Pr(A_i)$$

c. Example: True or False

- Suppose X and Y are real random values.
- $\Pr(X + Y \geq 0) \leq \Pr(X \geq 0) + \Pr(Y \geq 0)$
- True
- If $X + Y \geq 0$, at least one of X or Y should be greater than 0
- $\{X, Y: X + Y \geq 0\}$ is a subset of $\{X, Y: X \geq 0 \cup Y \geq 0\} \leq \Pr(X \geq 0) + \Pr(Y \geq 0)$

2. Variance

a. Is $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$? \rightarrow False

b. $\text{Var}(x) = E((x - \mu)^2)$

c. $\text{Var}(X + Y) = E((X + Y - \mu_x - \mu_y)^2)$
 $= E((X - \mu_x + Y - \mu_y)^2)$

d. Covariance

$$E((X - E(X))(Y - E(Y)))$$

$$= E(XY - X\mu_y - Y\mu_x + \mu_x\mu_y)$$

e. $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2 \cdot \text{coVar}(X, Y)$

3. Hats off (cont. from last time)

- a. Men with hats enter a room, and take off their hats. On their way out, hats are mixed up according to a random permutation, so each person receives a random hat. The probability each man gets his own hat is $1/n$

We define for each many i

$X_i = 1$ if that person receives his hat
 $= 0$ otherwise

$$\Pr(X_i = 1) = 1/n$$

$$\text{So, } E(X_i) = 1 \cdot 1/n + 0 \cdot (1 - 1/n) = 1/n$$

- Since the permutation is chosen uar, the probability each man gets his own hat is $1/n$

- Question: how many men in expectation will get their own hat?

- Answer:

$H_n = \#$ of men who get their own hat

$$E(H_n) = E(x_1 + \dots + x_n) = (\text{lin exp}) = (\sum_{i=1}^n E(x_i)) = n/n = 1$$

- Question: variance

- Answer:

$$\text{Var}(H_n) = \text{Var}(x_1 + \dots + x_n)$$

$$= E((x_1 + \dots + x_n - (\mu_1 + \dots + \mu_n))^2)$$

$$= E(((x_1 - \mu_1) + \dots + (x_n - \mu_n))^2)$$

$$= (\sum_{i=1}^n E((x_i - \mu_i)^2) + 2 * (\sum_{i < j} E((x_i - \mu_i)(x_j - \mu_j)))$$

$$\begin{aligned} \text{Var} \left[\sum_{i=1}^n X_i \right] &= \sum_{i=1}^n \text{Var}[X_i] + 2 \sum_{i < j} \text{Cov}[X_i, X_j] \\ &= \end{aligned}$$

$$\begin{aligned} \text{Reminder: } \text{Cov}[X, Y] &= [(X - EX)(Y - EY)] \\ &= E[XY] - (EX)(EY). \end{aligned}$$

$$\begin{aligned} \text{a. } \text{Var}(X_i) &= E((x_i - p)^2) \\ &= E[X_i^2] \\ &= 1/n \end{aligned}$$

$$\begin{aligned} \text{b. } \text{Cov}(X_i, X_j) &= E(X_i * X_j) - E(X_i) * E(X_j) \\ &= E(X_i * X_j) - 1 * 1 \end{aligned}$$

$$\text{Let } X_i * X_j = Z$$

$$Z = 1 \text{ if } X_i = X_j = 1$$

$$= 0 \text{ otherwise}$$

$$E(Z) = 1/(n) * 1/(n-1)$$

$$= n * 1/n * 1/n + n * (n-1) * 1/n * 1/(n-1) - 1$$

$$= 1 + 1 - 1$$

$$= 1$$

4. Linearity of Expectation

- a. Expected value of the sum of random variables is equal to the sum of their individual expectations
 - Important point: this holds regardless of whether they are independent
- b. Let X_1, \dots, X_k be random variables, c_1, \dots, c_k constants

$$\mathbb{E} \left[\sum_{i=1}^k c_i X_i \right] = \sum_{i=1}^k c_i \mathbb{E}[X_i]$$

c.

5. Digression

- a. Probability inequalities
 - Markov's inequality
 - Chebyshev's inequality
- b. Notions of convergence of random variables
- c. Limit theorems
 - Weak and strong laws of large numbers (LLN)
 - Central Limit Theorem (CLT)

6. Markov's inequality

- a. Let X be a non-negative random variable and suppose that $\mathbb{E}[X]$ exists

$$\Pr(X \geq t) \leq \frac{\mathbb{E}[X]}{t}$$

- b. For any $t > 0$,
- c. Example: $X \sim \text{Bin}(1000, 0.1)$. Then $\mathbb{E}[X] = 100$.
- d. The tail probability $\Pr(x \geq 400) \leq \frac{1}{4}$
- e. Proof:

$$\mathbb{E}[X] = \int_0^\infty x f(x) dx = \int_0^t x f(x) dx + \int_t^\infty x f(x) dx \geq \int_t^\infty x f(x) dx \geq t \int_t^\infty f(x) dx = t \Pr(X > t)$$

7.