

# 1 Sample Spaces

**Definition 1.** An experiment is any process that can be repeated with a well-defined set of outcomes. Example: roll a die, toss a coin.

**Sets:** discrete/finite, discrete/infinite, continuous

**Definition 2.** A sample space is the set of all possible outcomes of an experiment. Example:  $S = \{1, 2, 3, 4, 5, 6\}$ ,  $S = \{H, T\}$

**Definition 3.** The union of  $A$  and  $B$ ,  $A \cup B$ , is all elements in  $A$  or  $B$ .

**Definition 4.** The intersection of  $A$  and  $B$ ,  $A \cap B$ , is all elements in both  $A$  and  $B$ .

**Definition 5.**  $A$  and  $B$  are mutually exclusive (disjoint) if  $A \cap B = \emptyset$ .

**Definition 6.** The complement of  $A$ ,  $A^c$ , is the set of elements not in  $A$ :  $S - A$ .

# 2 Probability

**Definition 7.** The function  $P$  is a probability function if it satisfies:

1.  $P(A) \in [0, 1]$  for all  $A \subseteq S$
2.  $P(S) = 1$
3. If  $A$  and  $B$  are disjoint, then  $P(A \cup B) = P(A) + P(B)$
4. If  $A_1, A_2, \dots$  are disjoint, then  $P(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} P(A_i)$

**Lemma 1.**  $P(A^c) = 1 - P(A)$

**Proof:**  $A \cup A^c = S$ ,  $A \cap A^c = \emptyset$ , so  $P(S) = P(A \cup A^c) = P(A) + P(A^c) = 1$ . Therefore,  $P(A^c) = 1 - P(A)$ .

## 2.1 Fair Dice Example

If we roll a fair die,  $P(i) = \frac{1}{6}$  for  $i = 1, 2, \dots, 6$ .

$$P(\{1, 2\} \cup \{3, 4\} \cup \{5, 6\}) = P(\{1\}) + P(\{2\}) + \dots + P(\{6\}) = 6 \cdot P(\{1\}) \Rightarrow P(\{1\}) = \frac{1}{6}$$

Note: If all elements in  $S$  are equally likely,  $P(A) = \frac{\# \text{ in } A}{\# \text{ in } S}$

## 2.2 Card Example

Roll a fair dice, find  $P(\text{sum} = 5)$ .

	$A$	$B$	Total
	1	4	5
	2	3	5
	3	2	5
	4	1	5

There are 26 people, find  $P(\text{at least 2 same birthday})$ .

$1 - P(0 \text{ different})$

**Lemma 2** (Addition Rule).  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

If disjoint:  $P(A \cup B) = P(A) + P(B) - 0 = P(A) + P(B)$

Proof:  $P(A \cup B^c) + P(A \cap B^c) + P(B \cap A^c) = P(A \cup B)$

**Prove:** If  $A \subseteq B$ , then  $P(A) \leq P(B)$

If  $A \subseteq B$ ,  $B^c \cap A = \emptyset$

$B = A \cup (A^c \cap B)$  (disjoint union), so  $P(B) = P(A) + P(A^c \cap B)$

Because  $P(\cdot)$  is non-negative,  $P(B) \geq P(A)$

### 3 Counting

#### 3.1 Example 2.8

Given  $A, B, C$ :

- a) none occur:  $A^c \cap B^c \cap C^c$  (only  $A$  occurs:  $A \cap B^c \cap C^c$ )
- b) all three occur:  $A \cap B \cap C$  (exactly one occurs:  $(A \cap B^c \cap C^c) \cup (A^c \cap B \cap C^c) \cup (A^c \cap B^c \cap C)$ )
- c) at least one occurs:  $A \cup B \cup C$   $((A \cap B \cap C) \cup (A \cap B^c \cap C))$

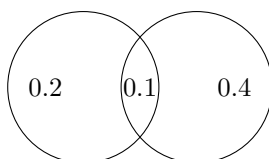
$$P(A) = 0.3, P(B) = 0.7, P(A \cap B) = 0.1, \text{ WANTED}$$

$$P(A \cup B) = 0.3 + 0.7 - P(A \cap B) = 0.9 - 0.1 = 0.8$$

Blue/green/red

$$P(A \cup B) = 0.5 + 0.7 = P(A \cap B) = 0.9$$

$$P(A \cap B) = 0.3$$



#### 3.2 Counting Example

Choose a card at random:  $P(A) = \frac{3}{7}, P(A \cap \text{red}) = \frac{15}{70} = \frac{3}{14}, P(A \cup \text{red}) = \frac{4}{7}$

Choose two at random without replacement (conditional probability): if the first card has an  $A$ , find

$$P(\text{second} = A) = \frac{2}{69}, P(\{2^{\text{nd}} = A\} \mid \{1^{\text{st}} = A\}) = \frac{1}{51} = \frac{1}{17}$$

$$\text{One card: } P(\text{red} \mid A) = \frac{15}{30} = 0.5$$

$$P(\text{red} \mid A) = \frac{P(\text{red} \cap A)}{P(A)} = \frac{3/7}{1} = \frac{3}{7}$$

### 4 Conditional Probability

**Definition 8.**  $P(A \mid B) = \frac{P(A \cap B)}{P(B)}$

Choose two without replacement:  $P(\text{both blue}) = P(\{1^{\text{st}} \text{ blue} \cap 2^{\text{nd}} \text{ blue}\}) = P(\text{blue})P(2^{\text{nd}} \text{ blue} \mid 1^{\text{st}} \text{ blue})$

$$= \frac{14}{70} \cdot \frac{13}{69} = \frac{182}{4830} = \frac{14}{70} = \frac{1}{5}$$

Choose 5 without replacement:  $P(\text{at least one is blue}) = 1 - P(\text{no blue}) = 1 - \frac{56 \cdot 55 \cdots}{70 \cdot 69 \cdots}$

Example: Two buckets - 11 blue/1 green, 2 blue/3 green. Choose one from bucket one and move to bucket 2, choose 1 from bucket 2:

$$P(1^{\text{st}} \text{ blue} \cap 2^{\text{nd}} \text{ blue}) = \text{WORK}$$

$$P(1^{\text{st}} \text{ green} \cap 2^{\text{nd}} \text{ blue}) = \frac{1}{4}$$

$$P(2^{\text{nd}} \text{ blue}) = \frac{6}{12}$$

#### 4.1 Total Probability - Given Events

Total probability: Given  $A_1, A_2, \dots, A_n$  with  $A_i \cap A_j = \emptyset$  ( $i \neq j$ ) and  $\bigcup A_i = S$ ,

$$\text{then } P(B) = \sum_{i=1}^n P(B \cap A_i) = \sum_{i=1}^n P(A_i)P(B \mid A_i)$$

## 5 Conditional Probability (continued)

Find  $P(1^{\text{st}} \text{ blue} \mid 2^{\text{nd}} \text{ blue}) = \frac{P(1^{\text{st}} \text{ blue} \cap 2^{\text{nd}} \text{ blue})}{P(2^{\text{nd}} \text{ blue})} = \frac{4/4}{5/12} = 0.6$

See setup for total probability.

**Theorem 1** (Bayes' Theorem).

$$P(A \mid B) = \frac{P(B \mid A)P(A)}{P(B)} = \frac{P(B \cap A)P(A)}{\sum_{\text{all } k} P(B \mid A_k)P(A_k)}$$

### 5.1 Medical Test Example

A test is 95% accurate:  $P(\{+\} \mid \text{disease}) = 0.95$ ,  $P(\{-\} \mid \text{not disease}) = 0.95$

If  $P(\text{disease}) = 0.01$ , find  $P(\text{test}+)$ :

$$0.01 \cdot 0.95 = (0.01)(0.95) + (1.00)(0.05) = 0.059$$

$$P(\text{disease} \mid +) = \frac{P(+ \mid \text{disease})}{P(+)} = \frac{(0.01)(0.95)}{0.059} = 0.16$$

Roll a die until you get a 6:

$$P(\{3\}) + P(\{4\}) + P(\{5\}) = \left(\frac{5}{6}\right)^2 \left(\frac{1}{6}\right) + \left(\frac{5}{6}\right)^3 \left(\frac{1}{6}\right) + \left(\frac{5}{6}\right)^4 \left(\frac{1}{6}\right)$$

Geometric sequence:  $a + ar + ar^2 + \dots = \frac{a}{1-r}$  if  $|r| < 1$

$$a = \frac{1}{6}, r = \frac{5}{6}, \text{ so } \left(\frac{5}{6}\right)^2 \left(\frac{1}{6}\right) = \frac{a \cdot r^2}{1-r} = \frac{(1/6)(5/6)^2}{1/6} = \frac{25}{36}$$

Question 2 if  $r \geq 1$

$P$  if  $A \subseteq B$ , then  $P(A) \leq P(B)$

## 6 Homework

### 6.1 Roll 2 dice

Find  $P(\text{sum} = 6)$ :  $S' = \{2, 3, 4, \dots, 12\}$ ,  $S_Y = \{0, 1, 2, \dots\}$

Roll 2 dice: Find  $P(\text{exactly 4 green})$

$$n = 10, k = 4, p = \frac{10}{50} = 0.2, P(4 \text{ green}) = \binom{10}{4} (0.2)^4 (0.8)^6$$

Binomial pdf  $(n, p, k)$ :  $n \binom{n}{k} = \frac{n!}{k!(n-k)!}$

Choose 5 without replacement:

### 6.2 Discrete Random Variables

**Example 1.** Roll 2 dice:  $X = \text{sum of two rolls}$ ,  $Y = \# \text{ of 6s}$

$$S_X = \{2, 3, 4, \dots, 12\}, S_Y = \{0, 1, 2\}$$

Example: Roll 1 dice, let  $X = (1^{\text{st}} \text{ roll})$ ,  $Y = \# \text{ of } \{$

$$S_X = \{2, 3, 4, \dots, 12\}, S_Y = \{0, 1, 2, 3\}$$

**Theorem 2.** A random variable  $X$  is a function from a sample set to the real numbers (with an associated probability distribution).

**Example:** Roll two dice:  $X = \text{sum of two rolls}$ ,  $Y = \# \text{ of } \{$   
 $S_X = \{2, 3, 4, \dots, 12\}, S_Y = \{0, 1, 2, 3\}$

**Theorem 3.** Associated with a random variable is a probability density function (pdf) which gives the probability of all elements.

**Theorem 4.** The cumulative distribution function (cdf) for  $X$  is:  $F_X(x) = P(X \leq x)$

## 7 Binomial Distribution

**Definition 9.**  $P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$ ,  $k = 0, 1, 2, \dots, n$

where  $n = \#$  of trials,  $k = \#$  of successes,  $p = P(\text{success in one trial})$ ,  $\binom{n}{k} = \text{number of ways choosing } k \text{ out of } n$

**Example 2.** 50 red, 40 red, 10 green

Choose 10 with replacement: Find  $P(\text{exactly 4 green})$

$$n = 10, k = 4, p = \frac{10}{50} = 0.2, P(4 \text{ green}) = \binom{10}{4} (0.2)^4 (0.8)^6$$

**Binompdf** ( $n, p, k$ ):  $n \binom{n}{k} = \frac{n!}{k!(n-k)!}$

Choose 5 without

## 8 Discrete Random Variables

**Example 3.** Roll 2 dice, find  $P(\text{sum} = 5)$

Choose 10 with replacement: Find  $P(\text{exactly 4 green})$

$$n = 10, k = 4, p = \frac{10}{50}, p = 0.2$$

$$P(4 \text{ green}) = \binom{10}{4} (0.2)^4 (0.8)^6$$

## 9 Continuous Random Variables

**Example 4.** Choose a number at random from the interval  $[0, 100]$  in  $\mathbb{R}$ :

$$P(X \leq 50) = 0.5, P(X \leq 10) = \frac{1}{10}, P(X \leq 1) = \frac{1}{100}, P(X = 0) = 0$$

$$P(X \leq V_1) = \int_0^{V_1} p_{2-x} dx = \frac{x^2}{2} \Big|_0^{V_1} = \frac{V_1}{4}$$

**Example 5.**  $P_X(x) = 2x$ ,  $0 \leq x \leq 1$

$$P(X \leq V_1) = P(X \leq V_1) = \int_0^{V_1}$$

**Definition 10.** For  $P_X(x)$  to be a pdf, we need:  $f_X(x) \geq 0$ ,  $\int_{-\infty}^{\infty} f_X(x) dx = 1$

$$\text{Find the cdf } F_X(x) = P(X \leq x) = \int_{-\infty}^x 2x dx = x^2 \Big|_0^x = x^2 \text{ for } 0 \leq x \leq 1$$

**Example 6.**  $f_X(x) = 2e^{-2x}$ ,  $x \geq 0$

$$P(X \geq 1) \int_1^{\infty} 2e^{-2x} dx = -e^{-2x} \Big|_1^{\infty} = 0 + e^{-2}$$

$$f_X(x) = cx^2 e^{x^3}, x \geq 0$$

Find  $c$  so this is a pdf:

$$1 = \int_0^{\infty} cx^2 e^{x^3} dx = c \left( -\frac{1}{3} e^{-x^3} \right) \Big|_0^{\infty} = \frac{c}{3} (0 - (-1)) = \frac{c}{3}$$

$$c = 3$$

$$P_X(x) = 3x^2 e^{-x^3}$$

Find  $c$  for pdf:  $1 = \int_0^{\infty} cxe^{-2x} dx$

$$u = -\frac{1}{2} e^{-2x}, dv = x dx$$

$$dv = 1, dv = e^{-x}$$

$$c \left[ -\frac{1}{2} x e^{-2x} - \frac{1}{4} e^{-2x} \right] \Big|_0^{\infty} = c \cdot \frac{1}{4} = 1, c = 4$$

$$f_X(x) = 2x, 0 \leq x \leq 1$$

If  $y = 2x + 1$ , find  $P_Y(y)$

$$\text{Go through: } F_Y(y) = P(Y \leq y) = P(2x + 1 \leq y) = P(x \leq \frac{y-1}{2})$$

$$= \int_0^{\frac{y-1}{2}} 2x dx = x^2 \Big|_0^{\frac{y-1}{2}} = \left( \frac{y-1}{2} \right)^2 = \frac{(y-1)^2}{4}, 1 \leq y \leq 3$$

$$f_Y(y) = \frac{d}{dy} (y-1)^2 = \frac{(y-1)}{2} \text{ pdf}$$

## 10 Expected Value

**Example 7.** We have: 3, 2.5, 4, 3.5, 2, 5.5. Mean:  $\frac{7.5}{20} = 3.75$   
 $2\left(\frac{3}{20}\right) + 3\left(\frac{8}{20}\right) + 5\left(\frac{2}{20}\right) = \sum_i n_i P(n_i)$

**Definition 11.** The expected value of a random variable  $X$  is:

$$E(X) = \begin{cases} \sum_{x \in S} x \cdot P(X = x) & \text{discrete} \\ \int_{-\infty}^{\infty} x \cdot f_X(x) dx & \text{continuous} \end{cases}$$

**Example 8.**  $P(X = x) = \frac{1}{10}$ ,  $x = 1, 2, 3, 4$   
 $E(X) = \frac{1}{10}[1 + 2 + 3 + 4] = 3$   
 $P(X = 1) = \frac{1}{10}$

Roll a die until the first 6, with # of rolls: Find  $E(X)$

$$P(X = 1) = \frac{1}{6}, P(X = 2) = \left(\frac{5}{6}\right)\left(\frac{1}{6}\right), P(X = k) = \left(\frac{5}{6}\right)^{k-1}\left(\frac{1}{6}\right), k = 1, 2, \dots$$

$$E(X) = \sum_{k=1}^{\infty} k \cdot \left(\frac{5}{6}\right)^{k-1}\left(\frac{1}{6}\right)$$

$$\sum ar^k = \frac{a}{1-r}, a + 2ar + 3ar^2 + \dots = \frac{a}{(1-r)^2}, E(X) = \frac{1/6}{(1-5/6)^2} = 6$$

Binomial:  $P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$ ,  $k = 0, 1, \dots, n$

$$E(X) = \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k} = np$$

### 10.1 Continuous Example

$$f_X(x) = 2x, 0 \leq x \leq 1$$

$$E(X) = \int_0^1 x(2x^2) dx = \frac{2}{3}$$

$$f_X(x) = 2e^{-2x}, x \geq 0$$

$$E(X) = \int_0^{\infty} x \cdot 2e^{-2x} dx \Rightarrow \hat{\lambda} = \frac{1}{2}$$

Toss a coin until the first  $T$ , if it's on 1st win 1

$$1 \cdot P(T \text{ on 1st}) + 2 \cdot P(T \text{ on 2nd}) + \dots = 2^n E(W_n) = 2\left(\frac{1}{2}\right) + 4\left(\frac{1}{4}\right) + \dots + 2^n\left(\frac{1}{2^n}\right) = \sum_{n=1}^{\infty} 1 = \infty$$

$$f_X(x) = \frac{1}{2}, x \geq 1$$

Show this is a pdf:

$$1 = \int_1^{\infty} \frac{c}{x^2} dx = -\frac{c}{x} \Big|_1^{\infty} = 0 + 1 = 1 \checkmark$$

$$E(X) = \int_1^{\infty} x \left(\frac{1}{x^2}\right) dx = \ln(x) \Big|_1^{\infty} = \infty$$

## 11 Median

**Definition 12.** The median  $m$  is the number so that  $P(X \leq m^*) \geq 0.5$  and  $P(X \geq m^*) \geq 0.5$ .

**Example 9.**  $P_X(x) = 3x^2$ ,  $0 \leq x \leq 1$

$$0.5 = \int_0^{m^*} 3x^2 dx, \text{ so } 0.5 = x^3 \Big|_0^{m^*} = (m^*)^3, \text{ thus } m^* = \sqrt[3]{0.5}$$

**Notation:**  $E(X) = \mu$ ,  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$  (population mean vs sample mean)

**Note 1.**  $E(g(x)) = \sum g(x)P(X = x)$  or  $\int_{-\infty}^{\infty} g(x)P_X(x)dx$

**Example 10.**  $x = [-1, 1, 3]$

$$E(X) = -0.3 + 0.4 + 0.9 = 1$$

$$E(\cos(x)) = \cos(-1)(0.3) + \cos(1)(0.2) + \cos(3)(0.3)$$

$$E(X) = \int_0^1 3x^3 dx, E(x^2) = \int_0^1 x^2(3x^2) dx$$

## 12 Variance

**Definition 13.**

$$\text{Var}(X) = E[(X - \mu)^2] = \sigma^2 \quad 1, 2, 5 \Rightarrow \sigma^2 = 2 \quad \text{Var} = 4$$

**Theorem 5.**  $E[(X - \mu)^2] = \text{Var}(X) = E(X^2) - \mu^2$

**Note 2.**  $E(aX + b) = aE(X) + b$

## 13 Variance Examples

**Example 11** (Finite discrete dist.).  $x = 1, 2, 3, 4$ ,  $E(X) = \frac{1}{4}(1 + 2 + 3 + 4) = 2.5$

$$\text{Var}(X) = \frac{1}{4}[(1 - 2.5)^2 + (2 - 2.5)^2 + (3 - 2.5)^2 + (4 - 2.5)^2] = 1.25$$

**Example 12** (Roll die).  $E(X) = \frac{1}{6}(1 + 2 + 3 + 4 + 5 + 6) = 3.5$

$$E(X^2) = \frac{1}{6}(1 + 4 + 9 + 16 + 25 + 36) = \frac{91}{6}$$

$$\text{Var}(X) = \frac{91}{6} - (3.5)^2 = \frac{35}{12}$$

**Example 13** (Binomial).  $E(X) = np$ ,  $\text{Var}(X) = np(1 - p)$

$$\text{Roll } 600, k \neq \text{of } 6\text{'s}: E(X) = 100, \text{Var}(X) = 600 \cdot \frac{1}{6} \cdot \frac{5}{6}$$

## 14 Further Properties of Mean and Variance

**Theorem 6** (Lemma).  $E(aX + bY) = aE(X) + bE(Y)$

**Useless:**  $\sigma = S(+) + EY + \sqrt{\text{VAR}}$

**Definition 14.**  $\bar{X} = \frac{1}{n} \sum X_i$  where  $X_i$  are from a list with  $E(X_i) = \mu$

$$E(\bar{X}) = \frac{1}{n} \sum E(X_i) = \mu$$

**Note 3.**  $\text{Var}(\bar{X}) = E[(\bar{X} - \mu)^2] = E\left[\left(\frac{1}{n} \sum X_i - \mu\right)^2\right] = E[X^2] - \mu^2$

**Note 4.** If  $X$  and  $Y$  are independent,  $\text{Var}(X) + \text{Var}(Y)$

$\text{Var}(aX + bY)$  if  $X$  and  $Y$  are independent:  $a^2\text{Var}(X) + b^2\text{Var}(Y)$

$\text{Var}(X_1 + \dots + X_n) = \text{Var}(X_1) + \dots + \text{Var}(X_n)$  if  $X$  and  $Y$  are independent

$\text{Var}(\bar{X}) = \sigma^2$  where  $\sigma^2 = \text{Var}(X_i) = \frac{\sigma^2}{n}$

**Notation (if points equally likely):**  $\sigma^2 = \frac{1}{n} \sum (X_i - \mu)^2$  = population variance

$s^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2$  = sample variance

$Y$	1	2	3	4	5	6
$X$	1	1	1	1	1	1
1	×					
2	×	×				
3	×	×	×			
4	×	×	×	×		
5					×	
6						×

PMF with respect to  $Y$

## 15 The k-th Moment

**Definition 15.** The  $k$ -th moment of  $X$  is  $E(X^k)$

**Notation (if points equally likely):**  $\sigma^2 = \frac{1}{n} \sum (X_i - \mu)^2$  = population variance = standard

$s^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2$  = sample var

## 16 Joint Densities

**Example 14.**  $f_{X,Y}(x,y) = \frac{2}{3}xy$ ,  $0 \leq x \leq 1$ ,  $0 \leq y \leq 2$ ,  $x+y > 1$

$$\text{Find } E(XY) = \int \int (xy) \left[ \frac{2}{3}xy \right] dydx = \frac{2}{3} \int_0^1 \int_{1-x}^2 x^2 y^2 dydx$$

$$\text{Find } F(X) = \int_0^2 \frac{2}{3}xydy = \frac{2}{3}xy \Big|_{y=0}^2 = \frac{2}{3}x(2) - \frac{2}{3}x(0) = \frac{4}{3}x, 0 \leq x \leq 1$$

$$\text{Expected } f(x): P(Y > X) = 1 - P(Y \leq X) = 1 - \int_0^1 \int_0^x \frac{2}{3}xydydx$$

$$y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}, z = \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$$

$$\text{Find } E(XY) = \sum \sum (xy)P(X,Y) = (.5)(2)(.5) + 1.5$$

$$E(Y) = 1(.3) + 1(.6) + 2(.1) = .5$$

$$\text{Find } E(XY) = (.1)(.1) + (.2) + 2(.2) + 2(.3) + 2(.2) + 2(.3) = .6$$

**Definition 16.** The covariance is the  $\text{Cov}(X,Y) = E[(X - \mu_X)(Y - \mu_Y)] = E(XY) - E(X)E(Y)$

Are the sets  $\{x = 13\}$  and  $\{y = 13\}$  independent?  $P(\{x = 13\} \cap \{y = 13\}) = P(\{x = 13\})P(\{y = 13\})$   
 $P(\{x = x\})$

**Theorem 7.** The random variables  $X$  and  $Y$  are independent if  $P(X \in A \cap Y \in B) = P(X \in A)P(Y \in B)$  for all sets  $A \in S_X$ ,  $B \in S_Y$

If  $X$  and  $Y$  are independent,  $E(XY) = E(X)E(Y) \Rightarrow \text{Cov}(X,Y) = 0$

**Lemma:** The continuous random variables  $X$  and  $Y$  are independent if  $f_{XY}(x,y) = g(x)h(y)$  where  $g(x) = kf_X(x)$  and  $h(y) = \ell f_Y(y)$  for all  $x \in S_X$ ,  $y \in S_Y$

**Example 15.**  $f_{XY}(x,y) = \frac{1}{33}xy$ ,  $0 \leq x \leq 1$ ,  $0 \leq y \leq 2$ ,  $x+y > 1$

Are  $X$  and  $Y$  independent? NO - need 5 to be rectangle

**Example 16.**  $f_{XY}(x,y) = d(x \cdot y)$ ,  $0 \leq x \leq 2$ ,  $0 \leq y \leq 5$

$$f_{XY}(x,y) = \frac{1}{(x+y)}p \neq h(x)g(y) \text{ NO}$$

**Example 17.**  $f_{XY}(x,y) = e^{-(x+y)}$ ,  $x \geq 0$ ,  $y \geq 0$

$$= e^{-x} \cdot e^{-y} = g(x)h(y) \text{ YES}$$

**Example 18.** If  $\text{Var } f_X(x) = 4x^3$ ,  $0 \leq x \leq 1$ ,  $f_Y(y) = 3y^2$ ,  $0 \leq y \leq 1$

Find  $P(Y > X)$ : If  $Y$  and  $X$  are independent

$$f_{XY}(x,y) = 4x^3(3y^2) \Rightarrow$$

$$\int_0^1 \int_x^1 (2x^3y^2)dydx = \int_0^1 4x^3y^3 \Big|_x^1 dx$$

...

## 17 Further Properties of Mean and Variance

**Lemma 3.**  $E(aX + bY) = aE(X) + bE(Y)$

**Definition 17.**  $\bar{X} = \frac{1}{n} \sum X_i$  where  $X_i$  are from a list with  $E(X_i) = \mu$

$$E(\bar{X}) = \frac{1}{n} \sum E(X_i) = \mu$$

$$\text{Note: } \text{Var}(\bar{X}) = E[(\bar{X} - \mu)^2] = E[(X_i - \mu)^2] = E[X^2] - \mu^2$$

**Note 5.** If  $X$  and  $Y$  are independent,  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$

$$\text{Var}(aX + bY) \text{ if } X \text{ and } Y \text{ are independent} = a^2\text{Var}(X) + b^2\text{Var}(Y)$$

$$\text{Var}(\bar{X}) \text{ for } X \text{ and } Y \text{ independent: } \sigma^2 = \text{Var}(X_i) = \frac{1}{n^2}[\sigma^2 + \dots + \sigma^2] = \frac{\sigma^2}{n}$$

$$\text{Var}(N) = 100\sigma^2(X) - 36\text{Var}(Y) - 110\sigma^2(X) + 36\text{Var}(Y) \text{ if } X \text{ and } Y \text{ are ind}$$

**Definition 18.** The correlation coefficient is:  $\rho = \frac{\text{Cov}(X,Y)}{\sigma_X \sigma_Y}$ ,  $-1 \leq \rho \leq 1$

PPP:

1.  $-1 \leq \rho \leq 1$
2.  $\rho = \pm 1 \Leftrightarrow y = a + bx$
3.  $\rho = 0$  no linear association
4.  $\rho$  gives the percent of the variation in  $y$  due to the change in  $x$

**Example 19** (20).  $X \sim \text{Binom}(n, p_x)$ ,  $Y \sim \text{Binom}(m, p_y)$ ,  $U = X + Y$

Find  $E(U)$ ,  $\text{Var}(U)$

$E(X) = np_x$ ,  $E(Y) = mp_y$ ,  $\text{Var}(X) = np_x(1 - p_x)$ ,  $\text{Var}(Y) = mp_y(1 - p_y)$

$E(U) = 4E(X) + 6E(Y) = 4np_x + 6mp_y$

$\text{Var}(U) = 10\sigma^2(X) - 36\text{Var}(Y) = 110p_x(1 - p_x) + 36mp_y(1 - p_y)$  if  $X$  and  $Y$  are ind

## 18 The Poisson Distribution

**Definition 19.** The Poisson distribution is:  $P(X = k) = \frac{e^{-\lambda} \lambda^k}{k!}$  for  $k \in \mathbb{N}$  (or  $k = 0, 1, \dots$ )  
 $E(X) = \lambda = \text{Var}(X)$

This usually gives the number of occurrences in a time interval. Something is Poisson if:

1. The occurrences in  $\Delta t$  happen simultaneously
2. For a short enough time frame, there are 0 or 1 occurrences
3. Occurrences are independent
4. The rate of occurrences is constant (only matters well if a  $\Delta t$ -age unit 2 small)

**Example 20.** A TV has 6 million pixels, where the probability that one is defective is  $\frac{1}{2}$  mill.  $\lambda = \text{mean} = 5$  A mill/2 mill = 2.5

$P(3 \text{ are defective}) \approx e^{-2.5} \frac{(2.5)^3}{3!} = P(\text{Poisson}(\lambda = 2.5, k)) = 0.2136$

Find  $P(X \geq 3)$ : WORK = Poisson  $p(\lambda = 2.5, k)$

**Example 21.** The number of people who walk into a store is Poisson with a mean of  $\frac{50}{1 \text{ hour}}$ . Find  $P(\text{at least in one hour with Poisson}(\lambda = 2.5, k))$  or  $P(3 \text{ in } 10 \text{ mins})$

$\lambda = 3$ , Poisson  $p(\lambda = 5, k) = 0.104$

**Note:** If the number of occurrences is Poisson, then the time between occurrences is exponential:  $P(X = k) = \lambda e^{-\lambda k}$  (ECN  $\Rightarrow$  X)

**Example 22.** The mean is 492 for two hours:  $\lambda = 492$  (for 2 hours)

Find  $\lambda$  (in 2 min):  $\lambda = \frac{492}{60} \approx 8.0333$ ,  $\lambda^{492}/60 = 8.0333$

Therefore Poisson  $p(\lambda = 492/60, k) = 0.029$  or  $P(E3; \text{ in } 2 \text{ min}) = \text{Poisson } p(\lambda = 492/60, k) = 0.014$

Find  $P(E3; \text{ in } 2 \text{ min}) = \text{Poisson } p(\lambda = 492/60, k)$

**Example 23.** 20000 People play, if  $P(\text{win}) = \frac{1}{5000}$ . Find  $P(2 > \text{People win})$

$\lambda = \frac{20000}{5000} = 4$ ,  $1 - P(X \leq 3) = 1 - \text{PoissonCDF}(4, 2) = \bar{7}(2)$



## 19 The Normal/Gaussian Distribution

**Definition 20.** Normal/Gaussian distribution:  $f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$ ,  $-\infty < x < \infty$ ,  $-\infty < \mu < \infty$ ,  $\sigma > 0$

Standard:  $f_Z(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$ , human

**Example 24.** Find  $P(2 \leq (1.2)) = \text{Normal CDF}(2, 100000, 1.2) \cap \mu = 0 \Rightarrow P = 0.8849$

**Lemma:** If  $X$  is Normal with  $\mu = 100$ ,  $\sigma = 10$

Find  $P(2 \leq (1.2))$  If we take a sample of 100  $P(\bar{X} > 52) = \text{Normal CDF}(52, 100000, 50)$

Find a so:  $P(\mu = k \leq \mu) \Rightarrow \text{inv Normal CDF}(.75, 100, 10, \text{center})$

HW 3

Bernoulli:  $P(X = 1) = p$ ,  $P(X = 0) = 1 - p$

a) Find  $E(X)$  and  $\text{Var}(X)$

$$E(X) = \sum xP(X = x) \text{ and } \text{Var}(X)$$

b) Let  $Y = X_1 + X_2 + \dots + X_n$  when  $X_i = \text{Bernoulli}$  with  $P(X_i = 1) = p$  and

$$E(Y) = \sum xP(X = x)$$

$$\text{Var}(Y) = \text{Var}(X_1 + \dots) = np(1 - p)$$

$$\text{Var}(Y) = \text{Var}(2X_i) \Rightarrow \text{Var}(\bar{X}_i)^2$$

## 20 Central Limit Theorem

Given a sample  $X_1, X_2, X_3, \dots, X_n$  all from the same population taken independently, then  $\bar{X} = \frac{1}{n} \sum X_i$  and the sum  $\sum X_i$  are normal

No outliers or strong skewed normal converges under sample  $E(\frac{\bar{X}-\mu}{\sigma/\sqrt{n}}) \Rightarrow$  so  $sd(\bar{X}) = \frac{\sigma}{\sqrt{n}}$ ,  $\text{Var}(\bar{X}) = n\sigma^2$

For  $X$  we know  $\mu = 50$ ,  $\sigma = 8$

Find  $P(\bar{X} > 52) = ?$  If we take a sample of 100  $P(\bar{X} > 52) = \text{Normal CDF}(52, 10000, 50, 30)$

Find  $P(X > 52) = ?$  Find a  $P(\mu \leq k \leq \mu) \Rightarrow a$  with  $P(X \geq 52) = \text{Normal CDF}(52, 10000, 50, 30) = 0.0062$

Find a so that  $P(\mu \leq k \leq \mu)$ : inv Normal (1.75, 100, 10, center)

**Example 25.**  $X$  is normal with  $\mu = 12$ ,  $\sigma = 42$

Find  $P(X \leq 150) = \text{Normal CDF}(10000, 130, 12, 41) = 0.0655$

With a sample of 400: Find  $P(10 \leq \bar{X} \leq 1(2)) = \text{Normal CDF}(112, 110, 112, 41/\sqrt{400})$

**Example 26.** Roll a dice 600 times, estimate  $P(90 \leq \# \text{ of } 6\text{'s} \leq 115)$ : Mean =  $np = 106$

variance =  $np(1 - p) = \frac{500}{6} \approx 71.27$

continuity correction:  $P(\bar{X} < 70) \Rightarrow P(\bar{X} \leq 70.5) \Rightarrow$

$P(10 \leq k \leq 115) = P(90.52k \leq 115) = \text{Normal CDF}(115, 90.5, 115.5, 70.5, 100.5)$

$k < 100 \Rightarrow X \leq 100.5$ ,  $X \geq 100 \Rightarrow X \geq 99.5$