#### Section 3.1 - Random Variable

**Definition 1** (Random Variable). A random variable is a function that associates a real number with each element in the sample space.

A random variable is used when you know the possible values it could have, but not what value it will actually take.

Example 1. Roll a 6-sided die. What values can it take on? What value will come up? Define a random variable for this scenario.

**Definition 2** (Discrete Random Variable). A random variable whose image is countable.

Example 2. We wait in line to see a movie. We know that the waiting time in line can be anywhere from 0 hours to maybe 3 hours. However, we don't know how long the wait is actually going to be. Define a random variable for this scenario.

**Definition 3** (Continuous Random Variable). A random variable whose image is an interval and whose CDF is continuous.

# Section 3.2 - Probability Mass Function (PMF)

**Definition 4** (Probability Distribution (Discrete Random Variable)). Writing down a formula or a table to represent all of the probabilities of the values of the random variable.

**Definition 5** (Probability Mass Function (PMF)). A function, f(x), associated with a random variable X.

PMF Properties:

Example 3 (Probability Calculations).

Toss a coin 2 times. Let X be a random variable representing the number of heads that occur.

- (a) Identify the sample space S.
- (b) Write the PMF for X.

- (c) Find  $P(X \le 1)$ .
- (d) Find P(X < 1).
- (e) Find  $P(X \le 2)$ .
- (f) Find P(X < 2).
- (g) Find  $P(X \ge 1)$ .
- (h) Find P(X > 1).

Example 4 (Valid PMF Check).

Suppose we know

$$f(x) = c(x+5), \ x = 1, 2, 3.$$

What value of c would make this a valid pmf?

Example 5. Find a formula for the probability distribution of the random variable (X) representing the outcome when a single die is rolled once.

# Section 4.1 - Expected Value

Example 6 (Expected Value Intro Example).

You play a game where the roll of 1 fair die determine whether you win or lose.

- If you roll a 5 or 6, you win \$5.
- If you roll a 3 or 4, you win \$0 (breakeven).
- If you roll a 1 or 2, you lose \$2.

How much do you win/lose on average, if you play a really long time?

**Definition 6** (Expected Value (General Formula)). Given a function g(x) from  $\mathbb{R}$  to  $\mathbb{R}$ , we define

$$\mu_{g(X)} = \mathbf{E}[g(X)] = \sum_{x} g(x)f(x)$$

Example 7. Suppose f(x), the pmf for a discrete random variable X is given in the table below:

$$\begin{array}{c|cccc} x & 5 & 10 & 15 \\ \hline f(x) & 0.50 & 0.10 & 0.40 \\ \end{array}$$

(a) What is P(X = 5)?

(c) What is P(X > 8)?

(b) What is P(X = 6)?

(d) What is P(X > 10)?

- (e) Compute  $\mu$ .
- (f) Compute  $\sigma$ .

#### Section 4.2 - Variance and Standard Deviation

**Definition 7** (Variance). The variance of a discrete random variable X is denoted by  $\sigma^2$  or  $\mathbf{V}(X)$ .

$$\sigma^2 = \mathbf{V}(X) = \mathbf{E}\left[ (X - \mu_X)^2 \right] = \sum_x (x - \mu_X)^2 f(x) = \mathbf{E}(X^2) - [\mathbf{E}(X)]^2$$

**Definition 8** (Standard Deviation). The standard deviation of X is denoted by  $\sigma$  or SD(X). The standard deviation of X is the square root of the variance of X:

$$\sigma = \mathbf{SD}(X) = \sqrt{\mathbf{V}(X)} = \sqrt{\sigma^2}$$

Example 8 (Citrus Farmer: Expected Value / Variance / Standard Deviation). A citrus farmer observed the following distribution for X, the number of oranges per tree.

(a) What is the expected value of X?

(b) What is the variance of X?

**Definition 9** (Variance (General Formula)). The variance of a function of X, say g(X) is given by

$$\sigma_{g(X)}^2 = \mathbf{V}[g(X)] = \mathbf{E}\left\{ \left[ g(X) - \mu_{g(X)} \right]^2 \right\} = \sum_x \left[ g(x) - \mu_{g(X)} \right]^2 f(x).$$

Example 9 (Example 8 continued).

(a) What is the expected value of g(X) = 2X + 1?

(b) What is the variance of g(X) = 2X + 1?

### Section 4.3 - Linear Combinations of Random Variables

**Theorem 1** (Expected Value Linear Combinations). Let a, b be constants. Let X be a random variable. Let g and h be functions of X. Then

- 1.  $\mathbf{E}(aX) = a \mathbf{E}(X)$
- 2.  $\mathbf{E}(b) = b$
- 3.  $\mathbf{E}(aX \pm b) = a\mathbf{E}(X) \pm b$
- 4.  $\mathbf{E}[g(X) \pm h(X)] = \mathbf{E}[g(X)] \pm \mathbf{E}[h(X)]$

**Theorem 2** (Variance Linear Combinations). Let a, b be constants. Let X be a random variable.

- 1.  $V(aX) = a^2 V(X)$
- 2.  $\mathbf{V}(X \pm b) = \mathbf{V}(X)$
- 3.  $\mathbf{V}(aX \pm b) = a^2 \mathbf{V}(X)$

Example 10 (Example 8 continued).

(a) What is the expected value of g(X) = 2X + 1? Use the shorter formula.

(b) What is the variance of g(X) = 2X + 1? Use the shorter formula.

(c) Find the standard deviation of g(X) = 2X + 1.

### Section 4.3 Proofs: Theorem 1 (#1 and 2 during lecture; #3 and 4 read on own)

1. Proof of  $\mathbf{E}(aX) = a\mathbf{E}(X)$ .

$$\mathbf{E}(aX) = \sum_{x} axf(x); \text{ definition } \mathbf{E}(X)$$
$$= a \sum_{x} xf(x); \text{ } a \text{ is a constant}$$
$$= a \mathbf{E}(X); \text{ definition } \mathbf{E}(X)$$

2. Proof of  $\mathbf{E}(b) = b$ .

$$\mathbf{E}(b) = \sum_{x} bf(x); \text{ definition } \mathbf{E}(X)$$

$$= b \sum_{x} f(x); b \text{ is a constant}$$

$$= b \text{ since } \sum_{x} f(x) = 1$$

3. Proof of  $\mathbf{E}(aX \pm b) = a \mathbf{E}(X) \pm b$ .

$$\mathbf{E}(aX \pm b) = \sum_{x} (ax \pm b) f(x); \text{ definition } \mathbf{E}(X)$$

$$= \sum_{x} (axf(x) \pm bf(x)); \text{ distribute}$$

$$= \sum_{x} axf(x) \pm \sum_{x} bf(x); \text{ split into 2 sums}$$

$$= \mathbf{E}(aX) \pm \mathbf{E}(b); \text{ see parts 1 and 2}$$

$$= a \mathbf{E}(X) \pm b; \text{ see parts 1 and 2}$$

4. Proof of  $\mathbf{E}[g(X) \pm h(X)] = \mathbf{E}(g(X)) \pm \mathbf{E}(h(X))$ .

$$\mathbf{E}\left[g(X) \pm h(X)\right] = \sum_{x} \left(g(x) \pm h(x)\right) f(x); \text{ definition } \mathbf{E}(X)$$

$$= \sum_{x} \left[g(x)f(x) \pm h(x)f(x)\right]; \text{ distribute}$$

$$= \sum_{x} g(x)f(x) \pm \sum_{x} h(x)f(x); \text{ split into 2 sums}$$

$$= \mathbf{E}(g(X)) \pm \mathbf{E}(h(X)); \text{ definition } \mathbf{E}(X)$$

Section 4.3 Proofs: Theorem 2 (#1 during lecture; #2 and 3 read on own) Recall  $\mathbf{V}(X) = \mathbf{E}(X^2) - [\mathbf{E}(X)]^2$ .

1. Proof of  $\mathbf{V}(aX) = a^2 \mathbf{V}(X)$ .

$$\mathbf{V}(aX) = \mathbf{E}\left[(aX)^2\right] - \left[\mathbf{E}(aX)\right]^2; \text{ definition of variance}$$

$$= \mathbf{E}\left(a^2X^2\right) - \left[a\mathbf{E}(X)\right]^2; \text{ multiply and pull out constant}$$

$$= a^2\mathbf{E}(X^2) - a^2\left[\mathbf{E}(X)\right]^2; \text{ pull out constant and multiply}$$

$$= a^2\left[\mathbf{E}(X^2) - \left[\mathbf{E}(X)\right]^2\right]; \text{ factor out } a^2$$

$$= a^2\mathbf{V}(X); \text{ definition of variance}$$

2. Proof of  $\mathbf{V}(X \pm b) = \mathbf{V}(X)$ .

$$\mathbf{V}(X \pm b) = \sum_{x} \left[ (x \pm b) - \mathbf{E}(X \pm b) \right]^{2} f(x); \text{ definition of variance}$$

$$= \sum_{x} \left[ x \pm b - \mathbf{E}(X) \mp b \right]^{2} f(x); \text{ property of expected value}$$

$$= \sum_{x} \left[ x - \mathbf{E}(X) \right]^{2} f(x)$$

$$= \mathbf{V}(X); \text{ definition of variance}$$

3. Proof of  $\mathbf{V}(aX \pm b) = a^2 \mathbf{V}(X)$ .

$$\mathbf{V}(aX \pm b) = \sum_{x} [ax \pm b - \mathbf{E}(aX \pm b)]^{2} f(x); \text{ definition of variance}$$

$$= \sum_{x} [ax \pm b - a \mathbf{E}(X) \mp b]^{2} f(x); \text{ property of expected value}$$

$$= \sum_{x} [ax - a \mathbf{E}(X)]^{2} f(x)$$

$$= \mathbf{V}(aX); \text{ definition of variance}$$

$$= a^{2} \mathbf{V}(X); \text{ see part 1}$$

Example 11 (YOU TRY).

Let X be a random variable with probability distribution

$$\begin{array}{c|ccccc} x & 0 & 1 & 3 \\ \hline f(x) & 1/3 & 1/2 & 1/6 \\ \end{array}$$

- (a) What is P(X = 2)?
- (b) What is P(X < 2)?
- (c) What is  $\mathbf{E}(X)$ ?
- (d) What is  $\mathbf{V}(X)$ ?

(e) What is  $\mathbf{E}(Y)$ , where  $Y = (X - 1)^2$ ?

Example R code for Expected Value and Variance from Distributions

Section 3.2 - Cumulative Distribution Function (CDF)

Example 12. Suppose we have the following probability distribution for X:

Draw a graph of the PMF and the CDF of X. Write down the CDF as a piecewise function.

**Definition 10** (Cumulative Distribution Function (CDF)). The cumulative distribution function, or CDF, of a discrete random variable X is defined as

$$F(x) = \mathsf{P}(X \le x) = \sum_{t < x} f(t), \ \ for - \infty < x < \infty.$$

- F(x) is defined for all real numbers.
- F(x) is a step function for discrete random variables with steps at each point of non-zero probability.
- The height of a step at each point of non-zero probability is the probability of that point.
- $\bullet \lim_{x \to -\infty} F(x) = 0$

$$\bullet \lim_{x \to +\infty} F(x) = 1$$

- $\lim_{x \downarrow x_0} F(x) = F(x_0)$  (F is right continuous)
- $P(a < X \le b) = F(b) F(a)$  for a < b

Example 13. Suppose we have the following probability distribution for X:

- (a) Find P(X = 2).
- (b) Find  $P(X \le 2)$ .
- (c) Find P(X < 2).
- (d) Find P(X > 3).
- (e) Find  $P(2 \le X \le 4)$ .

Example 14. Suppose f(x) = cx for x = 1, 2, 3, 4 is a pmf for a random variable X.

- (a) Find c.
- (b) Find the probability that X is even.
- (c) Determine the CDF, F(x). Write your answer as a piecewise function.
- (d) Graph the CDF, F(x).

Example 15. An insurance company offers its policyholders a number of different premium payment options. For a randomly selected policyholder, let X be the number of months between successive payments. The CDF of X is

$$F(x) = \begin{cases} 0, & \text{if } x < 1, \\ 0.4, & \text{if } 1 \le x < 3, \\ 0.6, & \text{if } 3 \le x < 5, \\ 0.8, & \text{if } 5 \le x < 7, \\ 1.0, & \text{if } x \ge 7. \end{cases}$$

Note: When F(x) is NOT based on functions of x (like above), it tells you that X is a discrete random variable.

(a) What is the PMF of X?

(b) Compute  $P(4 < X \le 7)$ .

(c) Compute  $P(4 \le X \le 7)$ .

(d) Compute  $P(3 \le X \le 7)$ .

Example 16 (YOU TRY). Suppose  $S = \{1, 2, 6, 9, 10\}$  for the random variable X. How would you find the following? What if the probabilities were 0.1, 0.2, 0.2, 0.1, 0.4, respectively?

(a) 
$$P(X \le 2)$$

(e) 
$$P(6 \le X \le 10)$$

(b) 
$$P(X < 2)$$

(f) 
$$P(6 < X < 10)$$

(c) 
$$P(X \ge 6)$$

(g) 
$$P(6 < X \le 10)$$

(d) 
$$P(X > 6)$$

(h) 
$$P(6 \le X < 10)$$

Example 17 (YOU TRY). Suppose you are given the following CDF. Find the PMF. Write your answer in table format.

$$F(x) = \begin{cases} 0, & x < 3 \\ 0.4, & 3 \le x < 6 \\ 0.45, & 6 \le x < 9 \\ 0.6, & 9 \le x < 10 \\ 0.98, & 10 \le x < 14 \\ 1, & 14 \le x \end{cases}$$

Example 18 (YOU TRY). A company is considering an investment that will earn \$10,000 with a probability of 0.65, lose \$5,000 with a probability of 0.30, and earn \$100,000 with probability 0.05. Compute the expected value and standard deviation of the amount earned by the investment.

Example 19 (YOU TRY). An overseas shipment of 5 foreign automobiles contains 2 that have slight paint blemishes. If an agency receives 3 of these automobiles at random, list the elements of the sample space S, using the letters B and N for blemished and non-blemished, respectively, then to each sample point, assign a value x of the random variable X representing the number of automobiles with paint blemishes purchased by the agency.

Example 20 (YOU TRY). Determine the value of c so that  $f(x) = cx^2$  for x = 1, 2, 3 is a probability distribution of the discrete random variable X.

Example 21 (YOU TRY). Suppose that the probabilities are 0.4, 0.3, 0.2, and 0.1, respectively, that 0, 1, 2, or 3 power failures will strike a certain subdivision in any given year. Find the mean and variance of the random variable X representing the number of power failures striking this subdivision.

Example 22 (YOU TRY - HARD). A salesperson has 2 appointments on a given day. At the first appointment, the salesperson believes that he/she has a 70% chance to make the deal, earning a \$1000 commission. At the second appointment, the salesperson believes that he/she has a 40% chance to make the deal, earning a \$1500 commission. Assume the appointments are independent. What is the expected commission?