

# Probability Refresher

Univariate

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# Introduction

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- MIT OCW MIT RES.6-012 Introduction to Probability, Spring 2018. (<https://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-041sc-probabilistic-systems-analysis-and-applied-probability-fall-2013/unit-i/lecture-1-introduction-to-probability/>)

## Sample Space

- A sample space is a set of all possible outcomes of an experiment.
- Typically denoted by  $\Omega$ .
- For example, if we toss a coin, the sample space is  $\Omega = \{H, T\}$ .
- If we toss two coins, the sample space is  $\{HH, HT, TH, TT\}$ .
- If we toss a coin and roll a die, the sample space is  $\{H1, H2, H3, H4, H5, H6, T1, T2, T3, T4, T5, T6\}$ .
- Continuous sample spaces are also possible. For example, if we measure the height of a person, the sample space is  $\mathbb{R}$ .

## Sample Space

Consider two rolls of a die. What is the sample space?

X/Y	1	2	3	4	5	6
1	11	12	13	14	15	16
2	21	22	23	24	25	26
3	31	32	33	34	35	36
4	41	42	43	44	45	46
5	51	52	53	54	55	56
6	61	62	63	64	65	66

## Sample Space

Consider you are throwing a dart at a dartboard (square from  $(0, 0)$  to  $(1, 1)$ ). What is the sample space?



The sample space is  $\Omega = \{(x, y) \in \mathbb{R}^2 : 0 \leq x \leq 1, 0 \leq y \leq 1\}$ .

# Events

- An event is a subset of the sample space.
- For example, if we toss a coin, the sample space is  $\Omega = \{H, T\}$ . The event that the coin lands heads is  $A = \{H\}$ .
- If we toss two coins, the sample space is  $\{HH, HT, TH, TT\}$ . The event that the first coin lands heads is  $A = \{HH, HT\}$ .
- If we toss a coin and roll a die, the sample space is  $\{H1, H2, H3, H4, H5, H6, T1, T2, T3, T4, T5, T6\}$ . The event that the coin lands heads is  $A = \{H1, H2, H3, H4, H5, H6\}$ .
- If we measure the height of a person, the sample space is  $\mathbb{R}$ . The event that the height is greater than 6 feet is  $A = \{x \in \mathbb{R} : x > 6\}$ .

# Axioms of Probability

1.  $P(A) \geq 0$  for all events  $A$ .
2.  $P(\Omega) = 1$ .
3. If  $A_1, A_2, \dots$  are disjoint events, then
$$P(\cup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} P(A_i).$$



## Consequences of Axioms

1.  $P(\emptyset) = 0$ .
2.  $P(A^c) = 1 - P(A)$ .
3.  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ .
4.  $P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(A \cap C) - P(B \cap C) + P(A \cap B \cap C)$ .

# Conditional Probability

- The conditional probability of  $A$  given  $B$  is defined as
$$P(A|B) = \frac{P(A \cap B)}{P(B)}.$$
- If  $A$  and  $B$  are independent, then  $P(A|B) = P(A)$ .
- If  $A$  and  $B$  are independent, then  $P(A \cap B) = P(A)P(B)$ .

# Random Variables

- A random variable is a function from the sample space to the real numbers, i.e.,  $X : \Omega \rightarrow \mathbb{R}$ .
- For example, if we toss a coin, the sample space is  $\Omega = \{H, T\}$ .
  - We can define a random variable  $X$  as  $X(H) = 1$  and  $X(T) = 0$ .
  - We could also have a random variable  $Y$  to denote our gain from the coin toss, i.e.,  $Y(H) = 1$  and  $Y(T) = -1$ .
- If we toss two coins, the sample space is  $\{HH, HT, TH, TT\}$ .
  - We can define a random variable  $X$  as  $X(HH) = 2$ ,  $X(HT) = 1$ ,  $X(TH) = 1$ , and  $X(TT) = 0$ , where  $X$  denotes the number of heads.

## Random Variables Notation

$X$  denotes a random variable.  $x$  denotes a particular value of the random variable.

# Probability Mass Function

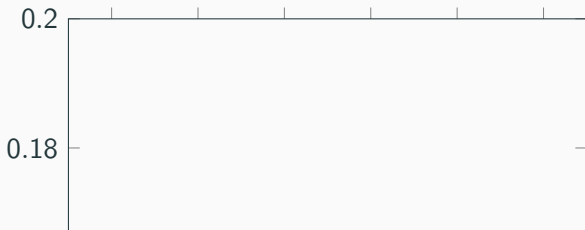
- The probability mass function (PMF) of a discrete random variable  $X$  is defined as  $p_X(x) = P(X = x)$ .
- For example, if we toss a coin, the sample space is  $\Omega = \{H, T\}$ .
  - We can define a random variable  $X$  as  $X(H) = 1$  and  $X(T) = 0$ .
  - The PMF of  $X$  is  $p_X(1) = P(X = 1) = P(H) = 0.5$  and  $p_X(0) = P(X = 0) = P(T) = 0.5$ .
- If we toss two coins, the sample space is  $\{HH, HT, TH, TT\}$ .
  - We can define a random variable  $X$  as  $X(HH) = 2$ ,  $X(HT) = 1$ ,  $X(TH) = 1$ , and  $X(TT) = 0$ , where  $X$  denotes the number of heads.
  - The PMF of  $X$  is  $p_X(2) = P(X = 2) = P(HH) = 0.25$ ,  $p_X(1) = P(X = 1) = P(HT) + P(TH) = 0.5$ , and  $p_X(0) = P(X = 0) = P(TT) = 0.25$ .

# Probability Mass Function

Consider two rolls of die. The following is the sample space.

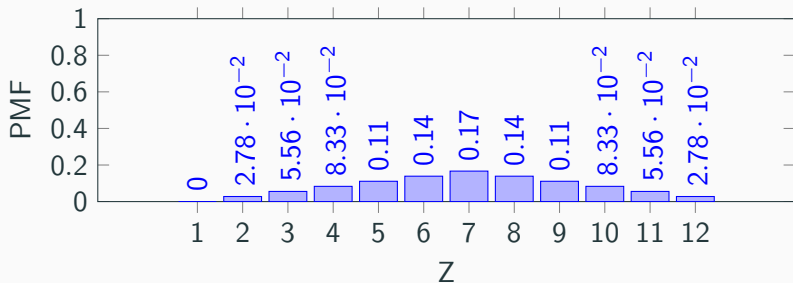
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- X and Y are random variables (value of the 1st and 2nd die)
- $p_X(1) = P(X = 1) = P(11) + P(21) + P(31) + P(41) + P(51) + P(61) = \frac{1}{6}$



# Probability Mass Function

- Let us create a new random variable  $Z = X + Y$ .
  - $p_Z(1) = 0$  as there is no way to get a sum of 1 from two die rolls.
  - $p_Z(2) = P(Z = 2) = P(11) = \frac{1}{36}$

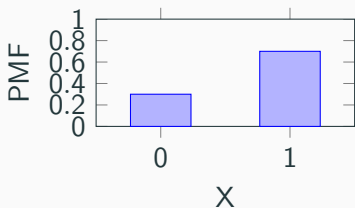


# Discrete Random Variables

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## Bernoulli Distribution

- A random variable  $X$  is said to follow a Bernoulli distribution with parameter  $p$  if  $X$  can take only two values, 0 and 1, and  $P(X = 1) = p$ .
- The PMF of  $X$  is  $p_X(x) = p^x(1 - p)^{1-x}$ .
- For example, if we toss a coin, the sample space is  $\Omega = \{H, T\}$ .
  - We can define a random variable  $X$  as  $X(H) = 1$  and  $X(T) = 0$ .
  - The PMF of  $X$  is  $p_X(1) = P(X = 1) = P(H) = p$  and  $p_X(0) = P(X = 0) = P(T) = 1 - p$ .





## Expected Value

- The expected value of a random variable  $X$  is defined as  $E[X] = \sum_{x \in \mathcal{X}} x \cdot p_X(x)$ .
- For example, consider a random variable  $X$  that follows a Bernoulli distribution with parameter  $p$ .
  - The expected value of  $X$  is  $E[X] = 0 \times (1 - p) + 1 \times p = p$ .
- Consider a random variable  $X$  that follows a uniform distribution over the set  $\{1, 2, 3, 4, 5, 6\}$ .
- The expected value of  $X$  is  $E[X] = \frac{1}{6} \times 1 + \frac{1}{6} \times 2 + \frac{1}{6} \times 3 + \frac{1}{6} \times 4 + \frac{1}{6} \times 5 + \frac{1}{6} \times 6 = 3.5$ .