
Probability Basics

Part 1

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Why Study Probability?



- | **Measure the uncertainty**
- | **“Mathematics is the logic of certainty; probability is the logic of uncertainty”**
 - Introduction to probability by Joseph K Blitzstein and Jessica Hwang

Probability Applications



| Biology

- Gene inheritance

| Medicine

- Medical treatment success rate

| Computer Science

- AI, Machine Learning

| Physics

- Quantum Physics

| Finance

- Risk in investment

Definitions

- | A **sample space** S is the set of all possible outcomes of an experiment.
- | An **event** $A \subseteq S$ is a subset of a sample space.
- | Assuming that all outcomes are equally likely and the space is finite:

$$P(A) = \frac{\# \text{ favorable outcomes}}{\# \text{ possible outcomes}}$$

is the **probability** that A occurs.

Definitions

$$P(\bar{A}) = 1 - P(A)$$

| If $A \subseteq B$, then $P(A) \leq P(B)$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

| Events A and B are independent iff

$$P(\underline{A \cap B}) = P(A)P(B)$$

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

| Probability that at least one of the A_i occurs:

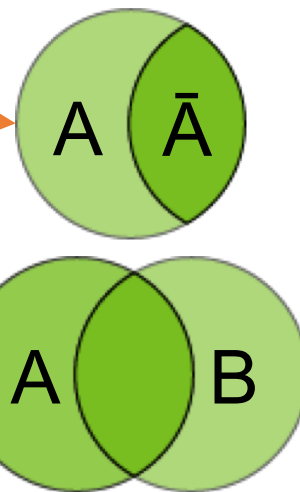
$$P(A_1 \cup \dots \cup A_n)$$

| Probability that none of the A_i occurs:

$$1 - P(A_1 \cup \dots \cup A_n)$$

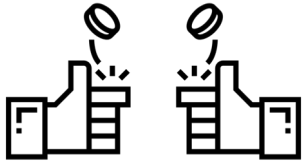
| Probability that all of the A_i occurs:

$$P(A_1 \cap \dots \cap A_n) = 1 - P(\bar{A}_1 \cup \dots \cup \bar{A}_n)$$



Example

Suppose we toss two coins. What is the probability that we see *two heads*? [Each coin has two possible outcomes: H (heads) and T (tails).]

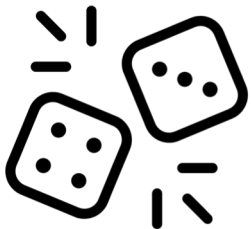


$\{(h, h), (t, t), (h, t), (t, h)\}$ $\frac{1}{4}$

What is the probability that we see at least one tail?

$\frac{3}{4}$

We have two dice that we roll them at the same time. What is the probability that we do not see any 6s?



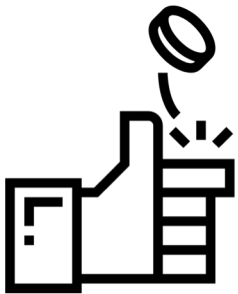
$1 - \left(\frac{1}{6} * \frac{1}{6}\right)$

Discrete Random Variables

Assuming the probability of head is 0.5, how many heads would you get if you toss a coin:

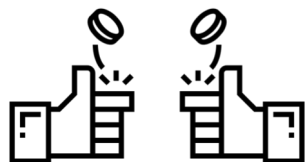
- 4 times $\{0, \dots, 4\}$
- 20 times $\{0, \dots, 20\}$
- 100 times $\{0, \dots, 100\}$

X is a **random variable (RV)** with **arity** of k if it can get exactly one value out of $\{x_1, \dots, x_k\}$



Probability of Discrete RV

| We toss a fair coin 2 times...



$$\begin{aligned}P(0 \text{ heads}) &= \frac{1}{4} \\P(2 \text{ heads}) &= \frac{1}{4} \\P(1 \text{ head}) &= \frac{1}{2}\end{aligned}$$

| Probability mass function: $P(X = x_i)$

$$\sum_i P(X = x_i) = 1$$

| $P(X = x_i \cup X = x_j) = P(X = x_i) + P(X = x_j)$ if $i \neq j$

| $P(X = x_i \cap X = x_j) = 0$ if $i \neq j$

↘ $P(X = x_1 \cup \dots \cup X = x_k) = 1$

Probability Function



- | We need a **mathematical function** that provides the **probabilities of occurrence** of different possible outcomes
- | How different events will be **distributed** throughout the sample space

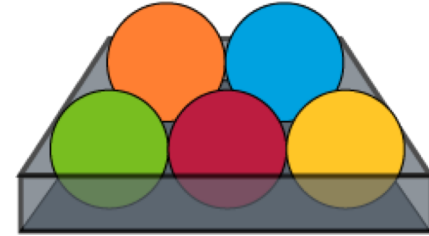
Common Distributions

| **Uniform** $X \sim U[1, \dots, N]$

– X takes values $1, 2, \dots, N$

$$P(X = i) = 1/N$$

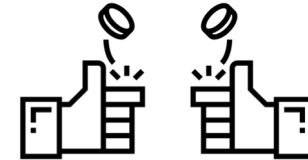
$\frac{1}{5}$



| **Binomial** $X \sim \text{Bin}(n, p)$

– X takes values $0, 1, \dots, N$

$$P(X = i) = \binom{n}{i} p^i (1 - p)^{n-i}$$



| **Joint:** Given two discrete RVs X and Y , their **joint distribution** is the distribution of X and Y together

$$\sum_x \sum_y P(X = x \cap Y = y) = 1$$

Conditional Probability

| $P(X = x|Y = y)$ is the probability of $X = x$, given the occurrence of $Y = y$

- You get A in Artificial Intelligence, given that you have studied the materials for the exam

$$P(X = x|Y = y) = \frac{P(X = x \cap Y = y)}{P(Y = y)}$$

Marginalization

- | Probability of a subset of a collection of random variables
- | Probability distribution of variables contained in the subset
- | If X and Y are two discrete RVs:

$$\begin{aligned} P(X = x_i) &= \sum_j P(X = x_i \cap Y = y_j) \\ &= \sum_j P(X = x_i | Y = y_j) P(Y = y_j) \end{aligned}$$

Bayes' Rule

$$P(X = x|Y = y) = \frac{P(X = x \cap Y = y)}{P(Y = y)}$$



$$P(X = x_i|Y = y_j) = \frac{P(Y = y_j|X = x_j)P(X = x_i)}{\sum_k P(Y = y_j|X = x_k)P(X = x_k)}$$

$$P(\text{★ A} \mid \text{📖}) = \frac{P(\text{📖} \mid \text{★ A})P(\text{★ A})}{\{P(\text{📖} \mid \text{★ A})P(\text{★ A}) + P(\text{📖} \mid \text{F})P(\text{F})\}}$$

Summary



- | **Probability definition**
- | **Discrete random variables and their probability**
- | **Distributions**
 - Uniform
 - Binomial
 - Joint
- | **Marginalization**
- | **Bayes' Rule**