

# Probability Cheat Sheet

A **set** is a collection of **distinct objects**, which can be considered an object in its own right. The **arrangement** of the objects inside the set **does not matter!**

A = {1, 9, 5, 3, 7} is the set of positive odd numbers less than 10.

B = {Red, Green, Blue} is the set containing these 3 colors.

C = {A, B} is the set containing the sets A and B!

$A \subset B$

$A \subseteq B$

$A = B$

$A \supseteq B$

$A \supset B$



Set A is a **subset** of set B.



Set A is **equal** to set B.



Set A is a **superset** of set B.

$\{\text{Cat, Dog}\} \subset \{\text{Cat, Dog, Mouse}\}$

The **sample space** ( $\Omega$ ) is the set of all possible outcomes of an experiment.

## Properties

Rolling a die:  $\Omega = \{1, 2, 3, 4, 5, 6\}$

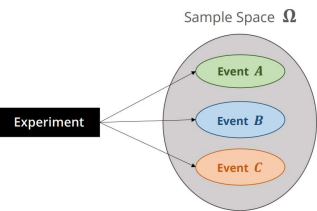
Taking a pass/fail class:  $\Omega = \{\text{Pass, Fail}\}$

Measuring a qubit:  $\Omega = \{|0\rangle, |1\rangle\}$

Collectively exhaustive

Mutually exclusive elements

In simple words, two outcomes cannot occur simultaneously

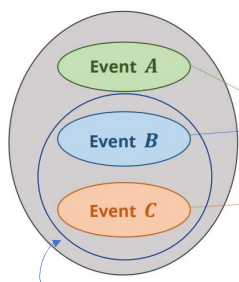


$\Omega = \{A, B, C\}$

(we denote all the events in the space using set notation)

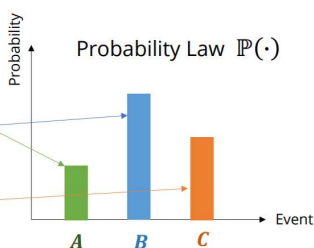
A **probability law** ( $\mathbb{P}$ ) assigns a probability to each element of the sample space.

Sample Space  $\Omega$



Event D

Events can combine multiple possible outcomes



If  $B \subseteq D$ , then  $\mathbb{P}(B) \leq \mathbb{P}(D)$

$\mathbb{P}(\text{empty set}) = \mathbb{P}(\{\}) = P(\emptyset) = 0$

$\mathbb{P}(A^c) = 1 - P(A)$

**Nonnegativity:**  $\mathbb{P}(A) \geq 0$ , for every event A

The probability of any event has to be **greater than or equal** to 0.

**Additivity:**  $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B)$

If A and B are **disjoint/mutually exclusive events**, then the above is true.

**Normalization:**  $\mathbb{P}(\Omega) = 1$

Since the sample space is **collectively exhaustive**, the probability of the outcome lying inside it is 1.

$\mathbb{P}(A \cap B) = 0$  when A and B are disjoint

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

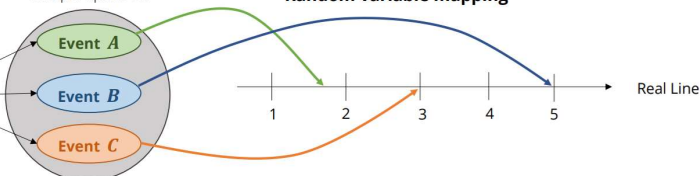
A **random variable** (RV) maps each possible outcome of an experiment to a real number.

A random variable is just a **function** from the sample space to the real line ( $\Omega \rightarrow \mathbb{R}$ ) !

Sample Space  $\Omega$

Random variable mapping

Experiment



The **Probability Mass Function (PMF)** associates a probability to each possible value of the random variable

HT & TH are disjoint events, they never occur simultaneously!

If we toss 2 **fair** coins, the possible outcomes are HH, HT, TH, HH, all **equally likely** (probability =  $\frac{1}{4}$ )

**Define** the RV as the number of heads. We get the following mapping:  
HH  $\rightarrow$  2, HT  $\rightarrow$  1, TH  $\rightarrow$  1, TT  $\rightarrow$  0

We obtain the following PMF:

$P(2) = P(HH) = \frac{1}{4}$

$P(1) = P(HT \cup TH) = P(HT) + P(TH) = \frac{1}{2}$

$P(0) = P(TT) = \frac{1}{4}$

The **expectation**  $E[X]$  of a random variable X is the **weighted average** of its possible values. The **variance** is the **expectation** of the random variable  $(X - E[X])^2$ .

$$E[X] = \langle X \rangle = \sum_x x \cdot \mathbb{P}(X = x)$$

The upper case "X" is the random variable while the lower case "x" is the different values "X" can take

$$\text{var}[X] = E[(X - E(X))^2] = \sum_x (x - E(X))^2 \cdot \mathbb{P}(X = x)$$

The **variance** of a random variable provides a measure of how **dispersed** the data individual values of X are, around the mean.

Even though the random variable might never equal its expected value, it is still a useful property.

**Law of Large Numbers:** As the number of trials of an experiment is increased, the observed average gets closer and closer to the expected value.

## The Bra-Ket notation

A **ket** is a column vector

A **bra** is the conjugate transpose of a ket (a row vector)

Ground state:  $|0\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  Excited state:  $|1\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

$$|v\rangle = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$$

$$\langle v| = |v\rangle^\dagger = (\overline{v_1} \ \overline{v_2} \ \dots \ \overline{v_n})$$

**Superposition state**  $|\Psi\rangle = \alpha|0\rangle + \beta|1\rangle$

$\alpha$  and  $\beta$  here are called **probability amplitudes**.

Measuring  $|\Psi\rangle$  will give  $|0\rangle$  w.p.  $|\alpha|^2$  and  $|1\rangle$  w.p.  $|\beta|^2$