

# Probability cheatsheet

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## 1. Probability expresses uncertainty:

$$\begin{array}{ll} P(X|C) = 0 & \Rightarrow \text{"X is definitely not true"} \\ P(X|C) = 1 & \Rightarrow \text{"X is definitely true"} \\ 0 < P(X|C) < 1 & \Rightarrow \text{we are uncertain about X} \end{array} \quad \left. \vphantom{\begin{array}{l} P(X|C) = 0 \\ P(X|C) = 1 \\ 0 < P(X|C) < 1 \end{array}} \right\} \text{Extends basic logic}$$

## 2. All probability is conditional:

$$\begin{array}{ccc} P(X) & \text{always means} & P(X|C) \\ \text{"probability of X"} & & \text{"probability of X given C"} \\ & & \text{or "probability of X conditional on C"} \end{array}$$

where **C** represents **background information** (such as an assumed statistical model). C is often dropped from notation but is always there.

Examples:  $P(\underbrace{\text{die}}_{\text{Model of die}} | \text{die is fair}) = 1/6$      $P(\text{transmission} | \text{model of infection})$     etc.

## 3. Probability is computable:

$$\begin{array}{ll} \text{I.} & P(\underbrace{\neg A}_{\text{"not A"}} | C) = 1 - P(A|C) \quad \text{(A either occurs or doesn't.)} \\ \text{II.} & P(A \text{ or } B | C) = P(A|C) + P(B|C) - P(A, B|C) \quad \text{(One of them must occur, but don't double-count)} \\ & \quad \quad \quad \text{"A and B given C"} \\ \text{III.} & P(A, B|C) = P(A|B, C) \cdot P(B|C) \quad \text{(Split up joint probabilities into conditional ones.)} \\ & \quad \quad \quad \text{"A and B given C"} \quad \text{"A given B and C"} \quad \text{"B given C"} \end{array}$$

"Independence" happens in when the middle term simplifies:  $P(A, B) = P(A)P(B)$

## 4. Practically useful tools for computation: (derived directly from the rules in 3)

**LOTP:** ("Law of total probability")

$$P(A|C) = \sum_X P(A, X|C)$$

Computes probability of A by summing (or integrating) over all the possible values of another variable.

Often seen as  $P(A|C) = \sum_X P(A|X, C) \cdot P(X|C)$  by applying rule III

**Bayes' rule:**

$$P(A|B, C) = \frac{P(B|A, C) \cdot P(A|C)}{P(B|C)}$$

Computes P(A given B) in terms of P(B given A)

## 4. Probability for variables

Often  $X$  will instead denote a variable whose value we are uncertain of. It is typical to use lower-case letters like  $x$  to denote specific possible values. So we will write:

$$P(X = x|C) \quad \text{"Probability X takes value x"}$$

For a continuous variable  $X$  we must instead think of  $P$  as a density function:

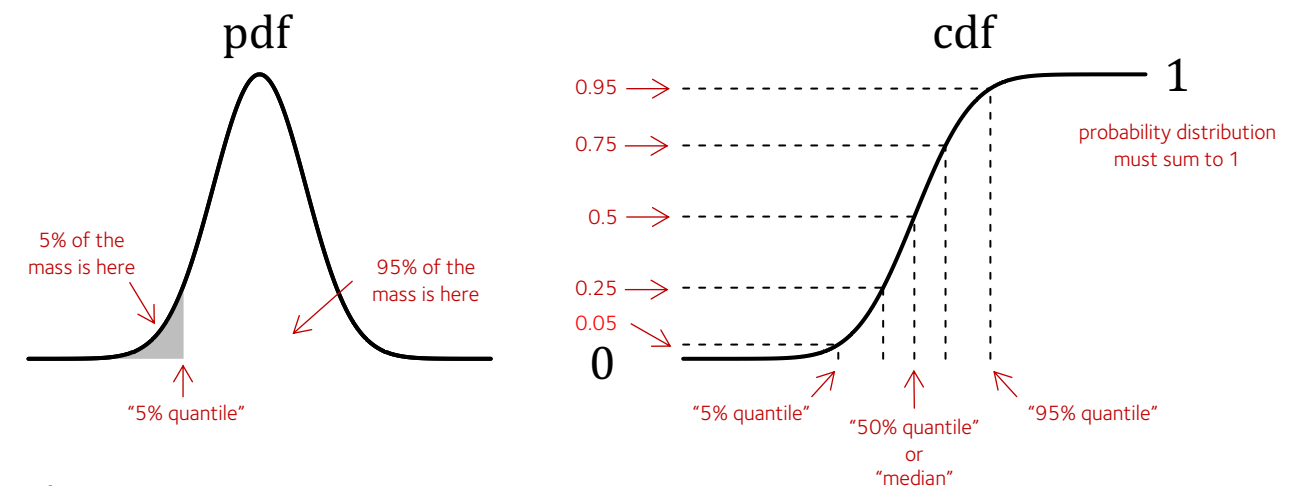
$$P(X = x|C) \quad \text{"Probability density function (pdf) evaluated at } X = x\text{"}$$

and integrate it to get an actual probability:

$$P(a \leq X \leq b|C) = \int_a^b P(X = x|C) dx \quad \text{"Probability that X is between a and b"}$$

$$P(X \leq b|C) = \int_{-\infty}^b P(X = x|C) dx \quad \text{"Cumulative distribution function (cdf) of X"}$$

Inverse of cdf is the **quantile function**.



## 5. Distributions

Probabilities are often expressed in terms of a set of well-known distributions

$X \sim N(0,1)$  "X has a standard normal (or Gaussian) distribution"

$X \sim \text{binom}(N, p)$  "X has binomial distribution with  $N$  draws and success probability  $p$ "

See the separate 'distributions cheatsheet' for more.

## 6. Statistical summaries of a distribution

**"Mean" or "expected value"**  $\mu = E(x) = \sum_x \overset{\text{value}}{x} \cdot \overset{\text{pdf}}{P(X = x)}$

sum (or integrate) over all possible values

**"Variance"**  $\text{var}(x) = E((x - \mu)^2) = \sum_x (x - \mu)^2 \cdot P(X = x)$

average squared distance to mean value (a measure of uncertainty)

**"Entropy"**  $H(x) = \sum_x \frac{1}{\log P(X = x)} P(X = x)$

average information provided by  $x$  (a measure of uncertainty)