# Probability

When we are looking into the future, an expected event will happen with a certain probability. This probability ranges from 0 ("will not happen" as in *Two suns will simultaneously rise tomorrow*) to 1 ("is guaranteed to happen" as in *The sun will set tonight*.)

Note that **probability** is firmly associated with **future**. The probability that something happened yesterday is always 0 or 1.

# Random variable

A random variable is a set of homogeneous objects (numbers, colours, letters) with a probability associated with each object. Considering that *probability* is part of the future, so is a *random variable*. The set can be infinite, etc.

Note that the sum of the probabilities of all the objects in the set must add to 1.

A random variable has no specific value; the whole set represents it.

Standard notation calls for naming random variables  $\mathcal{X}, \mathcal{Y}, \mathcal{Z}$ .

If the random variable has a countable ( $\subseteq \aleph_0$ ) set of values, the variable is called **discrete**; otherwise, it is called **continuous**.

The definitions vary somewhat for the two types, the former using summation and the latter integral calculus.

## **Probability distribution-discrete**

If the set of values of a random variable  $\mathcal{X}$  is finite, a probability is associated with each member of the set. **Probability distribution** is the fancy term for the collection of probabilities.

Quite formally, the probability distribution function  $f:\mathcal{X}\to<0,1>\text{or }f(\mathcal{X})\text{ of a discrete random variable}$   $\mathcal{X}$  is

$$f(x) = \begin{cases} p(x) & x \in \mathcal{X} \\ 0 & x \notin \mathcal{X} \end{cases}$$

For a continuous distribution  $\mathcal{X}$ , the probability  $P(\mathcal{X} = x) = 0$  for all values of x (artificial exceptions being ignored), so an alternate definition based on the cumulative distribution must be used. We call it the **probability density function**  $f(\mathcal{X})$  (name conveniently chosen to abbreviate **pdf** as well).

For practical purposes, a **pdf** is equivalent to the random variable that it represents.

#### **Cumulative distribution**

Cumulative distribution function:

$$F(x) = P(\mathcal{X} \le x)$$

If we know the  $f(\mathcal{X})$  of a variable  $\mathcal{X}$ ,

$$F(x) = \int_{-\infty}^{x} f(x) dx$$

(the integral degenerates to a summation for discrete variables).

The cumulative distribution function gives the actual distribution of probabilities by interval: the probability of a random variable taking a value from interval (a, b) equals F(b) - F(a).

For continuous distributions, the probability density function f(x) equals:

$$f(x) = \frac{d}{dx}F(x)$$

This pair of definitions is circular.

## Variate aka Random Number

When one picks a specific object from the set forming a random variable, that object becomes a **variate**, more commonly called a random number.

A variate is a constant and there is nothing *random* about it.

Intuitively, a variate is part of the past in the same way as a probability is part of the future.

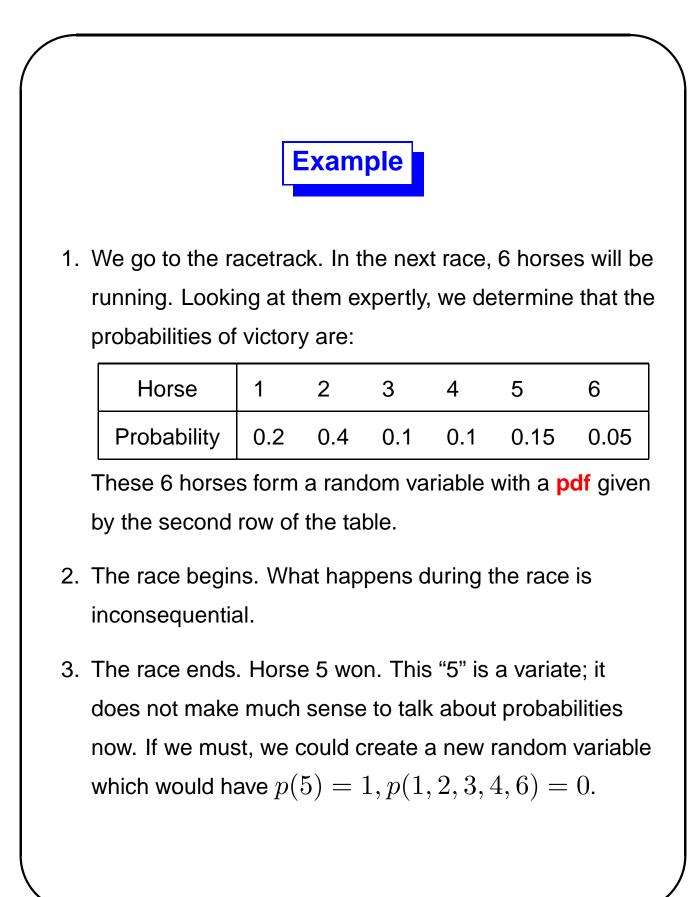
### Variate vs. Random Variable

Consider a deck of 52 cards spread on a table face down.

This is a random variable made of a set of 52 objects; a probability of  $\frac{1}{52}$  is associated with each object.

We touch one card; this card can be any of the 52, hence it represents the whole variable. The probability that this card is a  $2\clubsuit$  is  $\frac{1}{52}$ .

Now we turn the card face up and discover that it is a  $K \spadesuit$ . This  $K \spadesuit$  is now a fact (with probability 1) and there is nothing random about it anymore. We call it a variate. Note that it became a variate at the moment when it was turned face up, i.e. when the card lost its random nature.





Every random variable  $\mathcal{X}$  has two properties: a **mean** and a **variance**.

• Mean (or expectation)  $\mu$  or E()

The general formula for discrete distributions is:

$$\mu = E(\mathcal{X}) = \sum_{i=1}^{n} p(x_i) x_i$$

(The assumption is that  $\mathcal{X}$  has n members.)

For continuous variables:

$$\mu = \int_{-\infty}^{\infty} x f(x) dx$$

It is a mistake to assume that  $P(x \leq \mu) = 1/2$ 

The mean, used in isolation, tells us very little about the random variable. When combined with the variance, it becomes very useful.

• Variance V() or  $\sigma^2$ 

The general formula is:

$$\sigma^2 = E(\mathcal{X}^2) - \mu^2$$

For discrete distributions:

$$\sigma^{2} = V(X) = \sum_{i=1}^{n} x_{i}^{2} p(x_{i}) - \mu^{2}$$

The standard deviation  $\sigma = \sqrt{V(X)}$  has the same unit as the mean.

The variance (and standard deviation) is independent of the mean:

$$V(\mathcal{X}+c) = V(\mathcal{X})$$

# **Uniform distribution**

A random variable  $\mathcal{X}$  is **uniformly distributed** on the interval (a, b) if its **pdf** is given by:

$$f(x) = \begin{cases} \frac{1}{b-a} & a \le x \ge b\\ 0 & \text{otherwise} \end{cases}$$

The expert will quickly calculate

$$\mu = \int_{a}^{b} \frac{x}{b-a} dx = \frac{x^{2}}{2(b-a)} |_{a}^{b} = \frac{b^{2}-a^{2}}{2(b-a)} = \frac{b+a}{2}$$
$$\sigma^{2} = \frac{(b-a)^{2}}{12}$$

Uniform variates can be generated by calling a standard library function which returns a variate uniformly distributed on the interval (0,1). We denote a uniform variate as U(0,1).

Typical library function name: **drand48()**. when not available, the expression **(double)random() / MAXINT** will do.

