

# **Data 100, Discussion 6 – Probability and Modelling**

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## Random Variables

A **random variable** is a variable whose values are determined by probabilities. In other words, a random variable is a function that takes outcomes of a random process to real numbers.

Usually, we use  $X$  or  $Y$  to denote a random variable.

# Probability Mass Functions

The **probability mass function**, or **density function**, tells us the specific probabilities of each event occurring. In general, these can be discrete or continuous, but for the purposes of this class they will be discrete.

$P(X = x)$  represents the probability of random variable  $X$  taking on the value of  $x$ . ( $x$  is normally numerical, but it doesn't necessarily have to be.)  $\mathbb{X}$ , then, represents the set of the actual values themselves (i.e. dice rolls, cards, etc.)

*f.v.*  
↑  
*specific outcome*

$\mathcal{X} : \{1, 2, 3, 4, 5, 6\}$        $x \in \mathcal{X}$

A few properties that  $P(X = x)$  must satisfy:

$$0 \leq P(X = x) \leq 1, \forall x \in \mathbb{X}$$

$$\sum_{x \in \mathbb{X}} P(X = x) = 1$$

# Expectation

Often, we want to determine the expected, or average, outcome of a probabilistic event. For that, we define the **expectation** of a random variable:

$$\mathbb{E}[X] = \sum_{x \in \mathbb{X}} x \cdot P(X = x) = \sum \text{outcome} \cdot P(\text{outcome})$$

↑  
sum over all possible outcomes

This is a weighted average of all possible outcomes.

Properties that the expectation satisfies:

$$\mathbb{E}[aX] = a\mathbb{E}[X]$$

$$\mathbb{E}[c] = c, \quad \forall c \in \mathbb{R}$$

$$\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y], \quad \forall r.v. X, Y$$

Independent

$$P(X=x, Y=y) = P(X=x)P(Y=y)$$

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

independent  $\implies$  uncorrelated  
but not vice versa!

$X$ : fair, 6 sided die

$$\begin{aligned} \mathbb{E}[X] &= 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + \dots + 6 \cdot \frac{1}{6} \\ &= 3.5 \end{aligned}$$

# Variance

We define variance as being the mean squared distance from the mean, i.e.

$$\text{var}[X] = \mathbb{E}[(X - \mathbb{E}[X])^2]$$
$$E[(X - \mu)^2]$$

Properties that variance satisfies:

$$\text{var}[aX] = a^2 \text{var}[X]$$

$$\text{if } X, Y \text{ are independent : } \text{var}[X + Y] = \text{var}[X] + \text{var}[Y]$$

$$E[g(X)] = \sum_{x \in \mathcal{X}} g(x) \cdot P(X=x)$$

$$E[X^2] = \sum x^2 P(X=x)$$

## Bernoulli Random Variables

A Bernoulli random variable is one that only assumes two possible values – 0 or 1. It has a single parameter,  $p$ .

$$P(X = 1) = p$$

$$P(X = 0) = 1 - p$$

$$\sum P(X=x) = P(X=0) + P(X=1) = 1-p+p = 1$$

$$\mathbb{E}[X] = 0 \cdot P(X = 0) + 1 \cdot P(X = 1) = p$$

$$\text{var}[X] = p(1 - p)$$

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