

# Lecture 3: Discrete Random Variables

Professor Ilias Bilonis

## The probability mass function

# Probability mass function

Let  $X$  be a discrete random variable. The *probability mass function (pmf)* of  $X$  is:

$p(X = x)$  = Probability that the random variable  $X$  takes the value  $x$

# Probability mass function

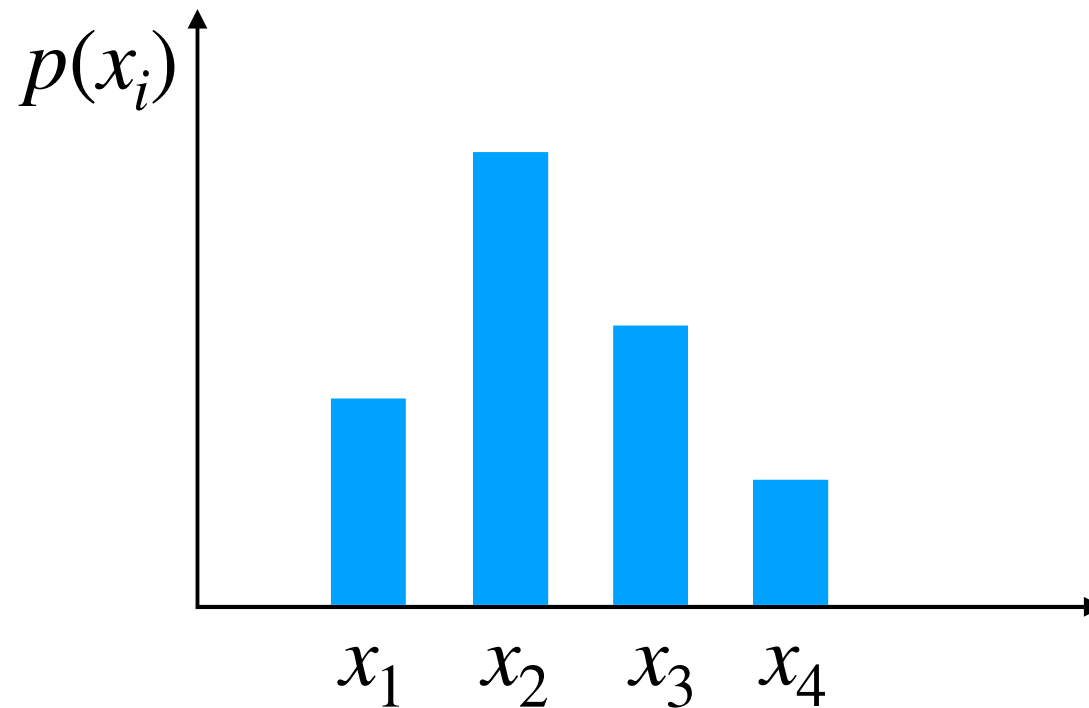
Let  $X$  be a discrete random variable. The *probability mass function (pmf) of  $X$*  is:

$p(X = x)$  = Probability that the random variable  $X$  takes the value  $x$

When there is no ambiguity:

$$p(x) \equiv p(X = x) .$$

# Visualization of the probability mass function



# Properties of the probability mass function

- The probability mass function is nonnegative:

$$\underline{p(x) \geq 0.}$$

- The probability mass function is normalized:

$$\underline{\sum_x p(x) = 1,}$$

where the summation is over all the possible values of  $X$ .

# Properties of the probability mass function

- Let  $X$  be a discrete random variable.
- The probability of  $X$  taking either the value  $x_1$  or the value  $x_2$  (assuming  $x_1 \neq x_2$ ) is:

$$\begin{aligned} \underline{p(X = x_1 \text{ or } X = x_2)} &\equiv p(X \in \{x_1, x_2\}) = p(X = x_1) + p(X = x_2) \\ &= p(x_1) + p(x_2) \end{aligned}$$

# Properties of the probability mass function

- More generally, the probability that the random variable  $X$  takes any value in a set  $A$  is given by:

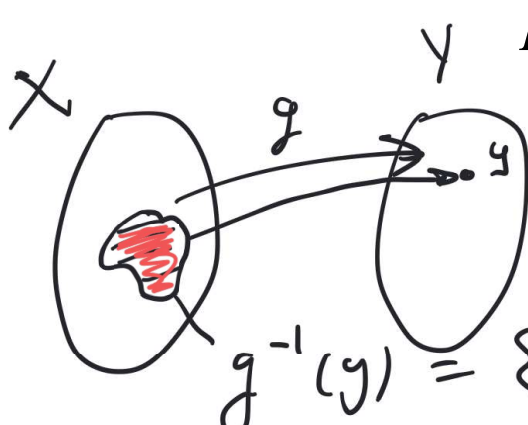
$$p(X \in A) = \sum_{x \in A} p(x)$$

# Functions of random variables

- Consider a function  $g(x)$ .
- We can now define a new random variable:

$$\underline{Y = g(X)}.$$

- It has its own probability mass function (pmf):

$$p(y) = p(Y = y) = \sum_{x \in g^{-1}(y)} p(x)$$


The diagram shows two sets,  $X$  and  $Y$ , represented as ovals. A function  $g$  maps elements from  $X$  to  $Y$ . A specific element  $y$  in  $Y$  is shown, and its pre-image in  $X$  is a red-shaded region. This pre-image is labeled  $g^{-1}(y) = \{x : g(x) = y\}$ .

where