

# Lecture 10 Random Variables, PMF, Expectation & Variance

BIO210 Biostatistics

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## What is a random variable (r.v.) ?

- An assignment of a value (a real number) to every possible outcome in the sample space.
- **Mathematically:** A real-valued **function** defined on a sample space  $\Omega$ . In a particular experiment, a random variable (r.v.) would be some function that assigns a real number to each possible outcome.

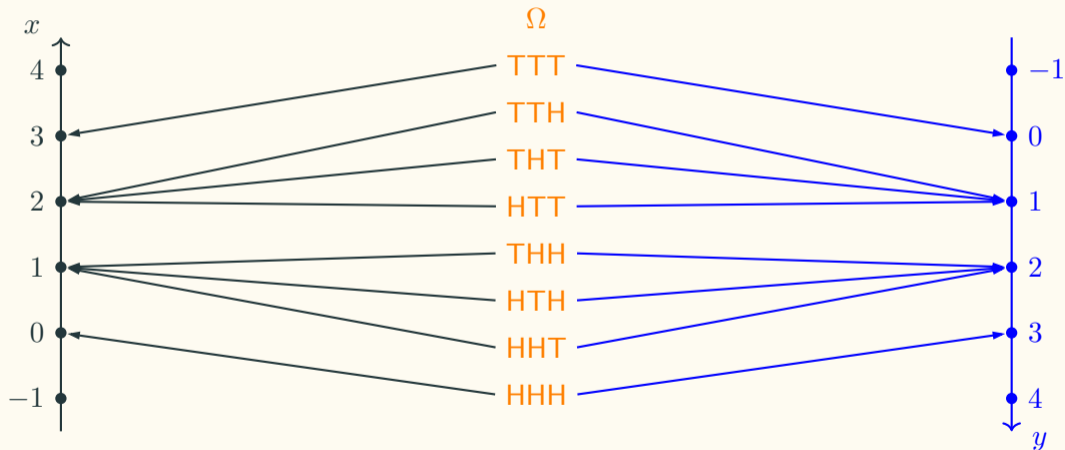
## More about random variables

- Discrete or continuous.
- Can have several random variables defined on the same sample space.
- Notation
  - random variable  $X$  : function  $\Omega \mapsto \mathbb{R}$
  - numerical value:  $x$  : value  $\in \mathbb{R}$

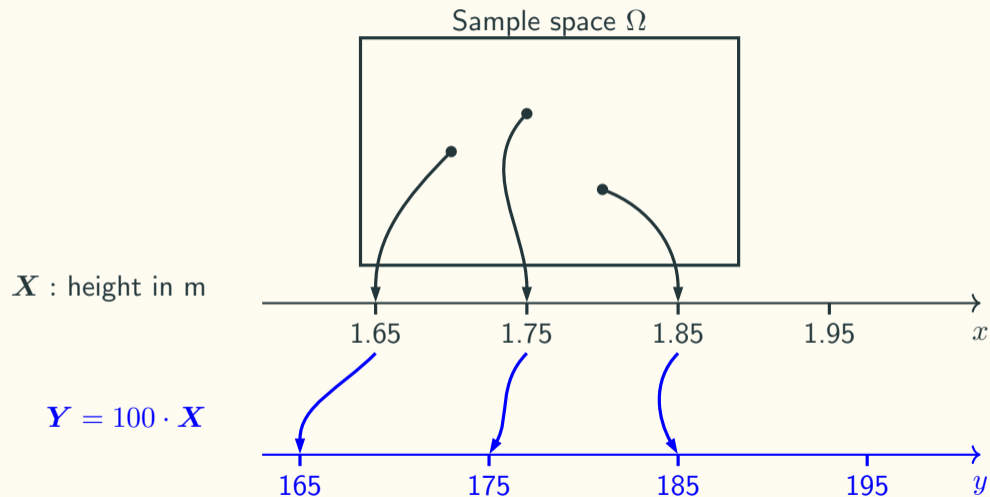
# Different random variables on the same sample space

$X$  : number of tails

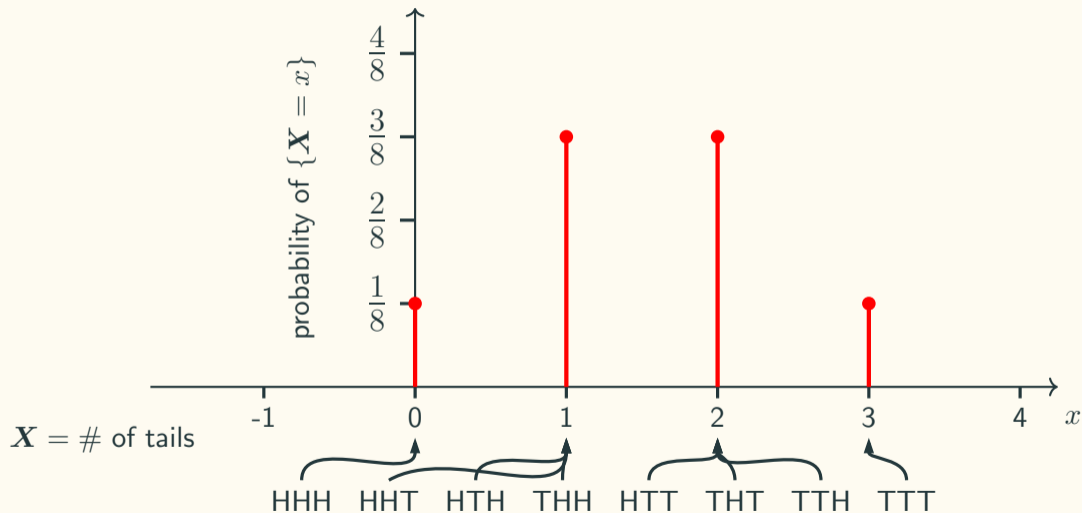
$Y$  : number of heads



## Function of a random variable is an r.v.



# Probability Mass Function (PMF)



## Probability Mass Function (PMF)

The PMF of  $\mathbf{X}$  = number of tails after three flips

$x$	$\mathbb{P}(\{\mathbf{X} = x\})$
0	1/8
1	3/8
2	3/8
3	1/8
otherwise	0

$$\mathbb{P}(\{\mathbf{X} = x\}) = \begin{cases} \frac{1}{8}, & x = 0, 3 \\ \frac{3}{8}, & x = 1, 2 \\ 0, & \text{otherwise} \end{cases}$$

# PMF Notation

## Probability Mass Function

- Notation

$$\begin{aligned}\mathbb{P}_{\mathbf{X}}(x) &= \mathbb{P}(\{\mathbf{X} = x\}) \\ &= \mathbb{P}(\{\omega \in \Omega \mid \mathbf{X}(\omega) = x\})\end{aligned}$$

- Properties

$$\begin{aligned}\mathbb{P}_{\mathbf{X}}(x) &\geq 0 \\ \sum_x \mathbb{P}_{\mathbf{X}}(x) &= 1\end{aligned}$$

$\omega$	$\mathbf{X}(\omega) = x$	$\mathbb{P}_{\mathbf{X}}(x) = \mathbb{P}(\{\mathbf{X} = x\})$
HHH	0	$\frac{1}{8}$
TTH, HTH, HHT	1	$\frac{3}{8}$
TTH, THT, THT	2	$\frac{3}{8}$
TTT	3	$\frac{1}{8}$



## Geometric PMF

**Experiment:** keep flipping a coin ( $\mathbb{P}(H) = p$ ) until a head comes up for the first time. Let the random variable  $X$  be the number of flips.

$\omega$	$X(\omega)$	$\mathbb{P}_X(x)$
H	1	$p$
TH	2	$(1-p)p$
TTH	3	$(1-p)^2p$
$\vdots$	$\vdots$	$\vdots$
$\underbrace{\text{TTT}\dots\text{TTT}}_{n-1}\text{H}$	$n$	$(1-p)^{n-1}p$

Geometric PMF.  $X$ : geometric random variable.

## How to compute a PMF $\mathbb{P}_{\mathbf{X}}(x)$

**To compute a PMF  $\mathbb{P}_{\mathbf{X}}(x)$ :**

1. Collect all possible outcomes for which  $\mathbf{X} = x$ ;
2. add their probabilities;
3. repeat for all  $x$ .

# Compute PMF

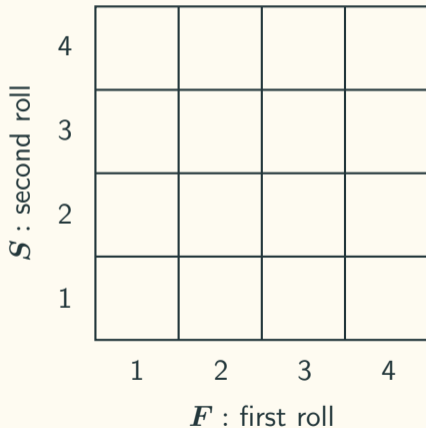
**Experiment:** two independent rolls of a fair tetrahedral die.

$F$ : outcome of the first roll

$S$ : outcome of the second roll

$X = \min(F, S)$

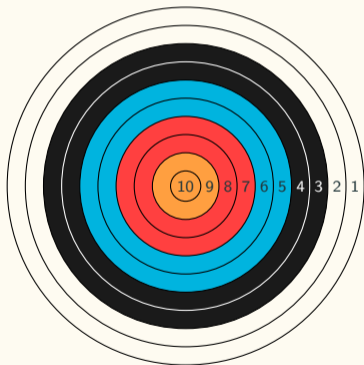
$\mathbb{P}_X(x) = ?$



## Expected value of a random variable (Expectation)

### Experiment: archery

Let  $X$  be the score you get for each shot. What is the expected value of  $X$  ?



Think: What is the average score you will get after a large number of trials?

$x$	$\mathbb{P}_X(x)$
1	0.19
2	0.17
3	0.15
4	0.13
5	0.11
6	0.09
7	0.07
8	0.05
9	0.03
10	0.01

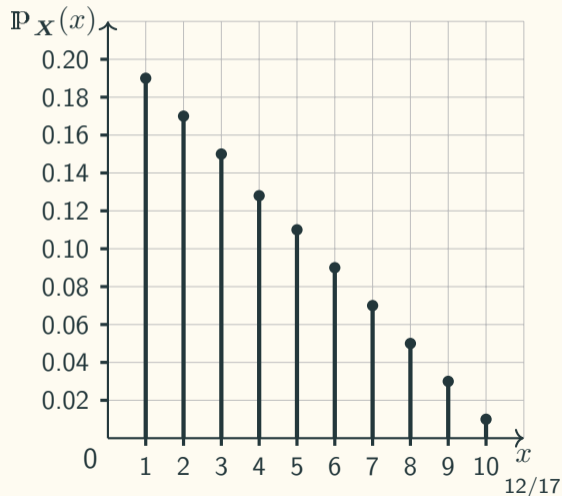
# Expected value (Expectation)

## Definition

$$\mathbb{E}[X] = \sum_x x \mathbb{P}_X(x)$$

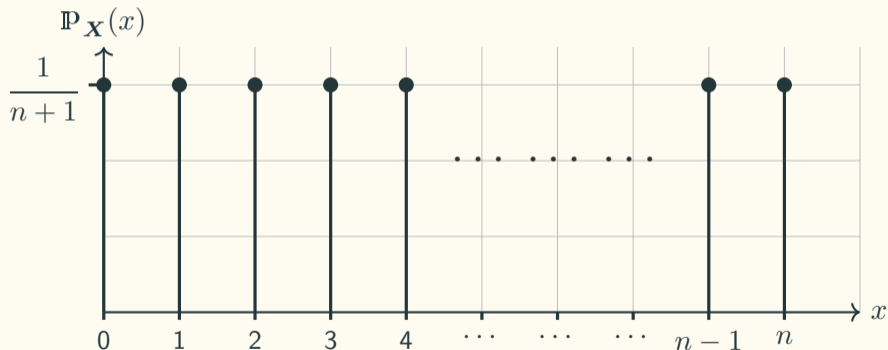
- Interpretation
  1. Centre of gravity of the PMF
  2. Average in large number of repetitions of the experiment

PMF of  $X$  from the archery experiment



# Expectation of a Uniform Distribution

**Example:** a uniform discrete random variable  $X$  on  $0, 1, 2, 3, \dots, n$



**What is  $\mathbb{E}[X]$  ?**

# Properties of expectations

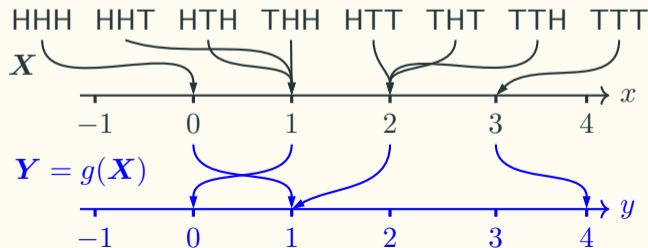
Let  $X$  be a random variable, and let  $Y = g(X)$ , what is  $\mathbb{E}[Y]$ ?

- The hard way:

$$\mathbb{E}[Y] = \sum_y y \mathbb{P}_Y(y)$$

- The easy way:

$$\mathbb{E}[Y] = \sum_x g(x) \mathbb{P}_X(x)$$



$y$	$\mathbb{P}_Y(y)$
0	3/8
1	4/8
4	1/8

$x$	$g(x)$	$\mathbb{P}_X(x)$
0	1	1/8
1	0	3/8
2	1	3/8
3	4	1/8

## Expectation of a linear function of r.v.

- Caution: in general  $\mathbb{E}[g(\mathbf{X})] \neq g(\mathbb{E}[\mathbf{X}])$
- Exception: if  $\alpha, \beta$  are constants, then we have:
  - $\mathbb{E}[\alpha] = \alpha$
  - $\mathbb{E}[\alpha\mathbf{X}] = \alpha\mathbb{E}[\mathbf{X}]$
  - $\mathbb{E}[\alpha\mathbf{X} + \beta] = \alpha\mathbb{E}[\mathbf{X}] + \beta$



# Variance and standard deviation of a random variable

## Definition of Variance

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

## Properties of Variance

- $\text{Var}(\mathbf{X}) = \mathbb{E}[\mathbf{X}^2] - (\mathbb{E}[\mathbf{X}])^2$
- If  $\alpha, \beta$  are constants, then  $\text{Var}(\alpha\mathbf{X} + \beta) = \alpha^2\text{Var}(\mathbf{X})$

## Definition of Standard Deviation

$$\sigma_{\mathbf{X}} = \sqrt{\text{Var}(\mathbf{X})}$$

# Discrete Random Variables (Summary slide)

