

General definition of independence

we say $\{A_1, A_2, \dots, A_n\}$ is independent
iff

$$P(A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_k}) = P(A_{i_1}) \cdot P(A_{i_2}) \cdot \dots \cdot P(A_{i_k})$$

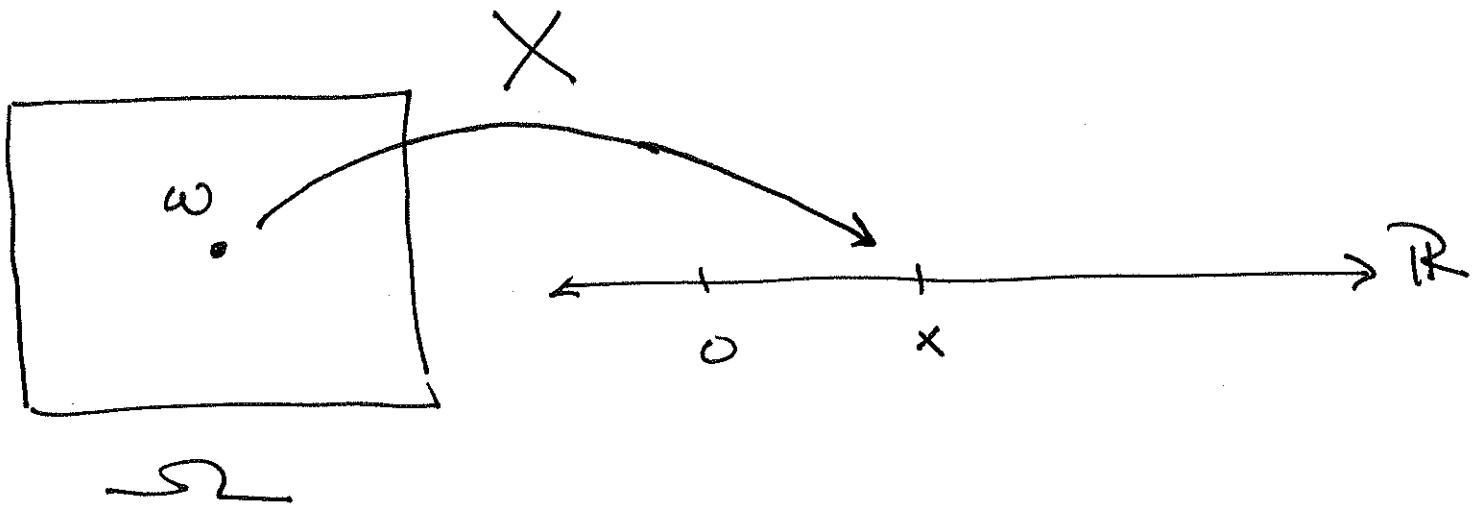
for every $\{i_1, i_2, \dots, i_k\} \subseteq \{1, 2, \dots, n\}$

2.1 Random Variables

Defn

A random variable is a function

$$X: \Omega \rightarrow \mathbb{R}$$



notation

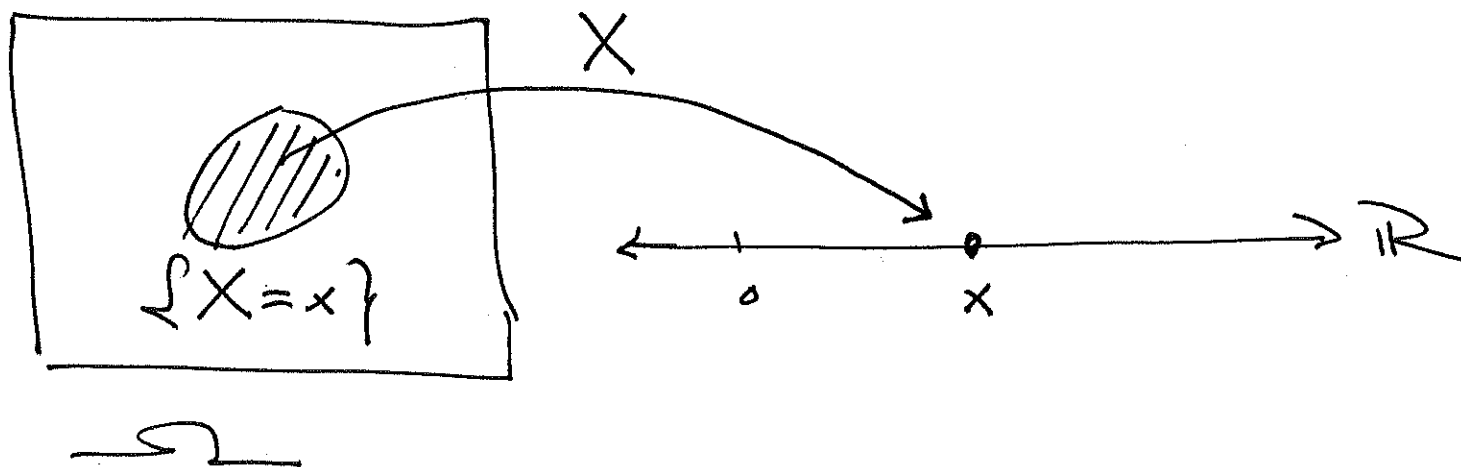
- use upper case for r.v.'s X
- use lower case for values of a r.v. x

$$X(\omega) = x$$

• denote by $\{X=x\}$ the event

$$\{X=x\} = \{\omega \in \Omega \mid X(\omega)=x\} \subseteq \Omega$$

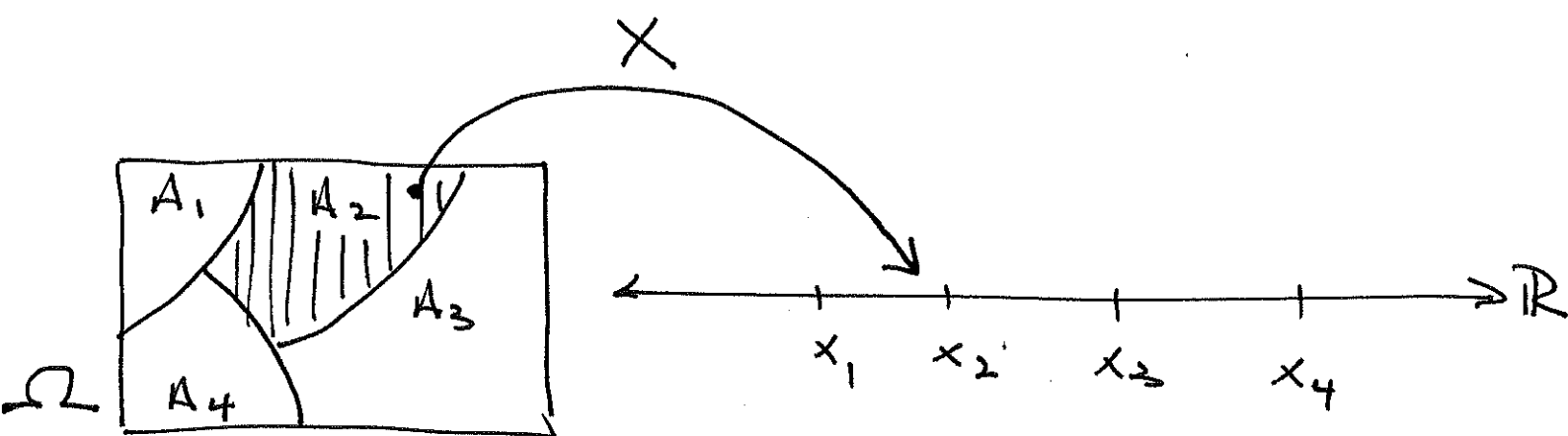
↑
called: Preimage of x
 under X .



• think of X as mapping 'mass' from Ω to $\mathbb{R}$.

- X is called a discrete r.v. iff its range is a discrete subset of $\mathbb{R}$

$$\text{range}(X) = \{X(\omega) \in \mathbb{R} \mid \omega \in \Omega\} \subseteq \mathbb{R}$$



here $A_i = \{X = x_i\}$

- X is called a continuous r.v.
later chap. 3.
- There are r.v.s that are neither discrete nor continuous.

Ex. (discrete r.v.)

Roll 2 fair independent dice.

let $X((i,j)) = i+j$ (sum).

$$\Omega = \left\{ \begin{array}{l} 11 \ 12 \ 13 \ \dots \ 16 \\ 21 \ 22 \ \dots \ 26 \\ \vdots \\ 61 \ 62 \ \dots \ 66 \end{array} \right\} \rightarrow \overbrace{\{2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}}^{\text{range}}$$

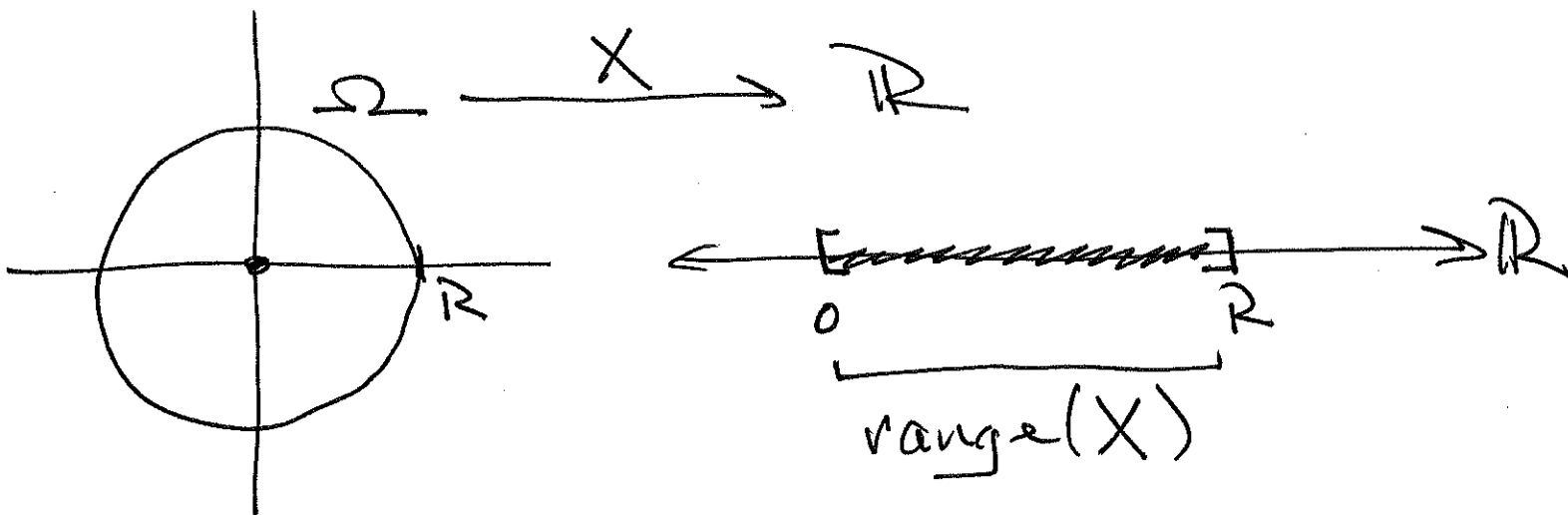
$$\Omega \xrightarrow{X} \mathbb{R}$$

Ex. (continuous r.v.)

Throw a dart at a circular target centered at $(0,0)$.

Let $X((a,b)) = \sqrt{a^2 + b^2}$,

i.e. distance from point (a,b) to $(0,0)$.



2.2 Probability mass functions

□

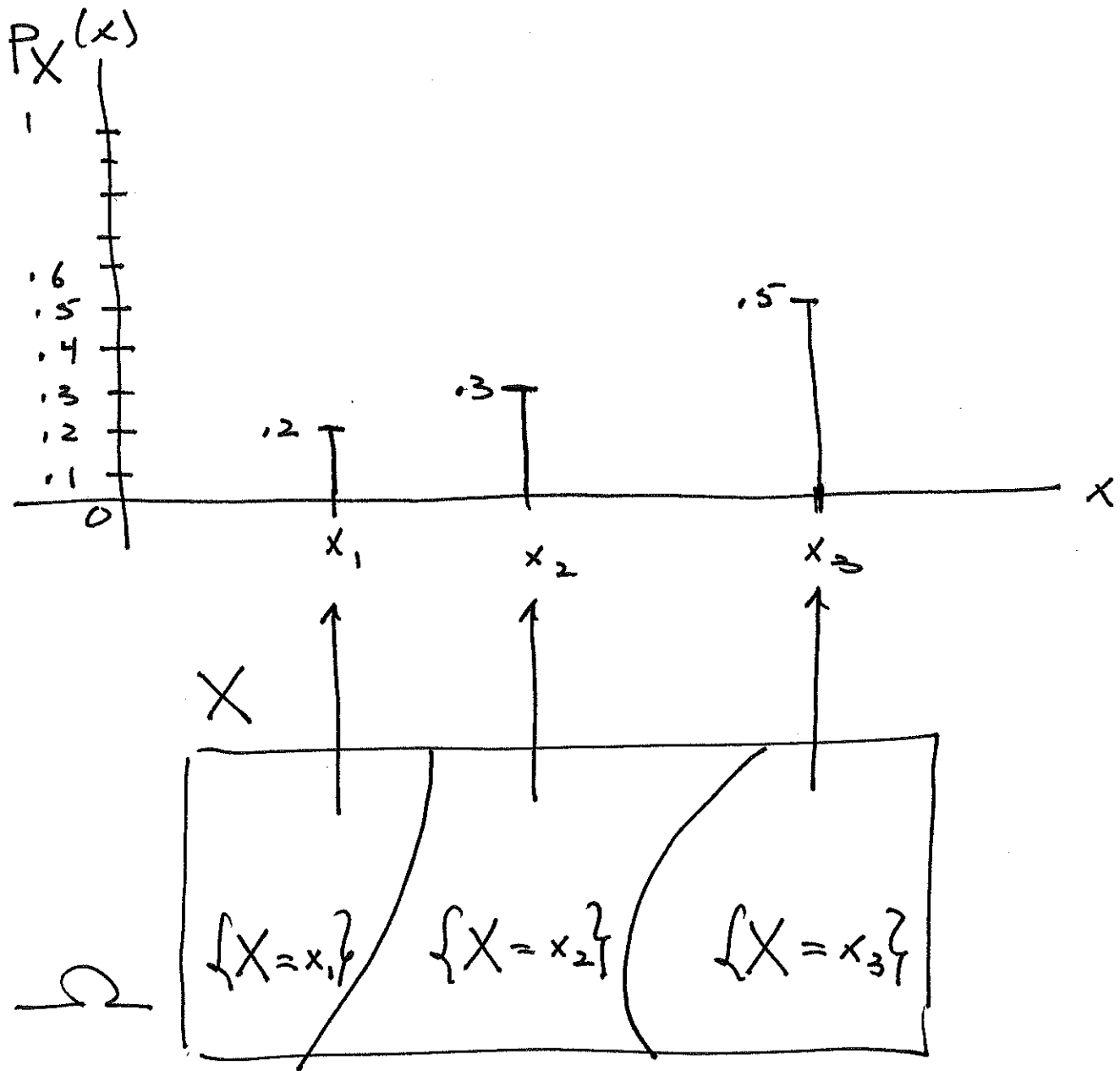
Let X be a discrete r.v. The Probability mass function (PMF) of X is the function

$$p_X(x) = P(\{X=x\}) = P(X=x)$$

observe $p_X : \mathbb{R} \rightarrow [0,1]$. Also

note that necessarily

$$\sum_{x \in \text{range}(X)} p_X(x) = 1$$



$$P_X(x) = \begin{cases} .2 & \text{if } x = x_1 \\ .3 & \text{if } x = x_2 \\ .5 & \text{if } x = x_3 \\ 0 & \text{otherwise} \end{cases}$$

note $\therefore \sum_x P_X(x) = .2 + .3 + .5 = 1.0$

The Bernoulli random variable

Perform one Bernoulli trial. let

$$X(\omega) = \begin{cases} 1 & \text{if } \omega = \text{Success} \\ 0 & \text{if } \omega = \text{failure} \end{cases}$$

where probability of success is p , and failure is $1-p$. Its PMF is

$$P_X(k) = \begin{cases} p & \text{if } k=1 \\ 1-p & \text{if } k=0 \end{cases}$$

p is called the parameter of the Bernoulli r.v.

The Binomial random variable

Perform a sequence of n independent Bernoulli trials.

Let $X(\omega) = \# \text{ of successes}$,

The PMF is

$$P_X(k) = P(X=k)$$

$$= \binom{n}{k} p^k \cdot (1-p)^{n-k}$$

for $k=0, 1, 2, \dots, n$ (and $P_X(k)=0$ otherwise)

check

$$\sum_{k=0}^n P_X(k) = \sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k}$$

$$= (p + (1-p))^n = 1^n = 1$$

✓

note. The Binomial r.v. has
2 Parameters:

$$p = P(\text{success in any trial})$$

$$n = \# \text{ of trials}$$

The Geometric random variable
Identical

Perform n independent Bernoulli trials
(with parameter p). Let X be
the # trials until 1st success.

What's the PMF?

$$P_X(k) = P(X = k) \quad (k = 1, 2, 3, \dots)$$

= Probability of $(k-1)$ failures
followed by 1 success

So

$$P_X(k) = (1-p)^{k-1} \cdot p \quad (k=1, 2, 3, \dots)$$

note:

$$\sum_{k=1}^{\infty} P_X(k) = \sum_{k=1}^{\infty} (1-p)^{k-1} \cdot p$$

$$= p \cdot \sum_{k=0}^{\infty} (1-p)^k$$

$$= p \cdot \frac{1}{1-(1-p)} = \cancel{p} \cdot \frac{1}{\cancel{p}} = 1 \checkmark$$

$$\text{recall: } \sum_{k=0}^{\infty} r^k = \frac{1}{1-r} \checkmark$$

where $-1 < r < 1$.(geometric series with common ratio r .)

Recall

$$\sum_{k=0}^{n-1} r^k = \frac{1-r^n}{1-r} \quad (r \neq 1)$$

Proof

$$S = \sum_{k=0}^{n-1} r^k$$

$$rS = \sum_{k=0}^{n-1} r^{k+1}$$

$$\begin{aligned} \therefore S - rS &= (r^0 + \cancel{r^1} + \dots + r^{n-1}) \\ &\quad - (\cancel{r^1} + \cancel{r^2} + \dots + \cancel{r^{n-1}} + r^n) \end{aligned}$$

$$S(1-r) = 1 - r^n$$

$$S = \frac{1-r^n}{1-r} \xrightarrow[\text{as } n \rightarrow \infty]{\text{lim}} \frac{1}{1-r}$$

$$\text{if } -1 < r < 1$$