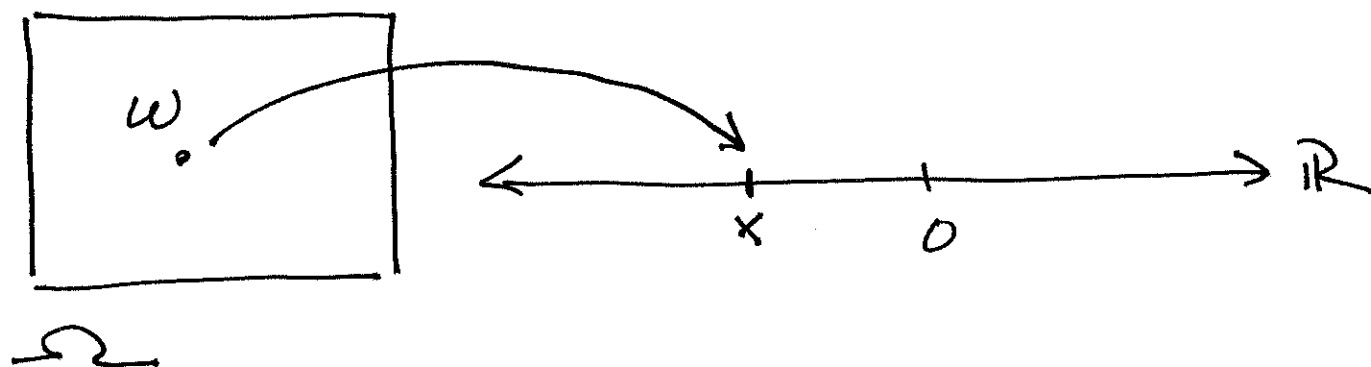


2.1 Random Variables

Defn

A random variable is a function

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

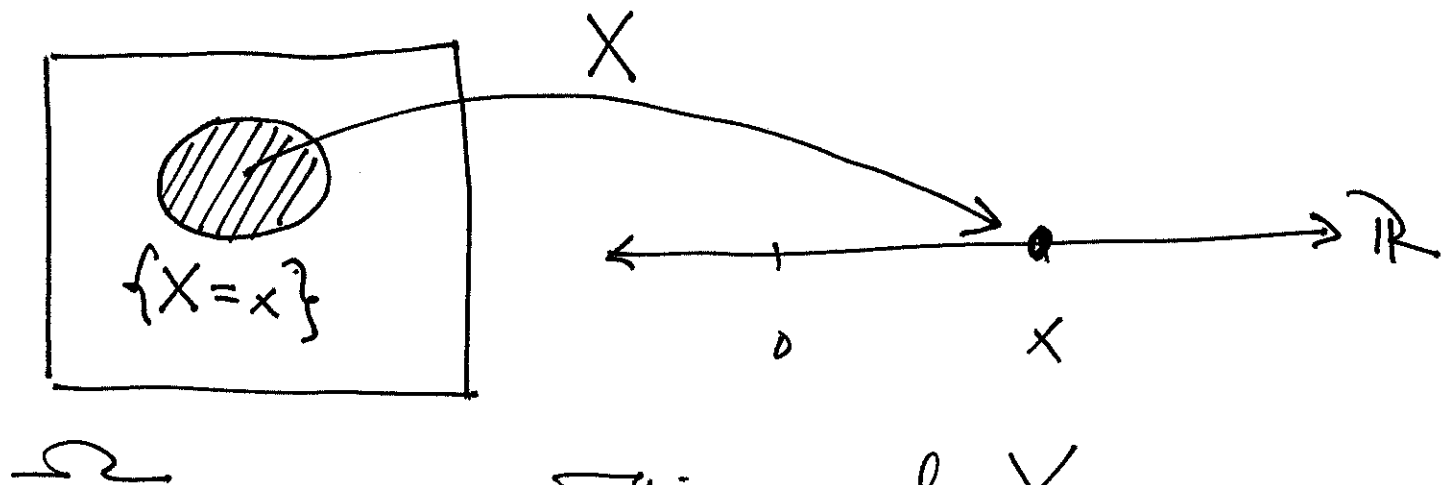


notation:

$$X(w) = x$$

• notation:

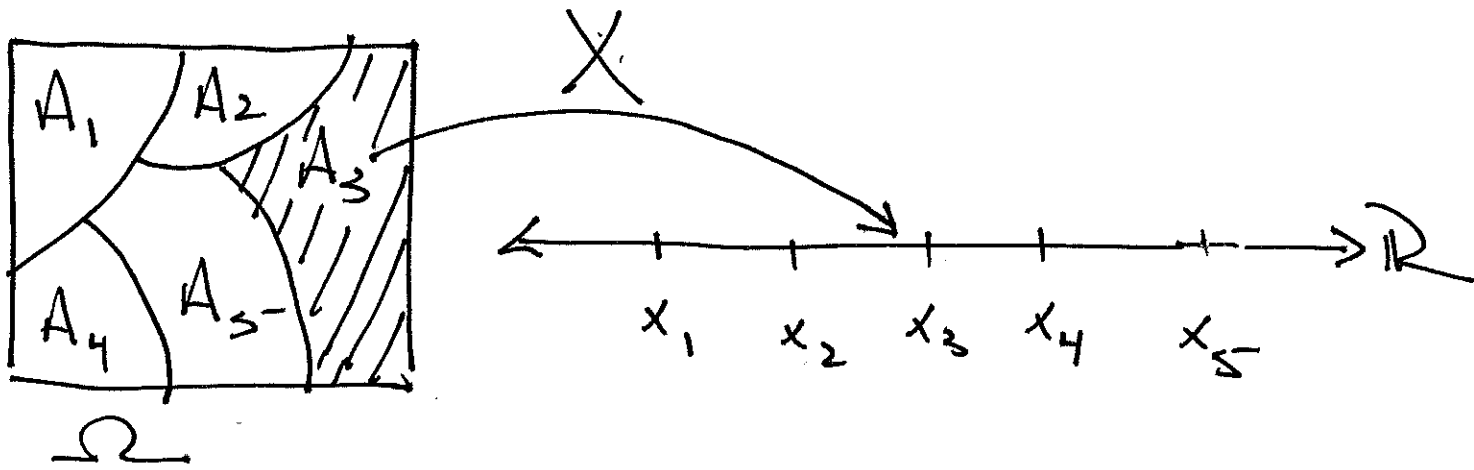
$$\{X=x\} = \{\omega \in \Omega \mid X(\omega) = x\}$$



Think of X as
transferring 'mass'.

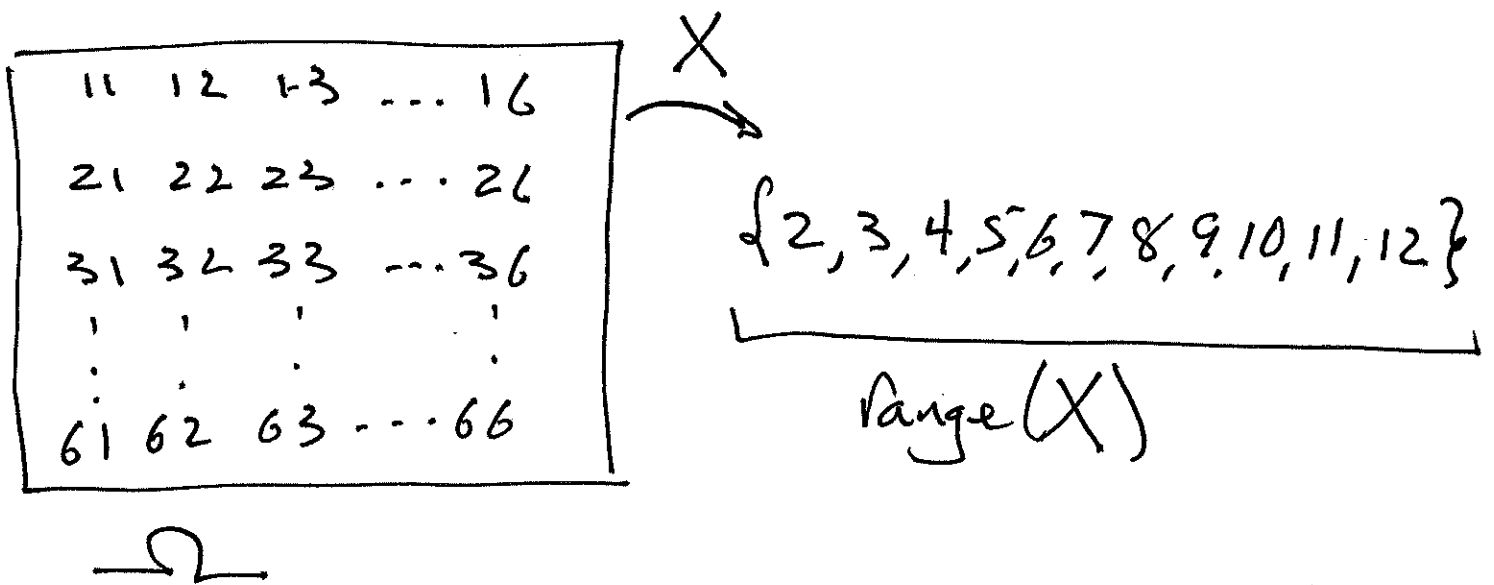
• X is called discrete if its
range

$\text{range}(X) = \{X(\omega) \mid \omega \in \Omega\} \subseteq \mathbb{R}$
is a discrete subset of $\mathbb{R}$.

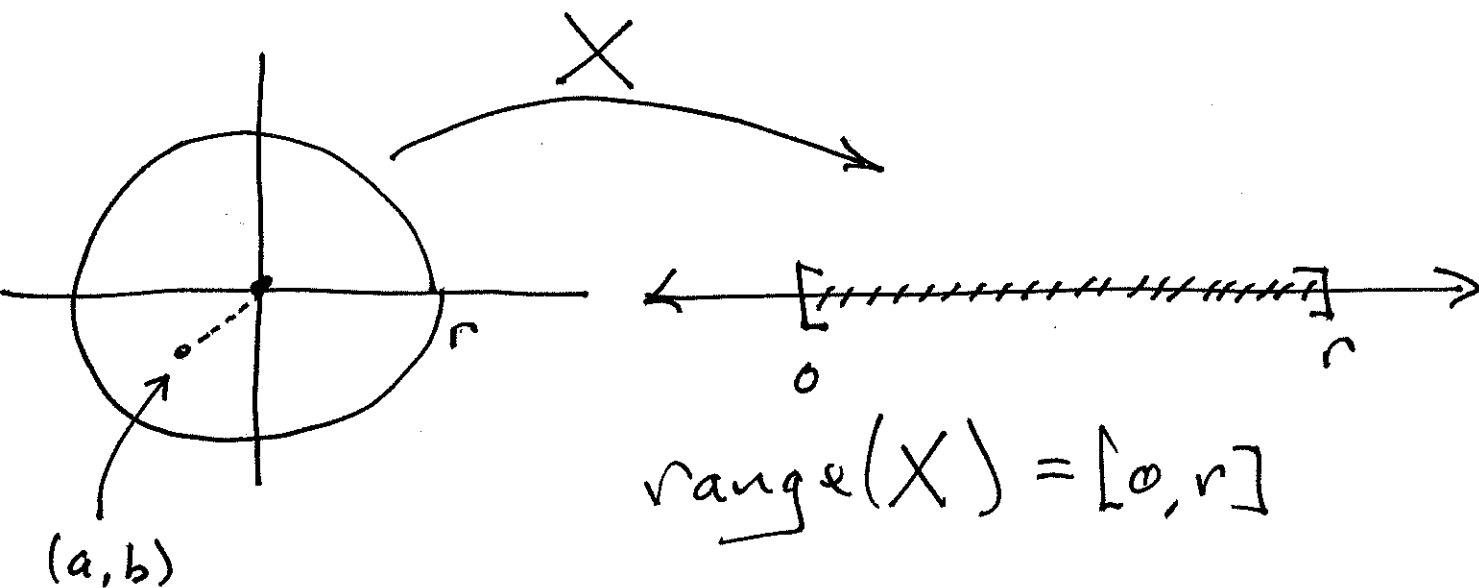


$$A_i = \{X = x_i\} \quad (i=1, 2, 3, 4, 5)$$

Ex. Roll 2 fair independent dice.
let $X((i,j)) = i+j$.



Ex. throw a dart at circular target centered at $(0,0)$
 let $X(a,b) = \sqrt{a^2 + b^2}$



This is a continuous r.v.

2.2 Probability mass function (PMF)

Defn

let X be a discrete r.v. The

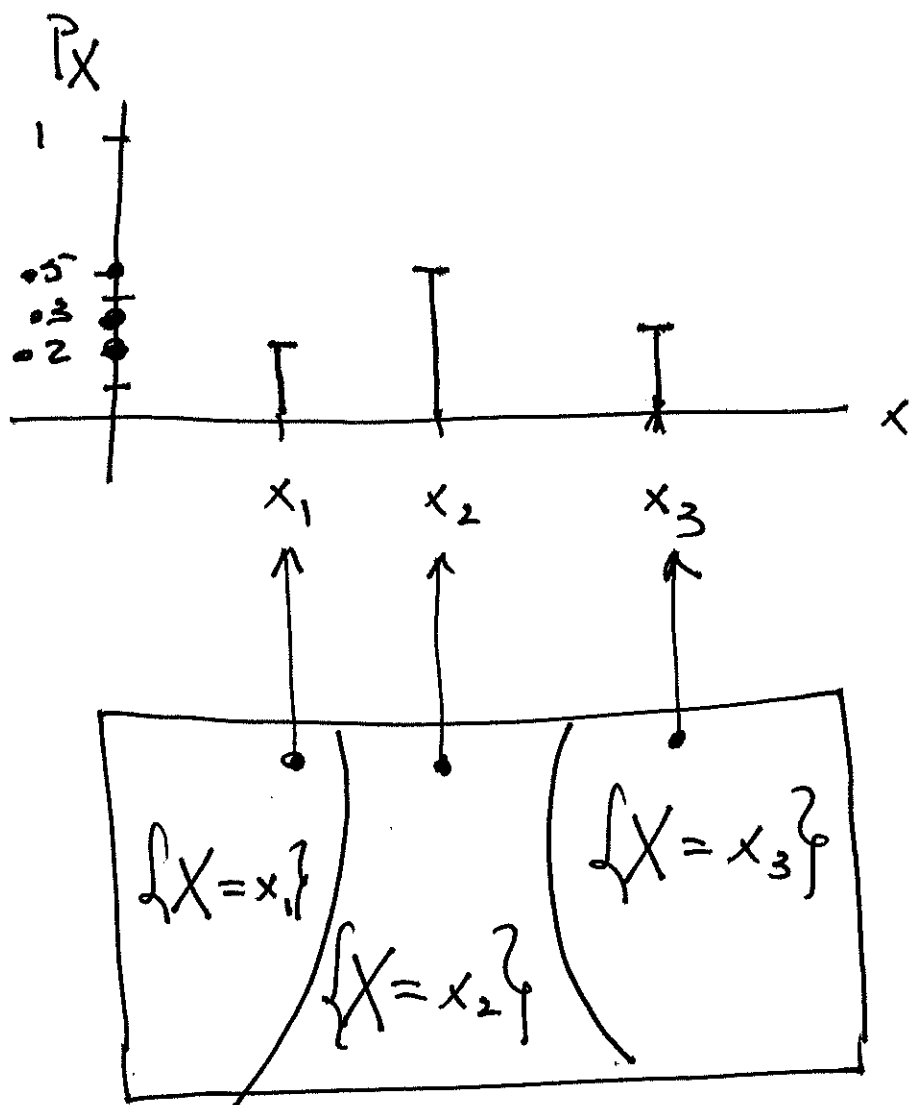
Probability Mass Function (PMF)

of X is

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

observe $P_X: \mathbb{R} \rightarrow [0,1]$, also note

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

$\exists x.$ 

$$P_X(x) = \begin{cases} 0.2 & \text{if } x = x_1 \\ 0.5 & \text{if } x = x_2 \\ 0.3 & \text{if } x = x_3 \\ 0 & \text{otherwise} \end{cases}$$

Bernoulli r.v.

Perform 1 Bernoulli trial. let

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

Its PMF is

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

where $p = P(\text{success})$, is called the parameter of the r.v. X .

Binomial Random Variable

Perform a sequence of independent Bernoulli trials of parameter p .

Let

$$X(\omega) = \# \text{ of successes in } n \text{ trials}$$

we get Binomial Probabilities

$$P_X(k) = P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$$

for $k=0, 1, 2, \dots, n$, and 0 otherwise.

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

✓

The Geometric Random Variable

Perform independent Bernoulli trials (param. p) until 1st success.
Let

$$X(\omega) = \# \text{ of trials until } 1^{\text{st}} \text{ success}$$

note: $\Omega = \{1, 01, 001, 0001, \dots, 00\dots0\dots\}$

$$X(\omega) = \text{length}(\omega).$$

$$\text{Range}(X) = \{1, 2, 3, \dots\} \cup \{\infty\}$$

Recall

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

is the Prob. of getting a seq.

of $k-1$ failures followed by 1 success.

Thus

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

for $k=1, 2, 3, \dots$, and 0 otherwise.

Note

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

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

$$= p \cdot \frac{1}{p} = \boxed{1} \quad \checkmark$$

The Poisson Random Variable

A Poisson r.v. has PMF

$$P_X(k) = e^{-\lambda} \cdot \frac{\lambda^k}{k!} \quad (\text{for } k=0, 1, 2, \dots)$$

and 0 otherwise. (here $\lambda > 0$ is the parameter)

observe

$$\begin{aligned} \sum_{k=0}^{\infty} e^{-\lambda} \cdot \frac{\lambda^k}{k!} &= e^{-\lambda} \cdot \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} \\ &= e^{-\lambda} \cdot e^{\lambda} \\ &= e^{-\lambda + \lambda} = e^0 = 1 \end{aligned}$$

Note: Poisson is the continuous time limit of Binomial.

We have the approximation

$$e^{-\lambda} \cdot \frac{\lambda^k}{k!} \approx \binom{n}{k} p^k (1-p)^{n-k} \quad (k=0, \dots, n)$$

where $\lambda = np$, n large, p small.

more precisely let $\lambda > 0$ be constant.
let $p = \frac{\lambda}{n}$, then

$$\lim_{n \rightarrow \infty} \binom{n}{k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k} = e^{-\lambda} \cdot \frac{\lambda^k}{k!}$$