

Mathematical Statistics

Lecture 1: Probability review, part 1

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Basic objects of probability:

Sample space S \leftarrow (results of some experiment)

Probability P assigns a real number in $[0, 1]$

to certain subsets $E \subset S$ $P(E)$

Subsets $E \subset S$ \leftrightarrow "event" = collection of potential outcomes

$P(E)$ = probability of obtaining an outcome in E

Axioms

$$P(S) = 1$$

$$P(\emptyset) = 0$$

if $E_i \subset S$

$E_i \cap E_j = \emptyset$ (mutually exclusive)

then

$$P\left(\bigcup_i E_i\right) = \sum_i P(E_i)$$

that's it.

Useful notion:

we say $E, F \subset S$ are independent

$$\text{if } P(E \cap F) = P(E)P(F)$$

Random variables

Def A random variable is a function $X: S \rightarrow \mathbb{R}$

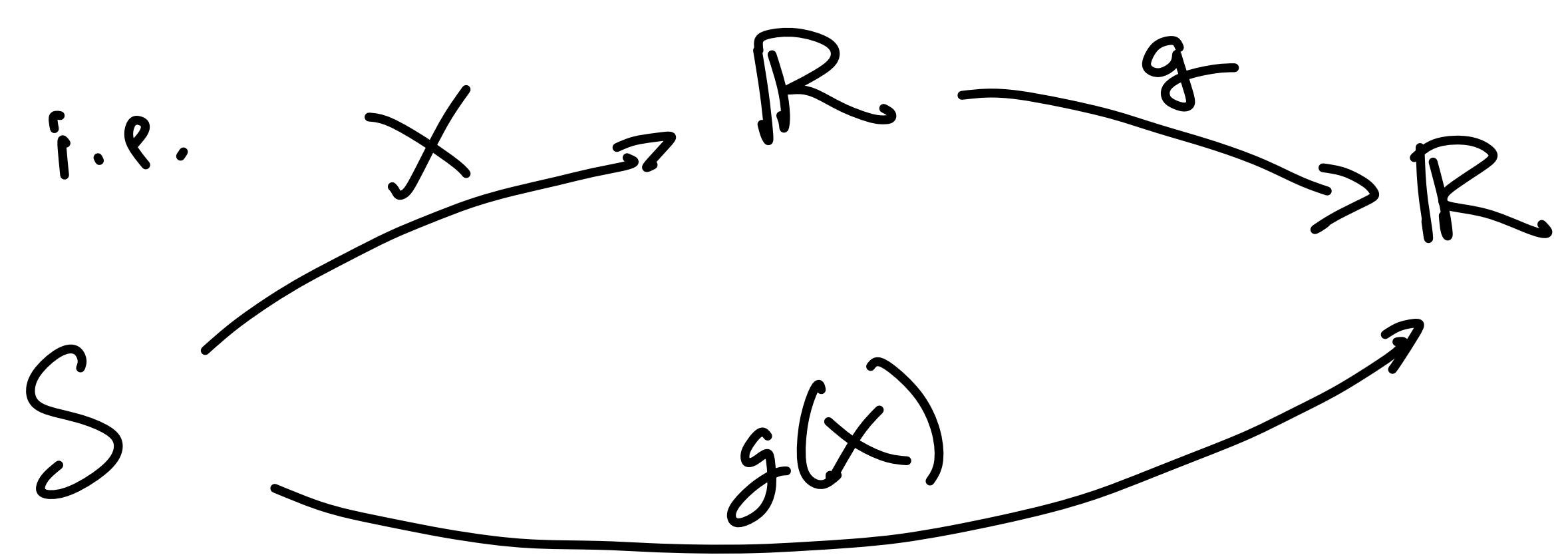
ex. $S =$ person chosen at random $X =$ ht of a person

Notation: $P(X \leq x) = P(\{s \in S \mid X(s) \leq x\})$

Similarly $P(x \leq X)$ $P(a \leq X \leq b)$ etc.

Functions of a random variable

Def: if $g: \mathbb{R} \rightarrow \mathbb{R}$ we make $g(X)$ for the new random variable $g \circ X$



How to encode the information of a random variable?

Cumulative distribution function (c.d.f.): $F(x) \equiv P(X \leq x)$

(Discrete) Probability distribution: $p(x) \equiv P(X=x)$

(Continuous) Probability density function (p.d.f.) is an $f(x)$
such that $P(a \leq X \leq b) = \int_a^b f(x) dx$

(Other way to encode: Moments)

First: Expectation

$$E[X] = \sum x p(x)$$

discrete case

$$\int x f(x) dx$$

continuous case

Properties: $E[X+Y] = E[X] + E[Y]$

$$E[\lambda X] = \lambda E[X]$$

Expectation of fcn of random var.

Fact: $E[g(x)] = \sum g(x)f(x)$ Discrete

$= \int g(x)f(x)dx$ Continuous

Moments

Def the i^{th} moment of X about a is
$$E[(X-a)^i]$$

Ex: $X \rightarrow f(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{else} \end{cases}$

0th moment anywhere = $E[(X-a)^0] = E[1] = 1$

1st moment about 0 = $E[X] = \int_0^1 x \cdot 1 = \frac{1}{2}$

2nd moment about 1 = $E[(X-1)^2] = \int_0^1 (x-1)^2 dx = \dots = \frac{1}{3}$

Moments encode very important info

Def $\mu = \text{mean} = E[X]$

Def $\sigma^2 = \text{variance} = E[(X - \mu)^2]$

Def $\mu_i = i^{\text{th}}$ moment about the mean $(\mu_2 = \sigma^2)$

Def $\mu'_i = i^{\text{th}}$ moment about 0 $(\mu'_1 = \mu = E[X])$

Fact: (Usually), X is determined by its moments about 0

$\mu_0^1, \mu_1^1, \mu_2^1, \dots$ a list of #'s.

Useful to "package" these into a power series

$$M_X(t) = \mu_0^1 + \mu_1^1 t + \mu_2^1 \frac{t^2}{2} + \mu_3^1 \frac{t^3}{3!} + \mu_4^1 \frac{t^4}{4!} + \dots$$

"moment generating function"

$$\text{So } \mu_i = \left. \frac{d^i}{dt^i} M_X(t) \right|_{t=0} .$$

$\Rightarrow X$ determined by $M_X(t)$.

Alternate descriptor of $M_X(t) = E[e^{tX}]$

Nice property $M_{X+Y}(t) = M_X(t) M_Y(t)$

Very useful tool for identifying random vars.

END