

CONTINUOUS RANDOM VARIABLES

Introduction

Continuous random variables

- A continuous random variable is a random variable that takes on an uncountably infinite number of possible values.
- Instead of speaking of computing the probability of specific values (as we do with discrete random variables) we speak of computing the probability of intervals.
- Examples:
 - Weight —
 - Blood pressure —
 - Viral load —

$$P(50 < Y < 60)$$

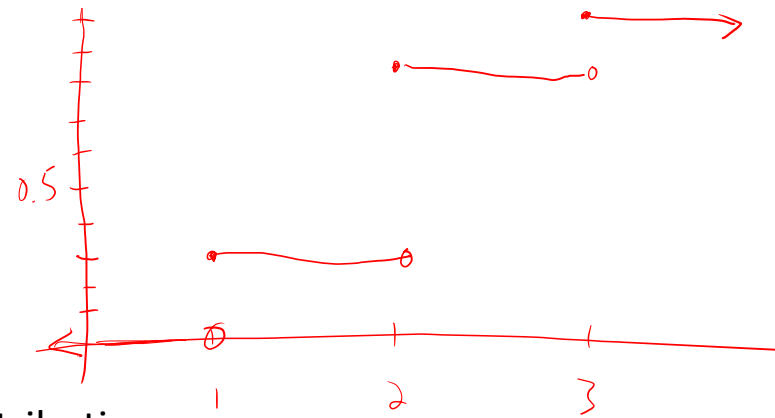
Cumulative distribution function

- Definition 1: The CUMULATIVE DISTRIBUTION FUNCTION (cdf) of a random variable Y , denoted $F(y)$ is

$$F(y) = P(Y \leq y), -\infty < y < \infty$$

- Both discrete and continuous random variables have cdfs.

Cumulative distribution function



- Example 1: A (discrete) random variable Y has distribution

y	1	2	3
$p(y)$	0.3	0.5	0.2

Find the cdf of Y and sketch its graph.

$$F(y) = P(Y \leq y)$$

$$P(Y \leq y) \text{ if } y < 1 \text{ is } 0$$

$$F(y) = \begin{cases} 0 & y < 1 \\ 0.3 & 1 \leq y < 2 \\ 0.8 & 2 \leq y < 3 \\ 1 & y \geq 3 \end{cases}$$

$$P(Y \leq 1) = P(Y = 1) = 0.3$$

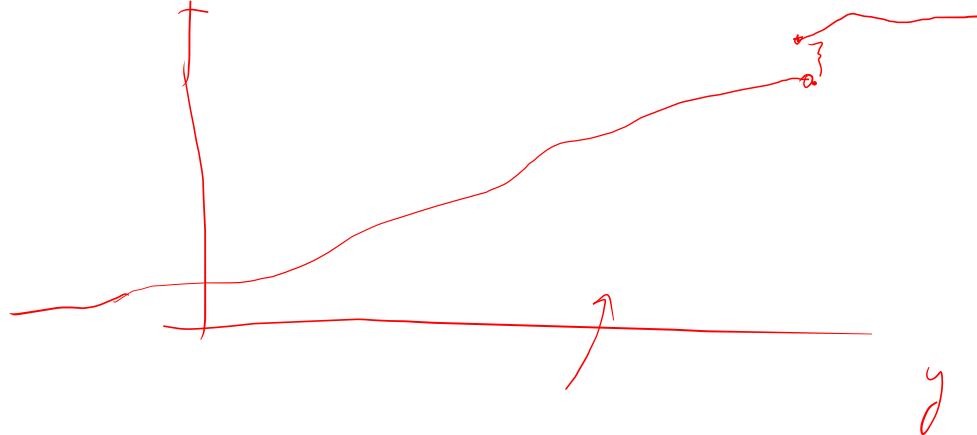
$$P(Y \leq y), 1 < y < 2 = 0.3$$

$$P(Y \leq 2) = P(Y = 1) + P(Y = 2) = 0.3 + 0.5 = 0.8$$

$$P(Y \leq y), 2 < y < 3 = 0.8$$

$$P(Y \leq 3) = 1 \quad P(Y \leq y) \text{ if } y > 3 \text{ is } 1$$

Distribution functions



- Properties of (cumulative) distribution functions:

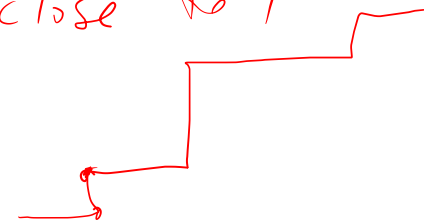
- $F(-\infty) = \lim_{y \rightarrow -\infty} F(y) = 0$ $P(Y \leq -1000000)$ small

- $F(\infty) = \lim_{y \rightarrow \infty} F(y) = 1$

$P(Y \leq 1000000)$ close to 1

- $F(y)$ is a nondecreasing function of y .

- The distribution function of a discrete random variable is a step function.



- Definition 2: A random variable is a CONTINUOUS RANDOM VARIABLE if its cdf is continuous (no steps!).



Probability mass functions, probability density functions

- A discrete random variable has a probability mass function – it puts mass on a countable number of values.
- A continuous random variable doesn't put mass on any value – instead, it has a *probability density function*.
- Definition 3: The PROBABILITY DENSITY FUNCTION of a random variable Y with distribution function $F(y)$ is

$$f(y) = \frac{d}{dy} F(y)$$

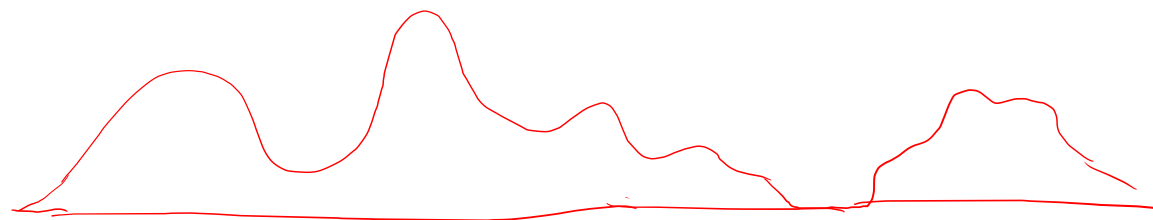
- Therefore, $P(Y \leq y) = \int_{-\infty}^y f(t) dt$.

$F(y) =$

FTOC

$$\sum \underline{P(y_i)} = 1$$

Density functions

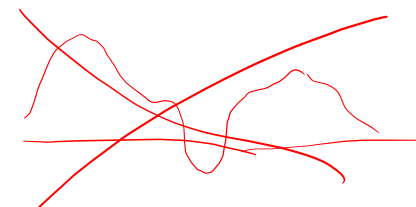


• Properties of a density function:

- $f(y) \geq 0$ for all $y, -\infty < y < \infty$ ✓

- $\int_{-\infty}^{\infty} f(y) dy = 1$ *total area* ✓

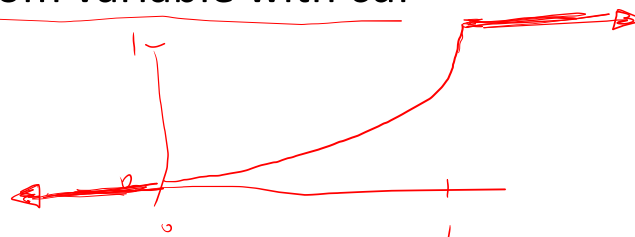
F non decreasing



• Example 2: Suppose Y is a random variable with cdf

($-\infty, \infty$)

$$F(y) = \begin{cases} 0, & y < 0 \\ y^3, & 0 \leq y < 1 \\ 1, & y \geq 1 \end{cases}$$



Find the pdf of Y .

$$f(y) = \frac{d}{dy} F(y) = \begin{cases} 0 & y < 0 \\ 3y^2 & 0 \leq y \leq 1 \\ 0 & y > 1 \end{cases}$$



Area under curve = 1 ?? check

Calculating probabilities of intervals

- With a continuous random variable Y , we can compute the probability that Y falls into a particular interval $[a, b]$, i.e., $P(a \leq Y \leq b)$.
- Theorem 1: If Y is a continuous random variable with pdf f , then for $a < b$,

$$P(a \leq Y \leq b) = \int_a^b f(y) dy$$

Proof $P(a \leq Y \leq b) = P(Y \leq b) - P(Y \leq a)$

$$= F(b) - F(a)$$

$$= \int_{-\infty}^b f(y) dy - \int_{-\infty}^a f(y) dy$$

$$= \int_{-\infty}^a f(y) dy + \int_a^b f(y) dy - \int_{-\infty}^a f(y) dy \quad \square$$

Calculating probabilities of intervals

$$f(y) = 3y^2, \quad 0 < y < 1$$

- Example 2 (continued): for the previous example, calculate.

$$\underline{P(0.4 \leq Y \leq 0.8)}$$

$$\int_{0.4}^{0.8} 3y^2 dy = y^3 \Big|_{0.4}^{0.8} = \underbrace{(0.8)^3}_{F(0.8)} - \underbrace{(0.4)^3}_{F(0.4)} = \dots = 0.448$$

$$F(y) = y^3 \quad 0 < y < 1$$

✗ Important for discrete RVs

Some notes about continuous random variables

(a, b) $[a, b)$ $(a, b]$ $[a, b]$

- The same way, we can compute probabilities of “one-sided” intervals, e.g., $[a, \infty)$, $(-\infty, b]$.

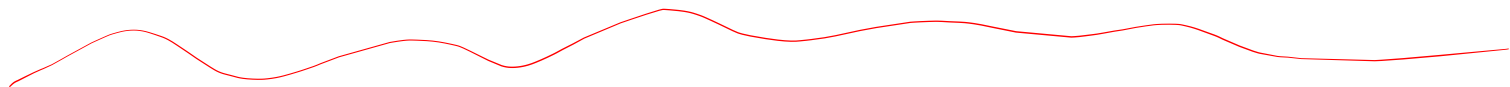
- With continuous random variables, we don't have to worry about “open” or “closed” intervals, i.e.,

$$\underline{P(a \leq Y \leq b) = P(a \leq Y < b) = P(a < Y \leq b) = P(a < Y < b)}$$

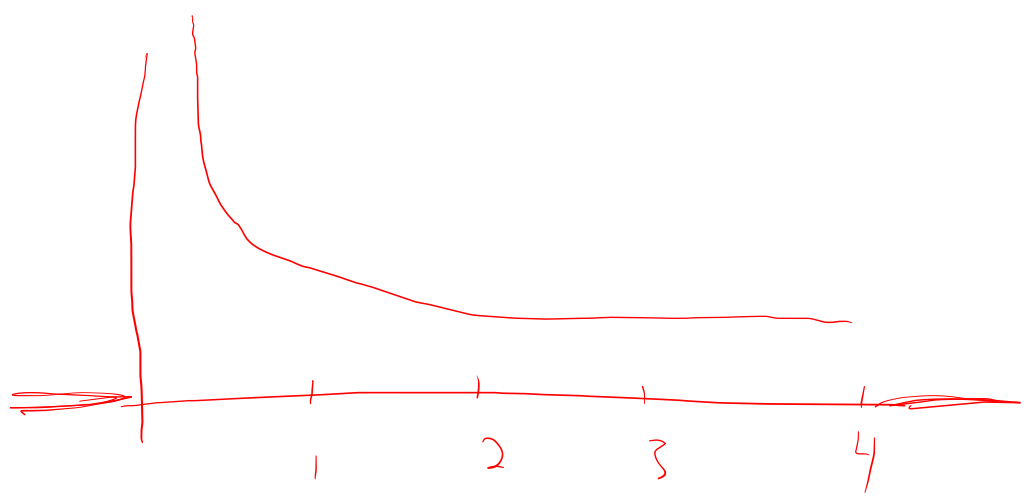


- So for a continuous random variable, $P(Y = a) = \text{????}$

$$P(Y = a) = P(a \leq Y \leq a) = \int_a^a f(y) dy = 0$$



Another example



- Example 3: Suppose Y is a (continuous) random variable with pdf

$$f(y) = \begin{cases} cy^{-1/2}, & 0 < y \leq 4 \\ 0, & \text{otherwise} \end{cases}$$

Find c , find the distribution function F , and calculate $P(Y < 1)$.

Total prob must be 1

$$\int_0^4 f(y) dy = 1 \quad \int_0^4 cy^{-1/2} = c \left(2y^{1/2} \Big|_0^4 \right) = 1 \quad \text{has to}$$

$$F(y) = \int_{-\infty}^y \frac{1}{4} t^{-1/2} dt = \frac{1}{4} \int_0^y t^{-1/2} dt = \frac{1}{4} \left(2t^{1/2} \Big|_0^y \right) = \frac{1}{4} (2\sqrt{y} - 0) = \frac{1}{2} \sqrt{y}$$

$c = \frac{1}{4}$

$$F(y) = \begin{cases} 0 & y < 0 \\ \frac{1}{2} \sqrt{y} & 0 \leq y \leq 4 \\ 1 & y > 4 \end{cases}$$

$$P(Y < 1) = F(1) = \frac{1}{2} \sqrt{1} = \boxed{\frac{1}{2}} \quad 0 < y < 4$$