

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 —

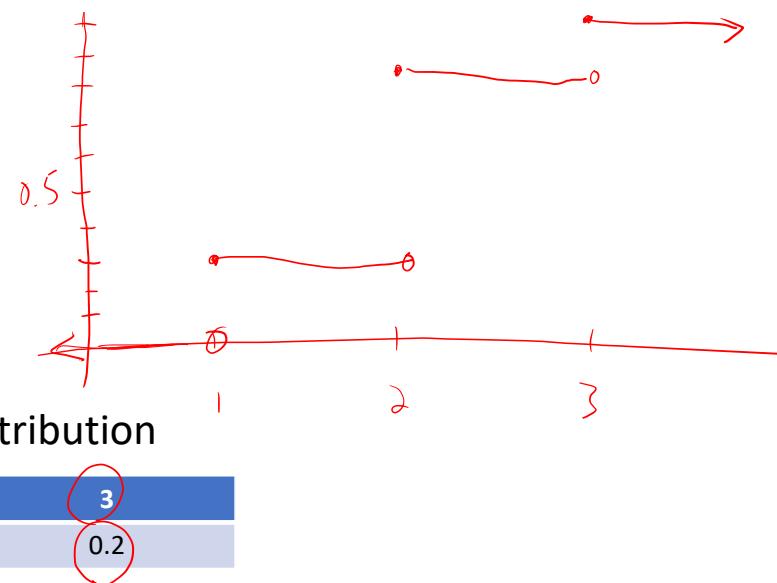
$$P(50 < Y < 60)$$

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

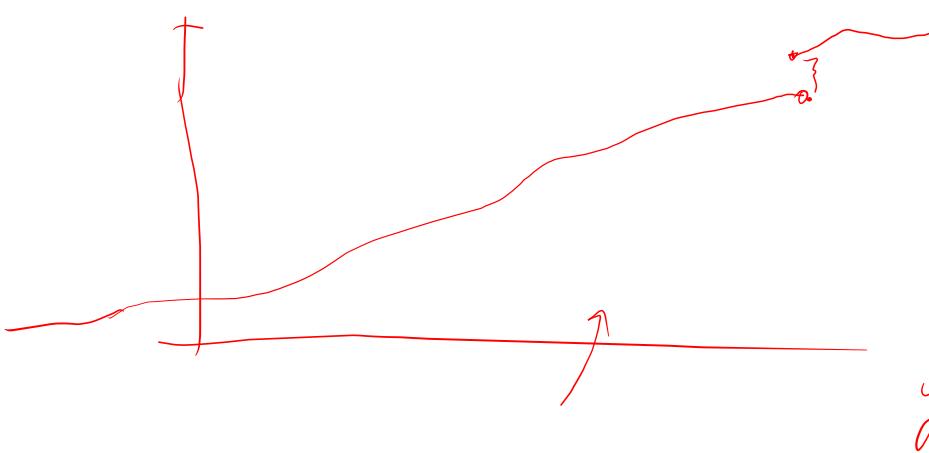
Find the cdf of Y and sketch its graph.

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

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

$$\begin{aligned}
 P(Y \leq 1) &= P(Y = 1) = 0.3 \\
 P(Y \leq y), \quad 1 < y < 2 &= 0.3 \\
 P(Y \leq 2) &= P(Y = 1) + P(Y = 2) = 0.3 + 0.5 = 0.8 \\
 P(Y \leq y), \quad 2 < y < 3 &= 0.8 \\
 P(Y \leq y) &= 1 \quad P(Y \leq y), \quad y > 3 \quad \text{is } 1
 \end{aligned}$$

Distribution functions



- Properties of (cumulative) distribution functions:

- $F(-\infty) = \lim_{y \rightarrow -\infty} F(y) = 0$

$P(Y \leq -1000000)$ small

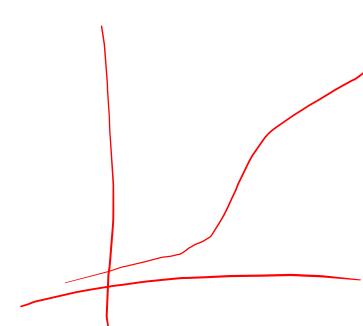
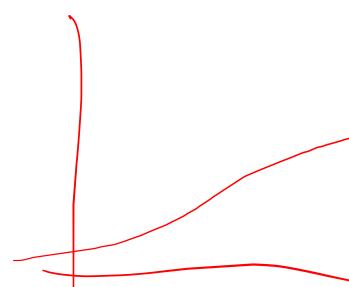
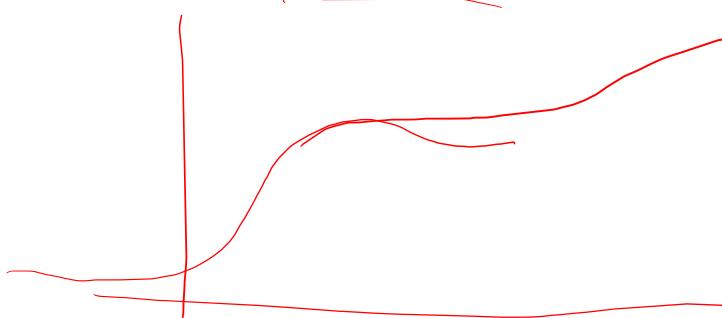
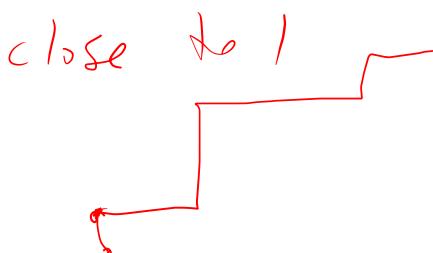
- $F(\infty) = \lim_{y \rightarrow \infty} F(y) = 1$

$P(Y \leq 1000000)$ 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

$$\sum p(y_i) = 1$$

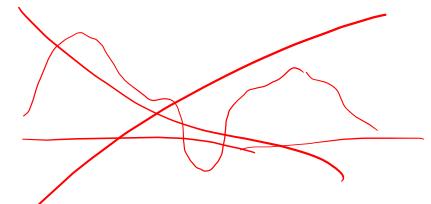
Density functions



- Properties of a density function:

- $f(y) \geq 0$ for all $y, -\infty < y < \infty$
- $\int_{-\infty}^{\infty} f(y) dy = 1$ total area

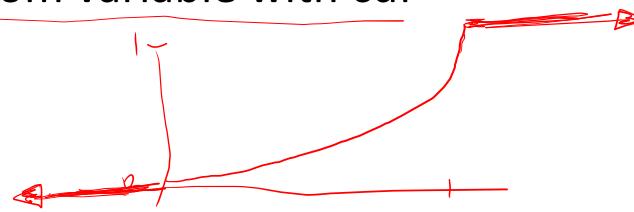
F non decreasing



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

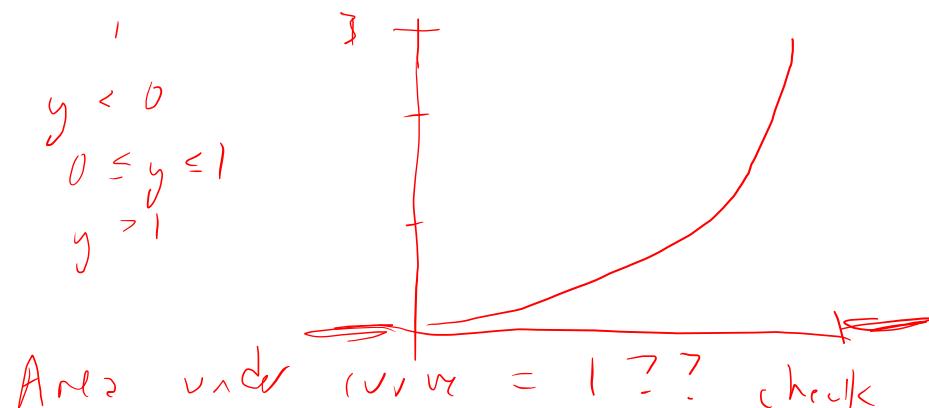
$(-\infty, \infty)$

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



Find the pdf of Y .

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



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., $\underline{P(a \leq Y \leq b)}$.
- Theorem 1: If Y is a continuous random variable with pdf f , then for $a < b$,

Proof

$$\begin{aligned} P(a \leq Y \leq b) &= P(Y \leq b) - P(Y \leq a) \\ &= F(b) - F(a) \\ &= \int_a^b f(y) dy - \int_{-\infty}^a f(y) dy \\ &= \int_{-\infty}^a f(y) dy + \int_a^b f(y) dy - \int_{-\infty}^a f(y) dy \end{aligned}$$

□

Calculating probabilities of intervals

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

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

$$P(0.4 \leq Y \leq 0.8)$$

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

$$F(y) = \frac{3}{4} y^4, \quad 0 < y < 1$$


 important
 discrete
 for
 RVs

Some notes about continuous random variables

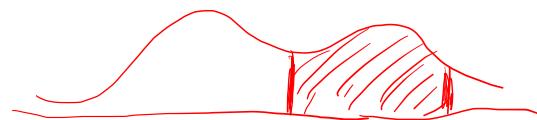
$$(a, b) \quad [a, b) \quad (a, b) \quad [a, b)$$

\uparrow

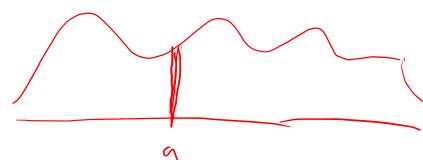
- The same way, we can compute probabilities of “one-sided” intervals, e.g., $[a, \infty)$, $(-\infty, b]$.

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

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



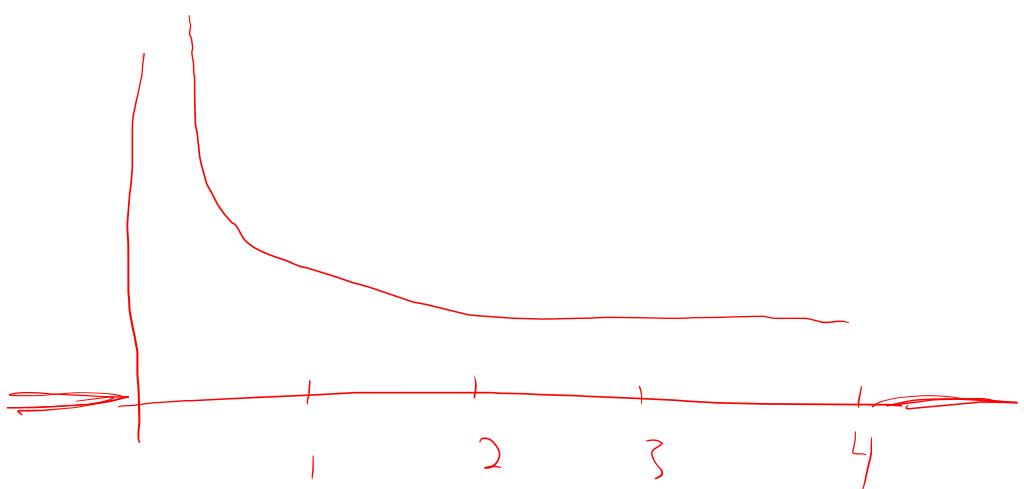
- So for a continuous random variable, $P(Y = a) = ????$



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



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

$$\int_0^4 f(y) dy = 1 \quad \int_0^4 cy^{-1/2} dy = c \left(2y^{1/2} \right) \Big|_0^4 = c(2\sqrt{4} - 2 \cdot 0) = c \cdot 4 = 1 \quad c = \frac{1}{4}$$

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

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

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