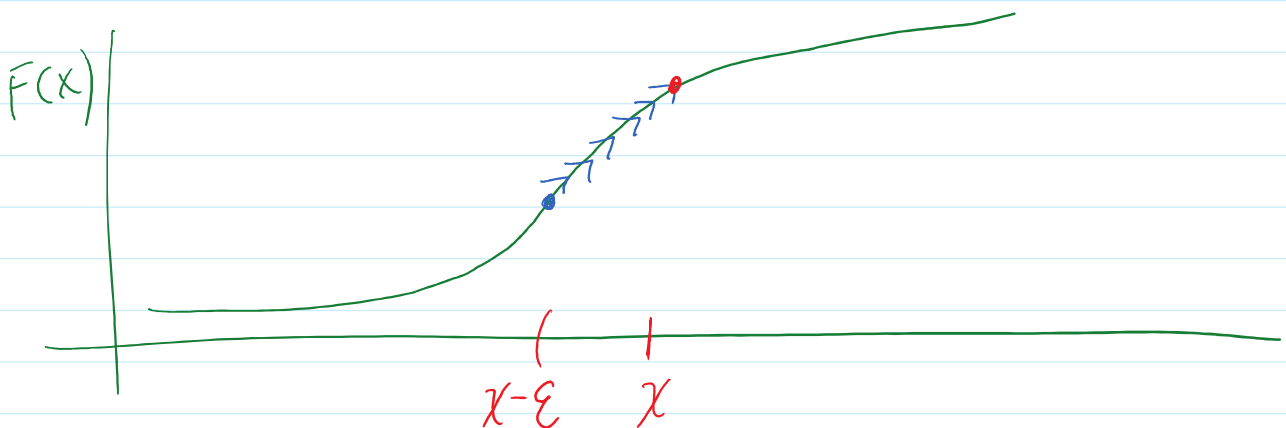


$$\begin{aligned}
 \lim_{\epsilon \rightarrow 0} P(x-\epsilon < X \leq x) &= \lim_{\epsilon \rightarrow 0} (F(x) - F(x-\epsilon)) \\
 &= F(x) - \lim_{\epsilon \rightarrow 0} F(x-\epsilon) \\
 &= \text{jump size} = f(x)
 \end{aligned}$$

Continuous Case:



$$\lim_{\epsilon \rightarrow 0} P(x-\epsilon < X \leq x) = \dots$$

$$= F(x) - \lim_{\varepsilon \rightarrow 0} F(x - \varepsilon)$$

$$= F(x) - F(x) = 0$$

Punchline: PMF formulation not as useful for cts RVs

Want: something for cts RVs that behaves like PMF

$$F(x) = \sum_{i \leq x} f(i)$$

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Defn: Probability Density Function (PDF)

Continuous analog of PMF.

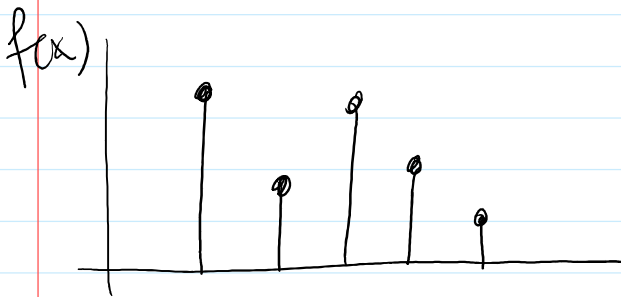
For a cts RV  $X$  the PDF is a function  $f: \mathbb{R} \rightarrow \mathbb{R}$  so that  $\forall x \in \mathbb{R}$

$$F(x) = \int_{-\infty}^x f(t) dt$$

Note: by Fund. Thm. of Calc.

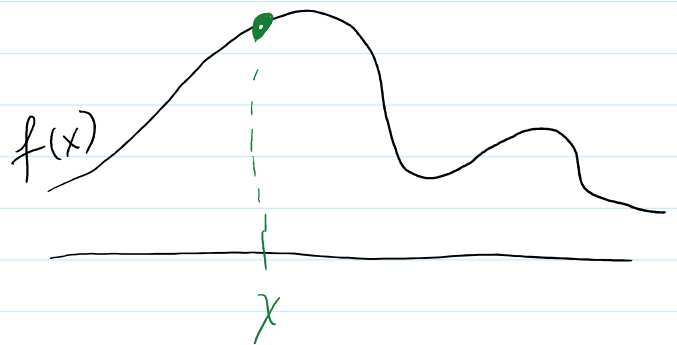
$$\frac{dF}{dx} = \frac{d}{dx} \int_{-\infty}^x f(t) dt = f(x)$$

discrete PMF



$$f(x) = P(X=x)$$

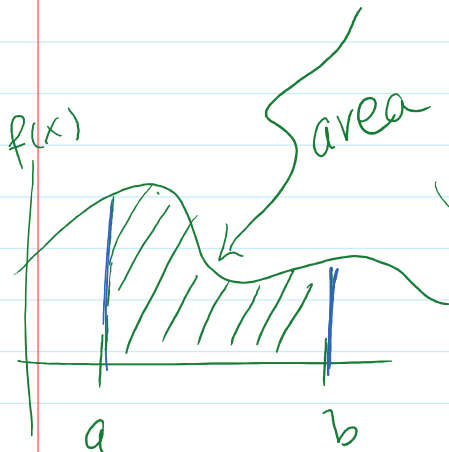
Continuous PDF



$$f(x) \neq P(X=x)$$

Properties:

$$P(a < X \leq b) = F(b) - F(a)$$



$$= \int_{-\infty}^b f(t) dt - \int_{-\infty}^a f(t) dt$$

$$= \int_a^b f(t) dt$$

Said  $P(X=a) = P(X=b) = 0$

be sloppy w/ end points

$$\left. \begin{aligned} P(a < X \leq b) &= P(a \leq X \leq b) \\ &= P(a \leq X < b) \\ &= P(a < X < b) \end{aligned} \right\} \begin{array}{l} \text{only for} \\ \text{cts} \end{array}$$

More general fact:

$$(\text{discrete}) : P(X \in A) = \sum_{x \in A} f(x)$$

$$(\text{cts}) : P(X \in A) = \int_A f(x) dx$$

Ex.

$$P(X \in (2, 3]) = \int_2^3 f(x) dx$$



$$P(X=2) = P(X \in \{2\}) = \int_2^2 f(x) dx = 0$$



$$F(x) = \frac{1}{1 + e^{-x}}$$

$$f(x) = \frac{dF}{dx} = \dots = \frac{e^{-x}}{(1+e^{-x})^2}$$
$$X \sim U(0, 1)$$
$$f(x) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{else} \end{cases}$$


$$F(x) = \int_{-\infty}^x f(t) dt$$

$$\underline{x < 0} : F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^x 0 dt = 0$$

$$\underline{0 < x < 1} : F(x) = \int_{-\infty}^x f(t) dt = \int_0^x 1 dt = x$$

$$\underline{x > 1} : F(x) = \int_{-\infty}^x f(t) dt = \int_0^1 1 dt = 1$$

$$F(x) = \begin{cases} 0 & x < 0 \\ x & 0 < x < 1 \\ 1 & x > 1 \end{cases}$$

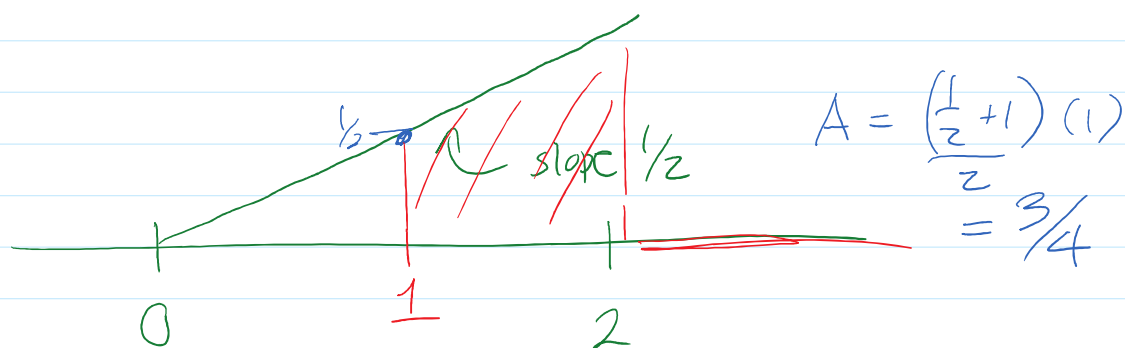
$F(x)$



$E X$

Ex.

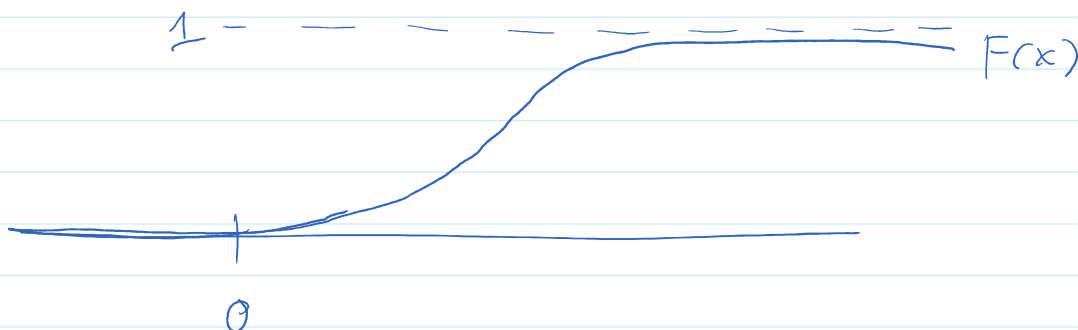
$$f(x) = \begin{cases} x/2 & 0 < x < 2 \\ 0 & \text{else} \end{cases}$$



$$\underline{P(X > 1)} = \int_1^{\infty} f(x) dx = \int_1^2 \frac{x}{2} dx = \frac{x^2}{4} \Big|_1^2 = \frac{4}{4} - \frac{1}{4} = \frac{3}{4}$$

Ex.

$$F(x) = 1 - e^{-x} \text{ for } x > 0$$



Q:  $P(1 < X < 2)$  ?

Way 1:  $P(1 < X < 2) = F(2) - F(1)$   
 $= (1 - e^{-2}) - (1 - e^{-1})$   
 $= e^{-1} - e^{-2}$

Way 2:  $f(x) = \frac{dF}{dx} = \frac{d}{dx}(1 - e^{-x}) = e^{-x}$  for  $x > 0$

$$P(1 < X < 2) = \int_1^2 e^{-x} dx = -e^{-x} \Big|_1^2 = e^{-1} - e^{-2}$$


---

Theorem: PMF/PDF characterization

A function  $f$  is the PMF/PDF of some RV  
iff

①  $f(x) \geq 0 \quad \forall x \in \mathbb{R}$

② (discrete)  $\sum_{x \in \mathbb{R}} f(x) = 1$

(cts)  $\int_{\mathbb{R}} f(x) dx = 1$



Fact: If  $g(x) \geq 0$

$$\text{and } \int_{\mathbb{R}} g(x) dx = C < \infty$$

If  $f(x) = \frac{1}{C} g(x)$  then  $f$  is a PDF.

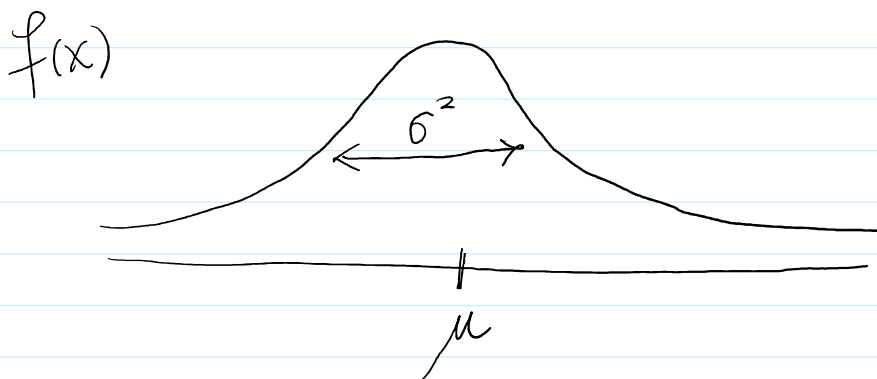
Ex. Normal Distribution (Gaussian)

notation:

$$X \sim N(\mu, \sigma^2)$$

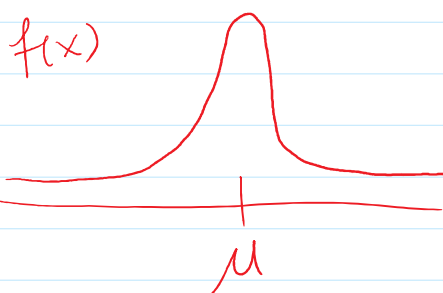
mean:  $\mu \in \mathbb{R}$

Variance:  $\sigma^2 > 0$



Small  $\sigma^2$

large  $\sigma^2$



Special Case: Standard Normal,  $\mu=0, \sigma^2=1$

$$X \sim N(0, 1)$$

$$f(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}x^2\right)$$

$\exp(a) = e^a$

Q: Is this a valid PDF?

①  $f(x) \geq 0 \quad \forall x \quad \checkmark$

②  $\int_{\mathbb{R}} f(x) dx = 1$

$$\int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}x^2\right) dx = 1$$

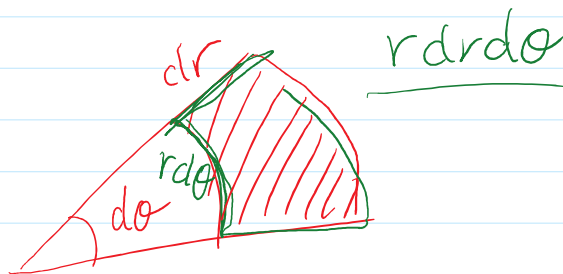
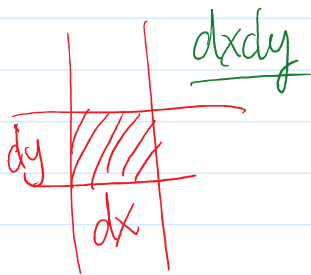
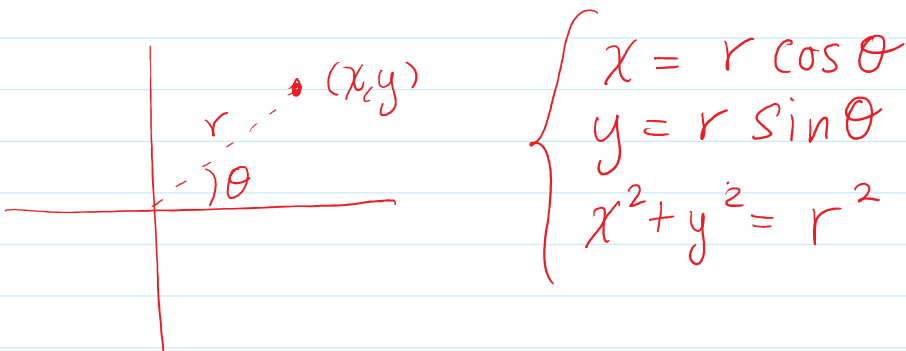
$I$

Want:  $I = 1 \iff \boxed{I^2 = 1}$

$$I^2 = I \cdot I = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy$$

$$\begin{aligned}
 I^2 = I \cdot I &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{1}{2\pi} e^{-\frac{1}{2}x^2} e^{-\frac{1}{2}y^2} dx dy & e^a e^b = e^{a+b} \\
 &= \iint_{\mathbb{R}^2} \frac{1}{2\pi} \exp\left(-\frac{1}{2}(x^2+y^2)\right) dx dy
 \end{aligned}$$

Polar coordinates:



$$\frac{1}{2\pi} \iint \exp\left(-\frac{1}{2}(x^2+y^2)\right) dx dy = \frac{1}{2\pi} \int_{\theta=0}^{2\pi} \int_{r=0}^{\infty} e^{-\frac{1}{2}r^2} r dr d\theta \quad (*)$$

1

Solve inner integral w/  $u$ -substitution

$$u = \frac{1}{2} r^2 \quad ; \quad du = r dr$$

$$\int_0^{\infty} e^{-u} du = -e^{-u} \Big|_0^{\infty} = 0 - (-1) = 1$$

$$\textcircled{*} = \frac{1}{2\pi} \int_0^{2\pi} d\theta = \frac{1}{2\pi} (2\pi) = 1.$$

Expected Value:

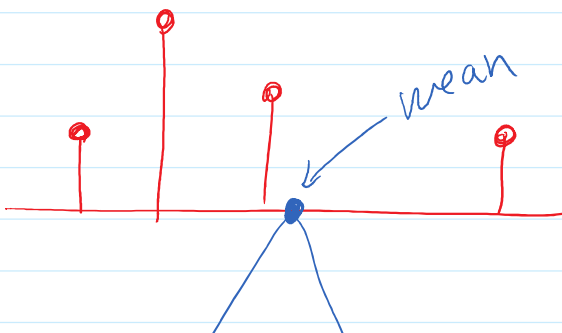
If  $X$  is a RV then the mean or expected value of  $X$

denoted  $E[X]$ ,

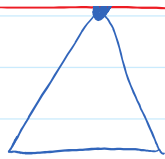
is defined

① discrete  $E[X] = \sum_{x \in \mathcal{R}} x f(x)$

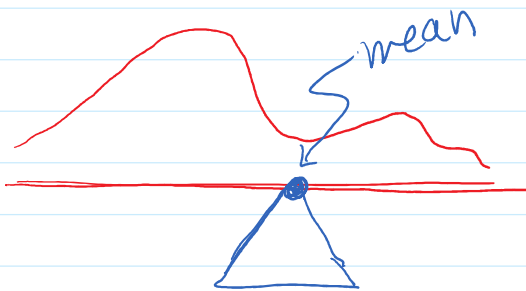
$$= \sum_{x \in \text{Support}(X)} x f(x)$$



weighted  
avg



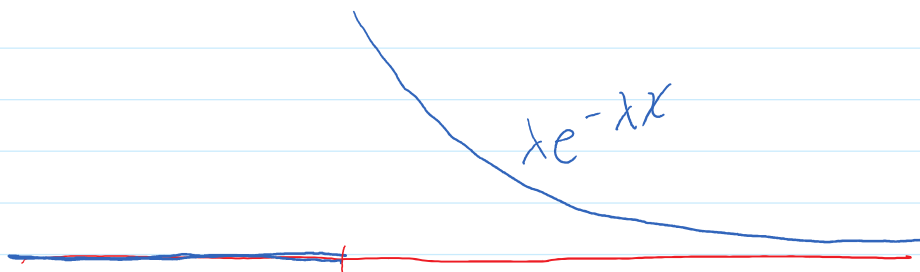
② Continuous :  $E[X] = \int_{\mathbb{R}} x f(x) dx$



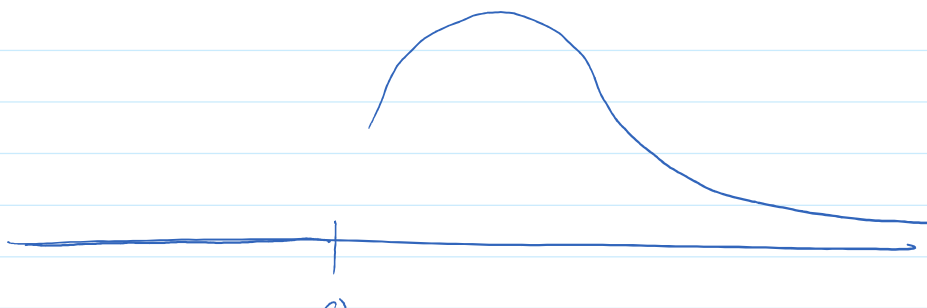
ex. let  $X \sim \text{Exp}(\lambda)$   $\lambda > 0$

↑ Exponential dist.

$$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$$



0



Q: what is  $E[X]$ ?

$$E[X] = \int_{\mathbb{R}} x f(x) dx = \int_0^{\infty} \underline{x} \lambda e^{-\underline{\lambda x}} dx$$

integration by parts?

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