

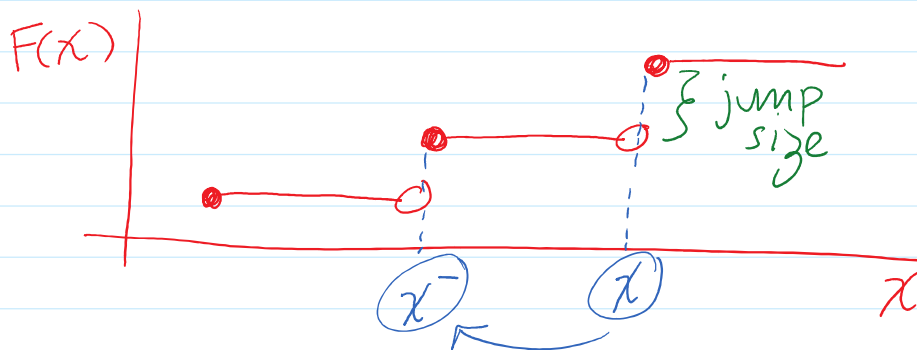
$$\text{PMF: } f(x) = P(X=x)$$

$$\text{CDF: } F(x) = P(X \leq x)$$

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

← from f to F

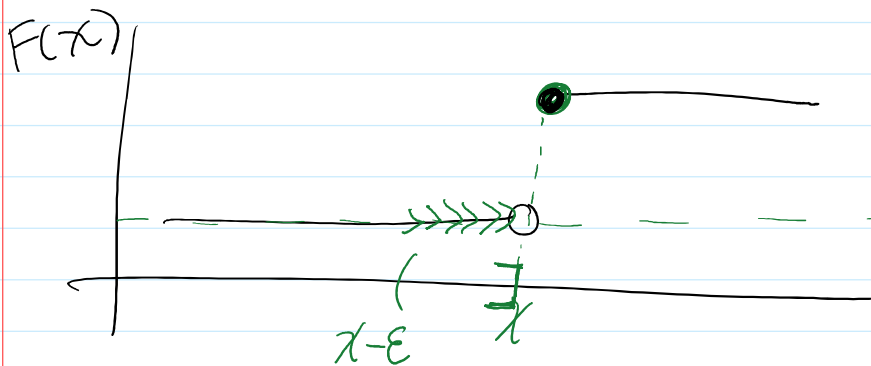
$$P(a < X \leq b) = F(b) - F(a)$$



$$\begin{aligned} P(X=x) &= P(x^- < X \leq x) \\ &= F(x) - F(x^-) \\ &= \text{jump at } x \end{aligned}$$

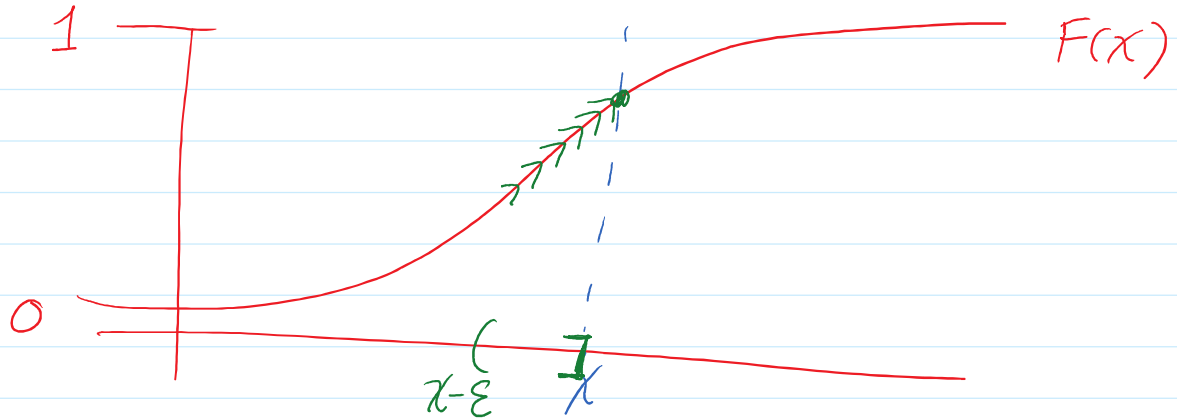
next lowest point below x

from F to f



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

What about continuous RVs?



$$\lim_{\varepsilon \downarrow 0} P(x - \varepsilon < X \leq x) = P(X = x)$$

$$= \lim_{\varepsilon \downarrow 0} \underbrace{F(x) - F(x - \varepsilon)}_{\forall \varepsilon \text{ } F \text{ is continuous}} \quad F(x - \varepsilon) \rightarrow F(x)$$

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

$$= 0$$

Punchline: this PMF formulation isn't useful for cts RVs

Can we create a continuous analog to the PMF?

Want:

$$F(x) = \sum_{i \leq x} f(i) \quad [\text{for discrete}]$$

Defn: Probability Density Function (PDF)

Analog of PMF for cts RVs.

The PDF for a CR RV is a function $f: \mathbb{R} \rightarrow \mathbb{R}$ defined $\forall x \in \mathbb{R}$ as the function for which

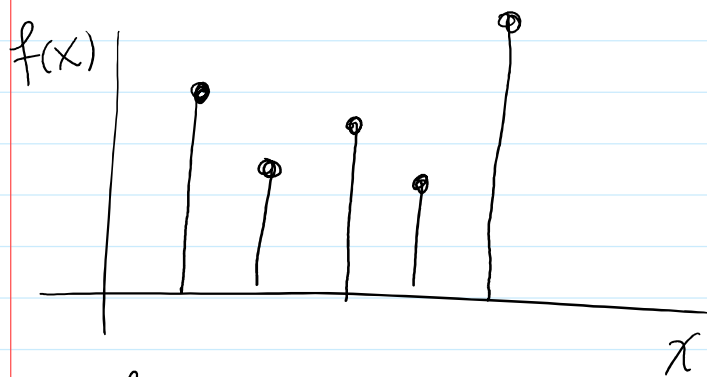
$$F(x) = \int_{-\infty}^x f(t) dt.$$

note by the Fundamental Theorem of Calculus

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

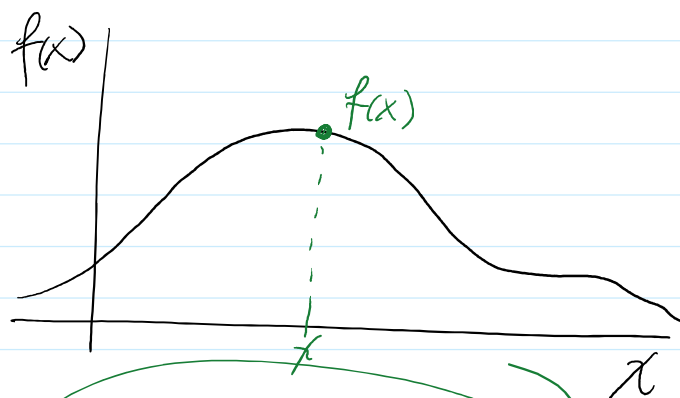
PDF = derivative of the CDF.

discrete PMF



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

continuous PDF



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

Properties of PDF

$$\begin{aligned}
 P(a < X \leq b) &= F(b) - F(a) \\
 &= \int_{-\infty}^b f(t) dt - \int_{-\infty}^a f(t) dt \\
 &= \int_a^b f(t) dt
 \end{aligned}$$

we said $P(X=a) = P(X=b) = 0$

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

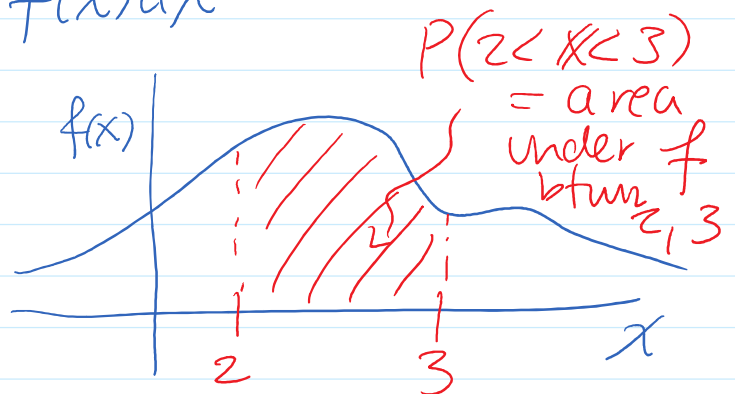
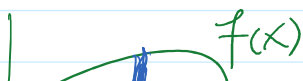
More general fact:

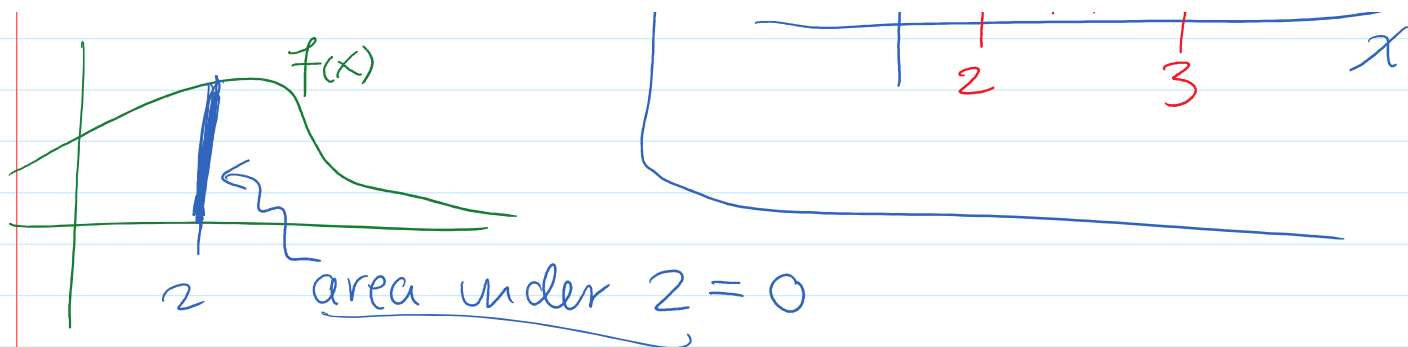
(discrete) : $P(X \in A) = \sum_{x \in A} f(x)$

(continuous) : $P(X \in A) = \int_A f(x) dx$

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

Ex. $P(X \in \{2\})$
 $= P(X=2) = 0$





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

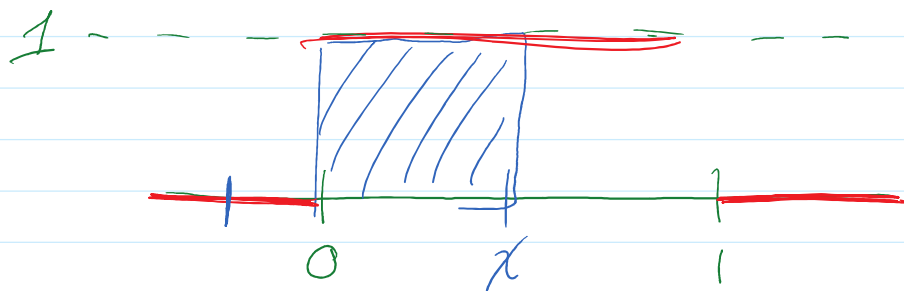
what is the associated PDF?

$$f(x) = \frac{d}{dx} F(x) = \dots = \frac{e^{-x}}{(1+e^{-x})^2}$$

Ex. Continuous Uniform Distribution (on $(0,1)$)

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

means $f(x) = \begin{cases} 1 & , \quad 0 < x < 1 \\ 0 & \text{else} \end{cases}$



What is the CDF? $F(x) = \int_{-\infty}^x f(t) dt$

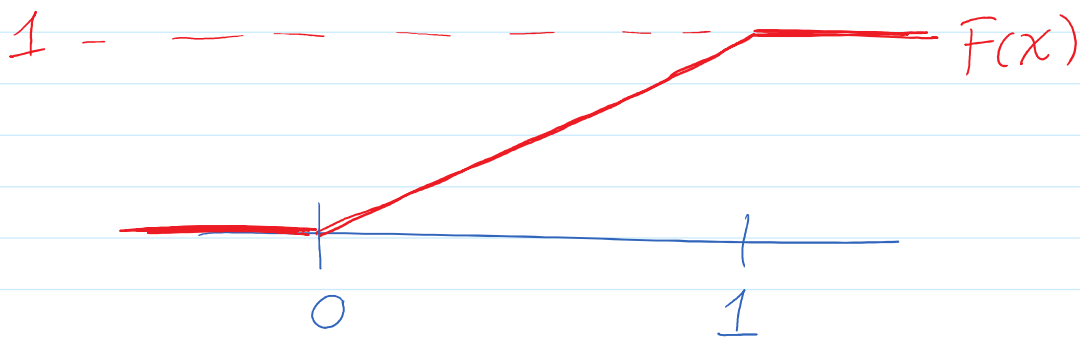
What is the CDF? $F(x) = \int_{-\infty}^x f(t) dt$

If $x < 0$ then $F(x) = \int_{-\infty}^x 0 dt = 0$

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

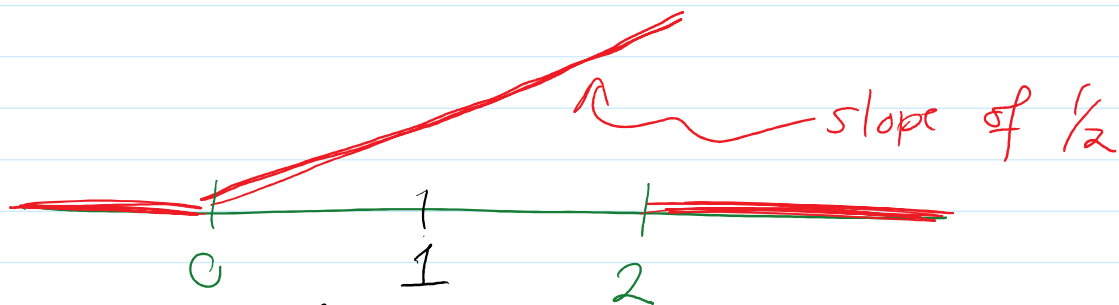
If $x > 1$ then $F(x) = \int_{-\infty}^x f(t) dt = \int_0^1 1 dt = 1$

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



Ex. Let

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

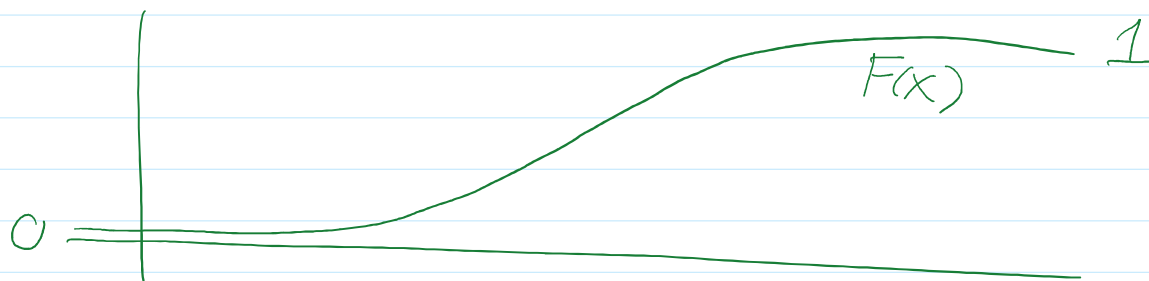


$$P(X > 1) = \int_1^{\infty} f(t) dt = \int_1^2 \frac{t}{2} dt$$

$$= \left. \frac{t^2}{4} \right|_1^2 = \frac{4 - 1}{4} = \frac{3}{4}$$

Ex. Let

$$F(x) = \begin{cases} 1 - e^{-x} & \text{for } x > 0 \\ 0 & \text{else} \end{cases}$$



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

$$\begin{aligned} \text{Way 1: } P(1 < X < 2) &= F(2) - F(1) \\ &= (1 - e^{-2}) - (1 - e^{-1}) \end{aligned}$$

$$= e^{-1} - e^{-2}$$

way 2: $f(x) = \frac{dF}{dx} = \frac{d}{dx} (1 - e^{-x}) = e^{-x}$

ad

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

$$= -e^{-2} - (-e^{-1})$$

$$= e^{-1} - e^{-2}$$

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

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

aside: cts case:

$$P(X \in A) = \int_A \underbrace{f(t)}_{\text{should be } \geq 0} dt \geq 0$$

so integral is

$$1 = P(S) = P(X \in \mathbb{R}) = \int_{\mathbb{R}} f(t) dt$$

Fact: If $g(x) \geq 0$ and $\int_{\mathbb{R}} g(x) dx = c < \infty$

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

$$① f(x) \geq 0$$

$$② \int f(x) dx = \int \frac{1}{c} g(x) dx = \frac{1}{c} c = 1.$$

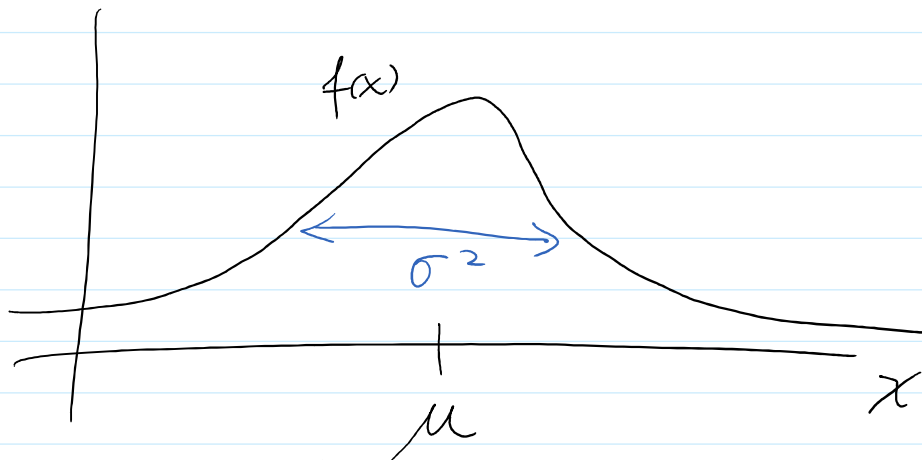
Ex. Normal Distribution (Gaussian)

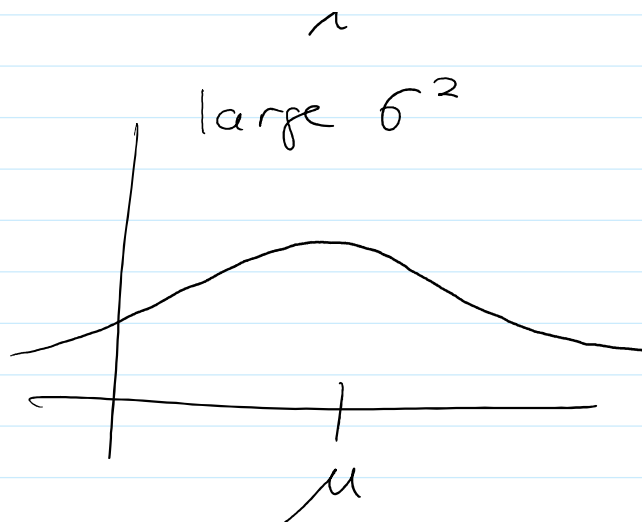
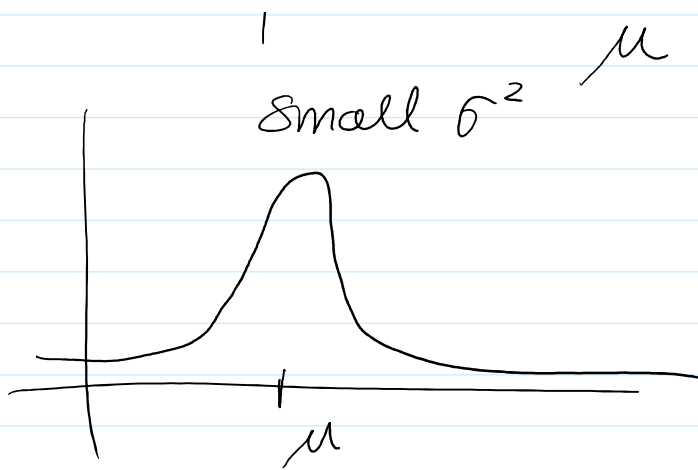
Notation:

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

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

variance: $\sigma^2 > 0$





Special case: standard normal

$$\mu = 0 \text{ and } \sigma^2 = 1$$

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

$$f(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}x^2\right) \quad \forall x \in \mathbb{R}$$

$$\exp(a) = e^a$$

Q: Is this a valid PDF?

$$\textcircled{1} f(x) \geq 0 \quad \checkmark$$

$$\textcircled{2} \int_{\mathbb{R}} f(x) dx = 1$$

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

I

$$\text{Want: } I = 1. \Leftrightarrow I^2 = 1.$$

want: $I = 1 \Rightarrow I = 1$.

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

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

$$= \iint_{\mathbb{R}^2} \frac{1}{2\pi} \exp\left(-\frac{1}{2}(x^2+y^2)\right) dx dy$$

Polar coordinates.