PMF: f(x) = P(X = x) $CDF: F(x) = P(X \le x)$ $F(x) = \sum_{i \le x} f(i)$

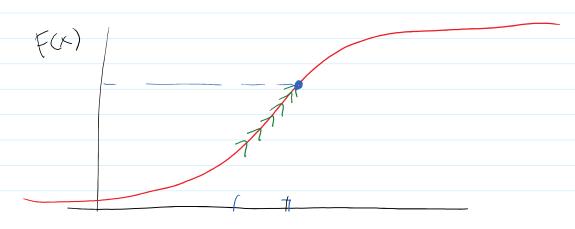
F(x) funp size = f(x) lim F(X-E) 670 x-E x

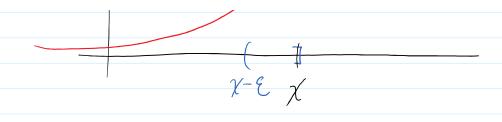
 $\mathbb{P}(a \angle X \leq b) = F(b) - F(a)$

 $\lim_{\epsilon \downarrow 0} P(\chi - \xi < \chi \leq \chi) = \lim_{\epsilon \downarrow 0} F(\chi) - F(\chi - \epsilon)$

 $P(X=X) = F(X) - \lim_{\epsilon \to 0} F(X-\epsilon)$ f(x) = jump size

What about a cts RV?





$$P(X=\chi) = \lim_{\epsilon \to 0} P(\chi - \epsilon \angle X \leq \chi)$$

$$= --- = F(\chi) - \lim_{\epsilon \to 0} F(\chi - \epsilon)$$

$$= F(\chi)$$

Wart: ets analog for PMF:

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

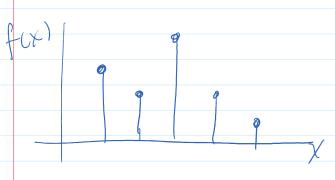
Defu: Probability Density Function (PDF) Cts version of PMF

The PDF for a Cts RV is a function $f: R \rightarrow R$, defined for $x \in R$, as the function flut satisfies

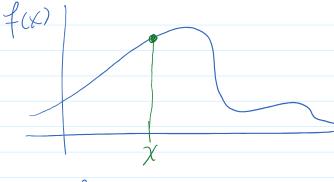
$$F(x) = \int_{-\infty}^{\infty} f(t) dt$$

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

So
$$f(x) = \frac{dF}{dx}$$
. (PDF = deriv. of CDF)



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



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

Properties of PDFs

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

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

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

$$P(x=a) = P(x=b) = 0$$

$$P(a < x < b) = P(a < x < b)$$

$$= P(a < x < b)$$

$$= P(a < x < b)$$

Generally:

(discrete) $P(X \in A) = Z f(i)$

(C+s)
$$P(X \in A) = \int_A f(x) dx$$

 $\frac{e_{X}}{F(x)} = \frac{1}{1 + e^{-x}}$ Q! What is its PDF?

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

Ex. Continuous Unif. Dist.

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

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

unat's the CDF? $F(x) = \int_{-\infty}^{x} f(t) dt$
 $f(x) = \int_{-\infty}^{x} f(t) dt = \int_{-\infty}^{x} f(t) dt = 0$
 $f(x) = \int_{-\infty}^{x} f(t) dt = \int_{-\infty}^{x} 1 dt = x$

$$F(x) = \int f(t)dt = \int 1 dt = 1$$

$$F(x) = \int f(x)dx = \int 1 dx = 1$$

$$0 \qquad x < 0$$

$$x > 1$$

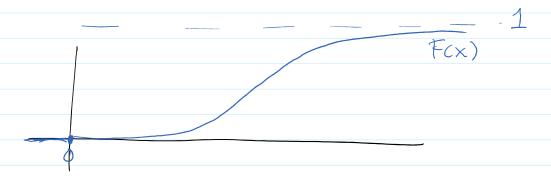
$$F(x) = \begin{cases} 0 \qquad x < 0 \end{cases}$$

$$x > 1$$

$$f(x) = \begin{cases} x > 1 \end{cases}$$

$$f(x)$$

$$= \frac{\chi^2}{4} = \frac{4}{4} - \frac{1}{4} = \frac{3}{4}$$



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

Wayl:
$$P(1 < x < 2) = F(2) - F(1)$$

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

$$\frac{\text{Wag 2:}}{\text{f(x)}} = \frac{dF}{dx} = \frac{1}{dx} \left(1 - e^{-x}\right) = e^{-x}$$

$$\text{for } x \neq 0$$

$$P(1 < x < 2) = \int_{1}^{2} f(x) dx = \int_{1}^{2} e^{-x} dx = -e^{-x} \Big|_{1}^{2}$$

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

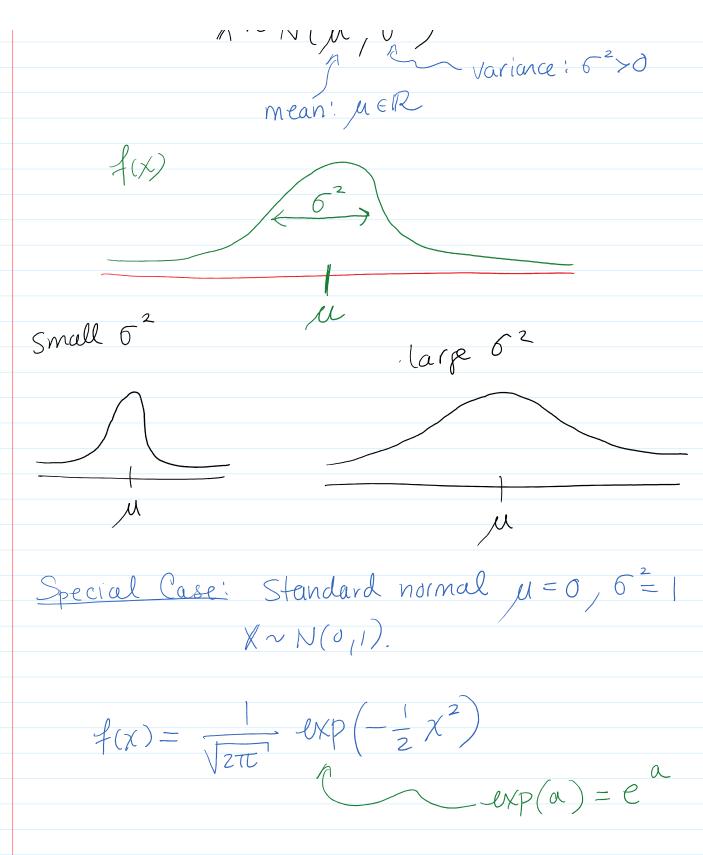
(2) (discrete)
$$\sum_{X \in \mathbb{R}} f(X) = 1$$

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

If
$$g(x) \ge 0$$
 and $\int_R g(x) dx = c < \infty$
 $f(x) = \frac{1}{c}g(x)$
then f is a PDF,

Ex. Normal Distribution (Gaussian)

Nofation:
$$\chi \sim N(\mu, 6^2)$$
 variance: $6^2 > 0$



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

$$\int_{\mathbb{R}} \int_{\mathbb{R}} \exp(-\frac{1}{2}x^{2}) dx = 1$$

$$I$$

$$I$$

$$I = 1 \Rightarrow I^{2} = 1$$

$$I^{2} = I \cdot I = \int_{\mathbb{R}} \exp(-\frac{1}{2}x^{2}) dx$$

Want:
$$T = 1 \Leftrightarrow T^2 = 1$$
.

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

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

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

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

$$\begin{array}{c}
\chi = r \cos \theta \\
y = r \sin \theta
\end{array}$$

$$\frac{1}{2} \int_{0}^{\infty} (x,y) dx dy$$

$$\frac{1$$

If X is a RV then the mean or

expected valve of X denoted E[X].

is defined as

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

$$= \sum_{\chi \in Support(\chi)} \chi \in Support(\chi)$$

) Continuous:
$$E[X] = \int x f(x) dx$$

PDF mean

