

Lecture 8

Ex. cont. from last time

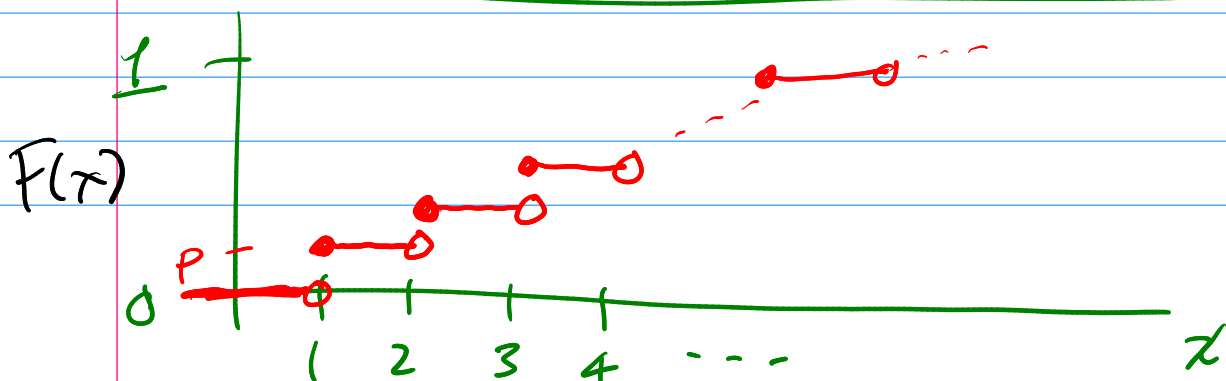
$$\begin{aligned} F(x) &= P(X \leq x) \\ &= P(X=1) + P(X=2) + P(X=3) + \dots + P(X=x) \\ &= \sum_{i=1}^x P(X=i) \\ &= \sum_{i=1}^x (1-p)^{i-1} p \\ &= p \sum_{i=0}^{x-1} (1-p)^i \end{aligned}$$

Geometric Sum:
 $\sum_{i=0}^{n-1} r^i = \frac{1-r^n}{1-r}$

$r = 1-p$

$$= p \frac{1 - (1-p)^x}{1 - (1-p)}$$

$$F(x) = 1 - (1-p)^x \quad \text{for } x=1, 2, 3, 4, \dots$$



$$F(x) = \begin{cases} 0, & x < 1 \\ 1 - (1-p)^{\lfloor x \rfloor}, & x \geq 1 \end{cases}$$

$\lfloor x \rfloor = \text{floor}(x)$

Defn: discrete / cts

A discrete RV is one whose CDF is a step function.

A continuous RV is one whose CDF is continuous.

Defn: Probability Mass Function (PMF)

For a discrete RV X the PMF is a function

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

So that $\forall x \in \mathbb{R}$

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

Theorem: For discrete RVs

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

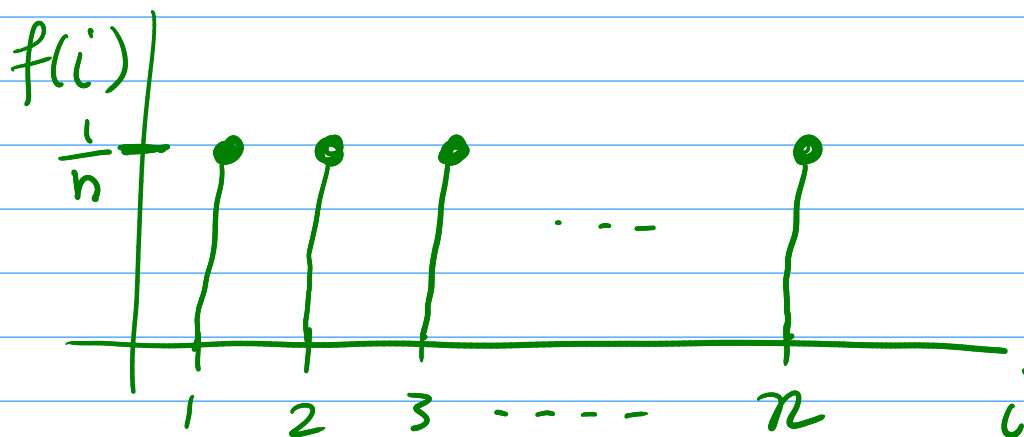
Ex. We say a RV X has a discrete Uniform dist over $1, \dots, n$

notation:

$$X \sim U(\{1, \dots, n\})$$

↑
read: "distributed as"

if the PMF has the form



algebraically:

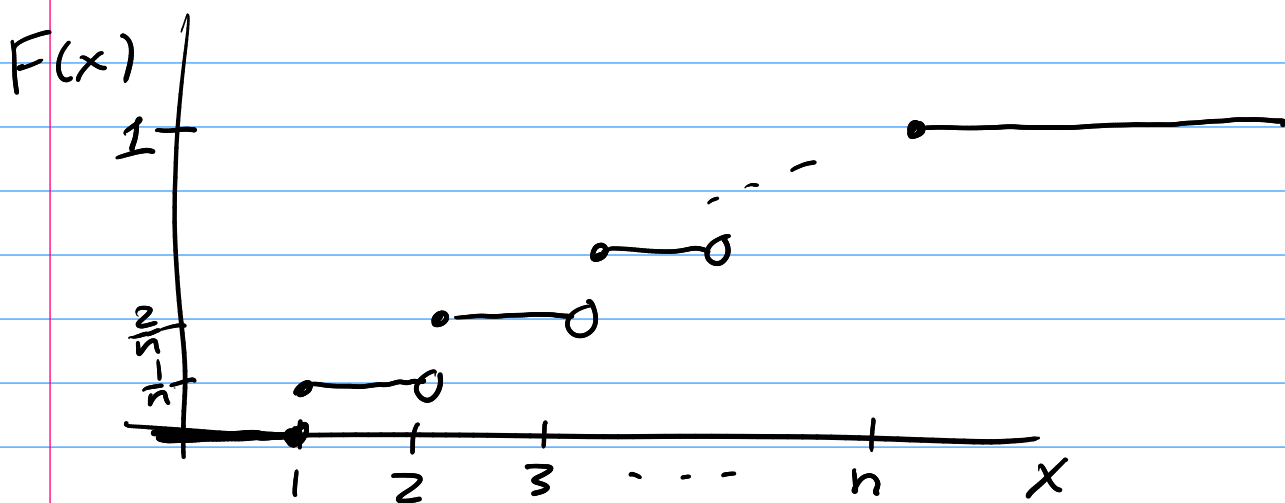
$$f(i) = \begin{cases} \frac{1}{n}, & i=1, 2, \dots, n \\ 0, & \text{else.} \end{cases}$$

Q: what's the CDF?

$$F(x) = \sum_{i \leq x} f(i) = \sum_{i=1}^x \frac{1}{n} = \frac{x}{n}$$

↖ for $x=1, 2, \dots, n$

$$F(x) = \begin{cases} 0, & x < 1 \\ \frac{\lfloor x \rfloor}{n}, & 1 \leq x \leq n \\ 1, & x > n \end{cases}$$



More generally, (for discrete)

$$P(X \in A) = \sum_{i \in A} f(i).$$

Ex. (continue) $X \sim U(\{1, \dots, 7\})$

$$P(2 \leq X \leq 5)$$

$$= P(X \in \{2, 3, 4, 5\})$$

$$= \sum_{x=2}^5 f(x)$$

$$= \sum_{x=2}^5 1/7 = 4/7.$$

Ex. Roll a die 60 times (indepdy)

$X = \# \text{ of } 5 \text{ among rolls}$

What's the PMF of X ?

$$\begin{aligned} f(0) &= P(X=0) = \underbrace{\left(\frac{5}{6}\right)\left(\frac{5}{6}\right)\left(\frac{5}{6}\right) \cdots \left(\frac{5}{6}\right)}_{60 \text{ times}} \\ &= \left(\frac{5}{6}\right)^{60}. \end{aligned}$$

$$f(1) = P(X=1) = \binom{60}{1} \left(\frac{1}{6}\right) \underbrace{\left(\frac{5}{6}\right) \cdots \left(\frac{5}{6}\right)}_{59 \text{ rolls}}$$

$$= \binom{60}{1} \left(\frac{1}{6}\right) \left(\frac{5}{6}\right)^{59}$$

$$\underbrace{\hspace{10em}}_6$$

$$\left(\frac{5}{6}\right) \left(\frac{5}{6}\right) \left(\frac{1}{6}\right) \cdots$$

$$f(2) = P(X=2) = \binom{60}{2} \left(\frac{1}{6}\right) \left(\frac{1}{6}\right) \underbrace{\left(\frac{5}{6}\right) \cdots \left(\frac{5}{6}\right)}_{58}$$

$$\cdots$$

$$= \binom{60}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^{58}$$

General pattern:

$$f(x) = \binom{60}{x} \left(\frac{1}{6}\right)^x \left(\frac{5}{6}\right)^{60-x}$$

Called a Binomial RV:

I do a series of n independent 0 or 1 experiments — each w/ a binary outcome — and the prob. of a 1 is $p \in [0, 1]$

let $X = \#$ of 1s among n experiments.

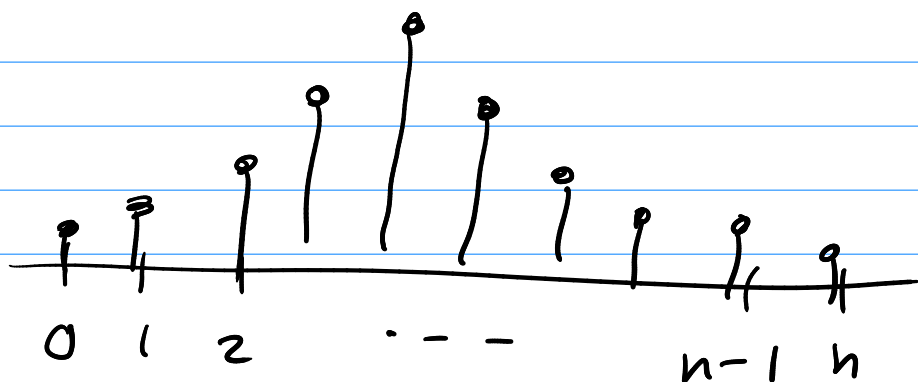
Then X has a binomial dist:

notation: $X \sim \text{Bin}(n, p)$.

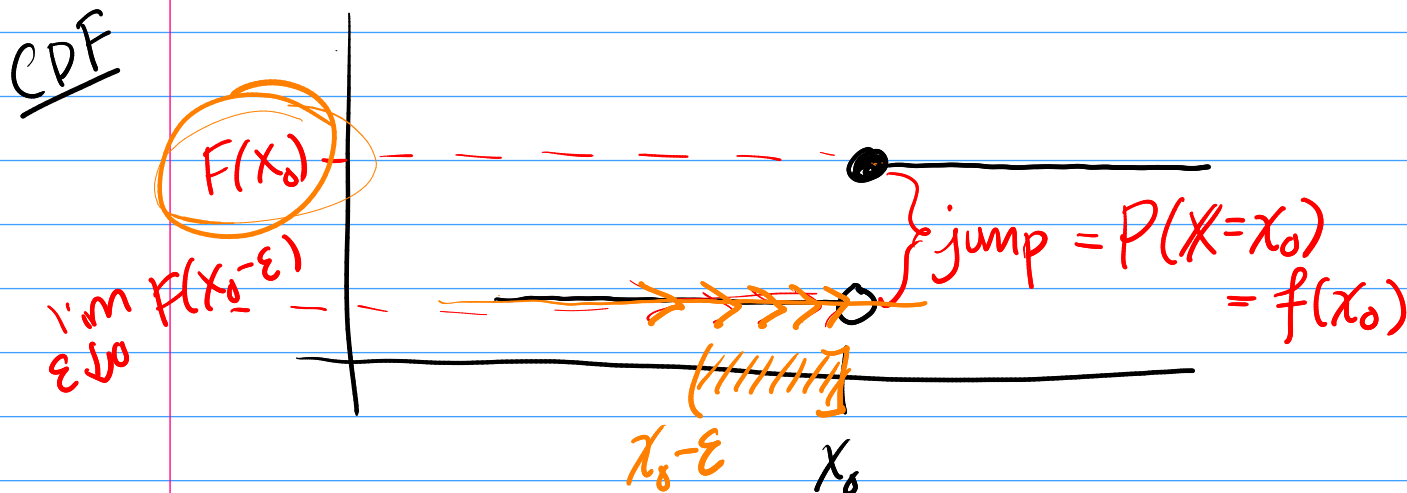
PMF:

$$f(x) = \binom{n}{x} p^x (1-p)^{n-x}$$

for $x = 0, 1, 2, \dots, n$.



Discrete:

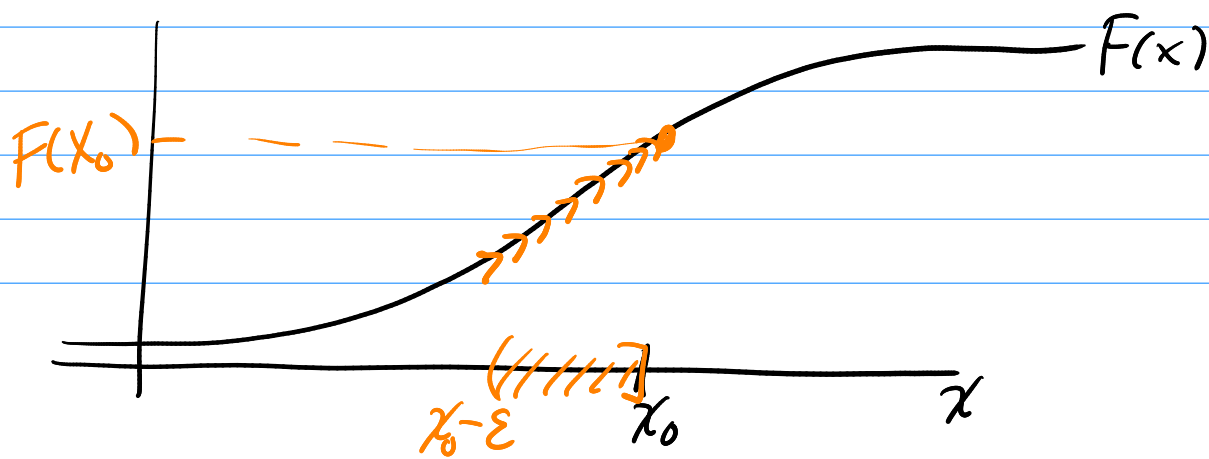


rule: $P(a < X \leq b) = F(b) - F(a)$

$\lim_{\epsilon \downarrow 0} P(x_0 - \epsilon < X \leq x_0) = \lim_{\epsilon \downarrow 0} F(x_0) - F(x_0 - \epsilon)$

$P(X=x_0) = F(x_0) - \lim_{\epsilon \downarrow 0} F(x_0 - \epsilon)$
 $\rightarrow = \text{jump size}$

what about a cts RV?



$$\begin{aligned}
 P(X=x_0) &= \dots = F(x_0) - \lim_{\varepsilon \downarrow 0} F(x_0 - \varepsilon) \\
 &= F(x_0) - F(x_0) \\
 &= 0
 \end{aligned}$$

Want is analogy for PMF:

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

Defn: Probability Density Function (PDF).

(cts ver of a PMF)

The PDF of a cts RV is a function

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

defined so that for $x \in \mathbb{R}$

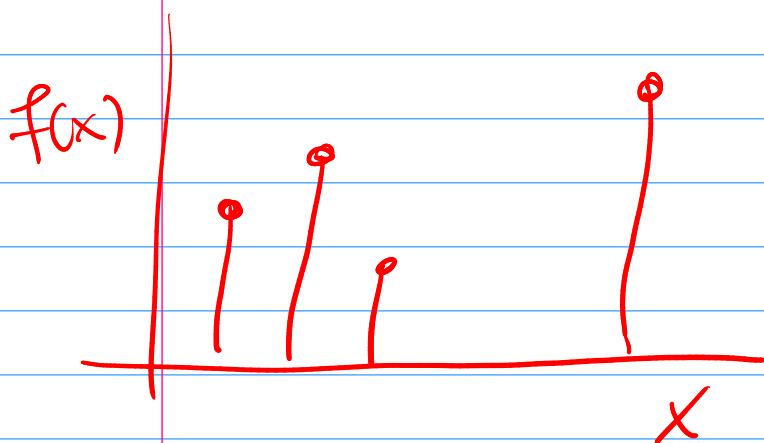
$$F(x) = \int_{-\infty}^x f(t) dt.$$

By Fund. Thm of Calc:

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

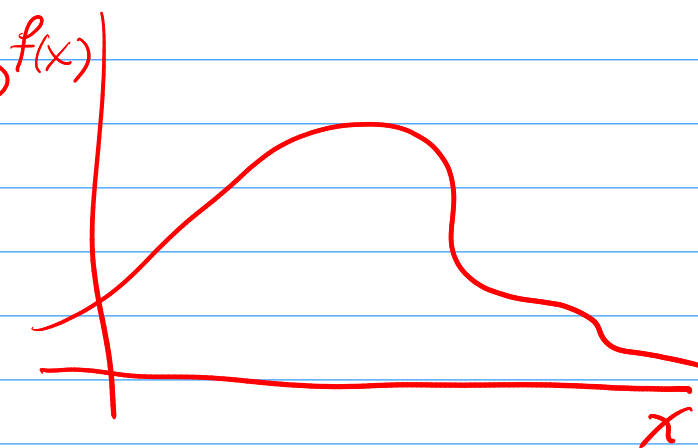
Key: PDF = deriv. of CDF.

discrete PMF



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

continuous PDF

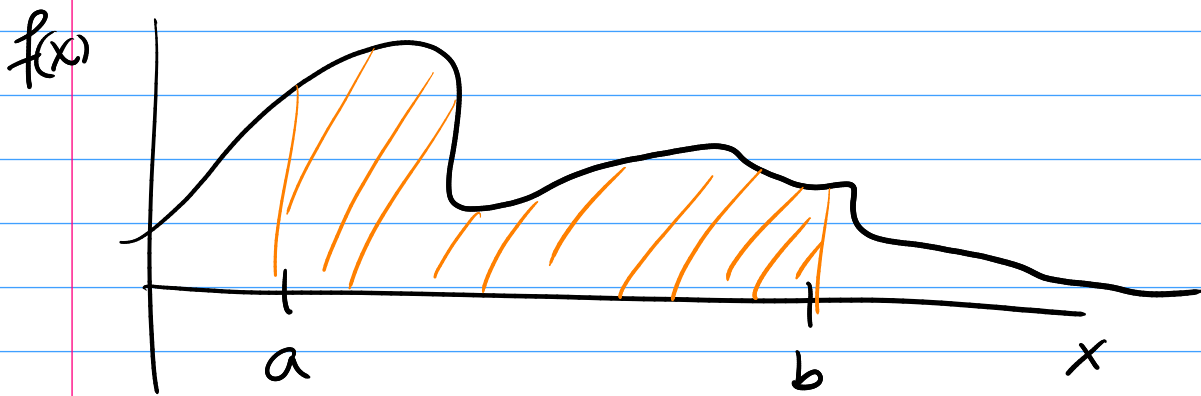


$$f(x) \neq P(X=x).$$

Properties of PDFs

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

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



Note: $P(X=a) = 0 = P(X=b)$

so,
$$\begin{aligned} P(a < X \leq b) &= P(a \leq X \leq b) \\ &= P(a \leq X < b) \\ &= P(a < X < b). \end{aligned}$$
 cts

Generally:

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

(cts)
$$P(X \in A) = \int_A f(t) dt$$