

## Oct 11<sup>th</sup> Continuous Random Variables (Ch 5 in Ross)

Uncountable infinite values for the RV

↪  $\mathbb{R}$

↪ interval of  $\mathbb{R}$  ex:  $[a,b]$ ,  $[0,1]$

examples "exact" weight  
time  
temp  
share price

} if we can divide the  
units w/o stopping

A random variable  $X$  is said to have a continuous distribution if there exists a non-negative  $f = f_x$  such that

$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

area under  
the curve of  $f(x)$

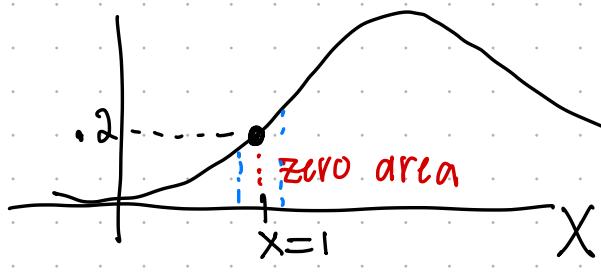
for every  $a$  and  $b$ .

$f$ : prob density function (pdf)

\* We consider intervals for continuous R.V.

\* prob at one specific point = 0

$$P(1 \leq X \leq 1) = \int_1^1 f(x) dx = 0$$



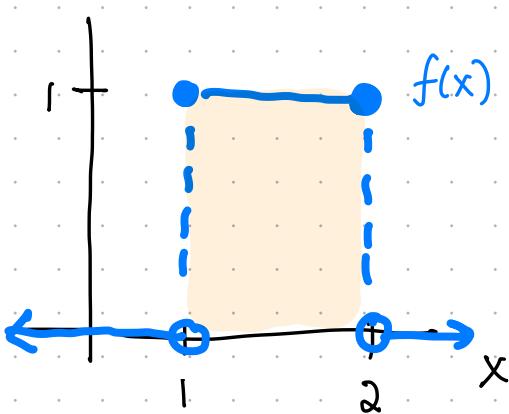
$$P(.99999 \leq X \leq 1.00001)$$

$$= \int_{.99999}^{1.00001} f(x) dx > 0$$

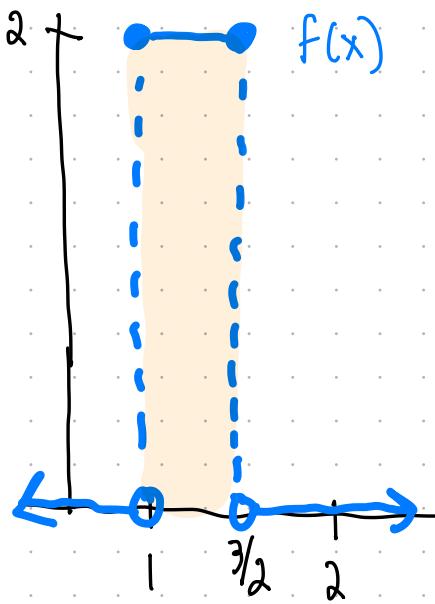
Note:  $f(x)$  is not the prob

↪ the prob is the integral of  $f(x)$  of a certain interval

Imagine two examples



$$\begin{aligned} P(1 \leq X \leq 2) &= \text{area of rectangle} \\ &= \text{base} \cdot \text{height} \\ &= (2-1) \cdot 1 = 1 \cdot 1 = 1 \end{aligned}$$



$$\begin{aligned} P\left(1 \leq X \leq \frac{3}{2}\right) &= \text{area of rectangle} \\ &= \text{base} \cdot \text{height} \\ &= \left(\frac{3}{2}-1\right) \cdot 2 = \frac{1}{2} \cdot 2 = 1 \end{aligned}$$

\*  $f(x) > 1$  can occur for some values of  $x \in \mathbb{R}$

→ If we integrate over the entire domain → should get 1

$$P(-\infty < X < \infty) = \int_{-\infty}^{\infty} f(x) dx = 1$$

ex Let  $X$  be a cont R.V. with PDF:

C Constant

$$f(x) = \begin{cases} c(4x - 2x^2) & 0 < x < 2 \\ 0 & \text{Otherwise} \end{cases}$$

a) What is  $c$ ?

b)  $P(X > 1)$ ?

$$\begin{aligned} a) 1 &= \int_{-\infty}^{\infty} f(x) dx = \int_0^2 c(4x - 2x^2) dx = c \int_0^2 (4x - 2x^2) dx \\ &= c \left( \frac{4x^2}{2} - \frac{2x^3}{3} \right) \Big|_{x=0}^{x=2} \\ &= c \left( \frac{4(2)^2}{2} - \frac{2(2)^3}{3} - \left( \frac{4(0)^2}{2} - \frac{2(0)^3}{3} \right) \right) \end{aligned}$$

$$1 = c \left( 8 - \frac{16}{3} \right) \approx c \left( \frac{24 - 16}{3} \right) = c \frac{8}{3}$$

$$C = \frac{3}{8}$$

$$b) P(X > 1) = P(X \geq 1) = \int_1^{\infty} f(x) dx$$

$$\begin{aligned} &= \int_1^2 \frac{3}{8} (4x - 2x^2) dx \\ &= \frac{3}{8} \left( 2x^2 - \frac{2}{3}x^3 \right) \Big|_1^2 \end{aligned}$$

$$= \frac{3}{8} \left( 8 - \frac{2}{3} \cdot 8 - \left( 2 - \frac{2}{3} \right) \right)$$

$$= \frac{3}{8} \left( 6 - \frac{16}{3} + \frac{2}{3} \right)$$

$$= \frac{3}{8} \left( 6 - \frac{14}{3} \right) = \frac{3}{8} \left( \frac{18}{3} - \frac{14}{3} \right)$$

$$= \frac{3}{8} \left( \frac{4}{3} \right) = \frac{1}{2}$$

$\lambda = 2$

Assume you receive an average of 2 phone calls per hour. You make the Poisson assumption so the time between calls follows an exponential density,

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \in [0, \infty) \\ 0 & \text{otherwise} \end{cases}$$

Assume you just hung up the phone

a) what is the prob that you are going to have to wait more than an hour for the next call?

b) You'll receive a call w/in 10 min?

c) Wait more than 10 min but less than an hour?

$$a) P(X > 1) = P(X \geq 1)$$

$$= \int_1^\infty f(x) dx$$

$\lambda$ : constant

$$= \int_1^\infty \lambda e^{-\lambda x} dx$$

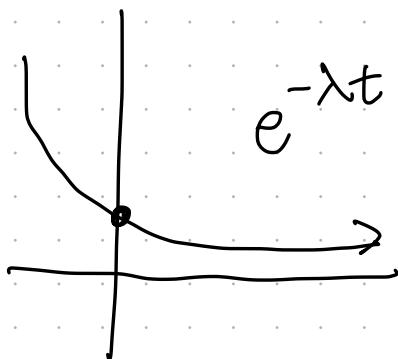
$$= \left[ \frac{e^{-\lambda x}}{-\lambda} \right]_1^\infty$$

$$= \lim_{t \rightarrow \infty} \left[ -e^{-\lambda x} \right]_1^t$$

$$= \lim_{t \rightarrow \infty} \left( e^{-\lambda t} - e^{-\lambda \cdot 1} \right)$$

$$= 0 + e^{-\lambda} = e^{-\lambda} = e^{-2}$$

$x=2$   
 $e^{-2} \approx 0.135$

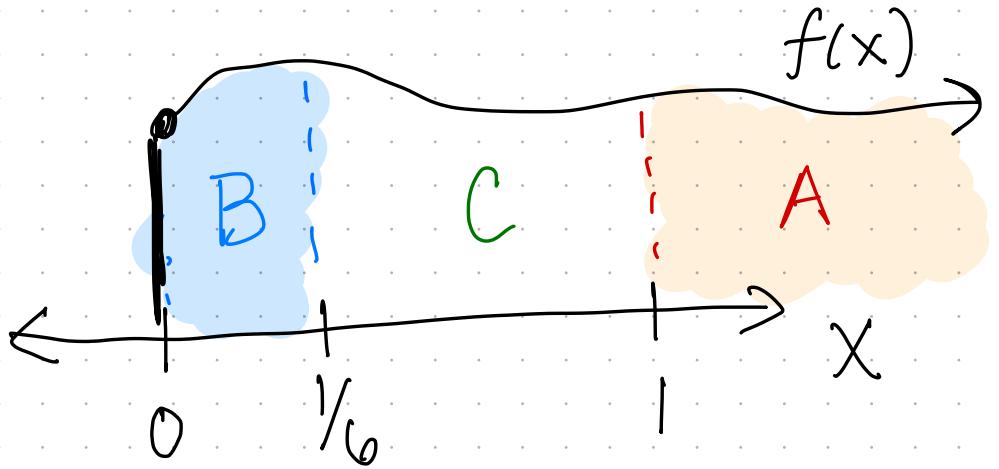


$$b) P\left(X < \frac{10}{60}\right) = P\left(X \leq \frac{1}{6}\right)$$

$$= \int_0^{\frac{1}{6}} 2e^{-2x} dx$$

$$\begin{aligned}
 &= \left[ -\frac{e^{-2x}}{2} \right]_0^{1/6} \\
 &= -e^{-1/3} - e^0 = -e^{-1/3} \\
 &= 1 - e^{-1/3} = 0.2834
 \end{aligned}$$

c) More than 10 min but less than 1hr



$$P\left(\frac{1}{6} \leq X \leq 1\right) = 1 - .135 - .2834$$

$$= .5816$$

OR

$$\int_{1/6}^1 \lambda e^{-\lambda x} dx$$

The expected value of a cont R.V.  $X$  is

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

More generally:  $E[g(x)] = \int_{-\infty}^{\infty} g(x) f(x) dx$

Linearity:  $E[aX+b] = a E[X] + b$

$\downarrow \quad \downarrow$   
constants

$\hookrightarrow E[b] = b$

$$E[X+Y] = E[X] + E[Y]$$

$$\begin{aligned} \text{Var}(x) &= E[(x-\mu)^2] & \mu &= E[x] \\ &= E[X^2] - (E[X])^2 \end{aligned}$$

ex  $X$  R.V. with density:

$$f_X(x) = \begin{cases} 2x & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

find  $E[X]$  &  $\text{Var}(x)$

$$E[X] = \int_{-\infty}^{\infty} xf(x) dx = \int_0^1 2x^2 dx = \left. \frac{2x^3}{3} \right|_0^1 = \frac{2}{3}(1^3 - 0^3) = \frac{2}{3}$$

$$E[X^2] = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_0^1 2x^3 dx = \left. \frac{2x^4}{4} \right|_0^1 = \frac{2}{4}(1 - 0) = \frac{1}{2}$$

$$\text{Var}(x) = E[X^2] - (E[X])^2$$

$$= \frac{1}{2} - \left( \frac{2}{3} \right)^2 = \frac{1}{2} - \frac{4}{9}$$

$$= \frac{9}{18} - \frac{8}{18} = \frac{1}{18}$$

ex The density of  $X$  is given by:

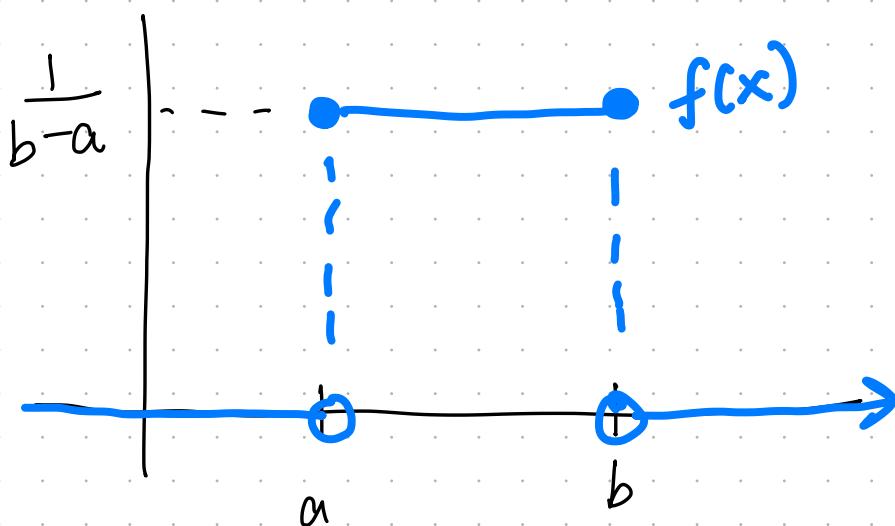
$$f(x) = \begin{cases} \frac{1}{2} & 0 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} \text{find } E[e^x] &= \int_0^2 e^x \cdot \frac{1}{2} dx = \frac{1}{2} \int_0^2 e^x dx \\ &= \left. \frac{1}{2} e^x \right|_0^2 = \frac{1}{2} (e^2 - e^0) = \underline{\frac{1}{2}(e^2 - 1)} \end{aligned}$$

We say that a R.V.  $X$  has a uniform distribution on  $[a, b]$  if

$$f(x) = \begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

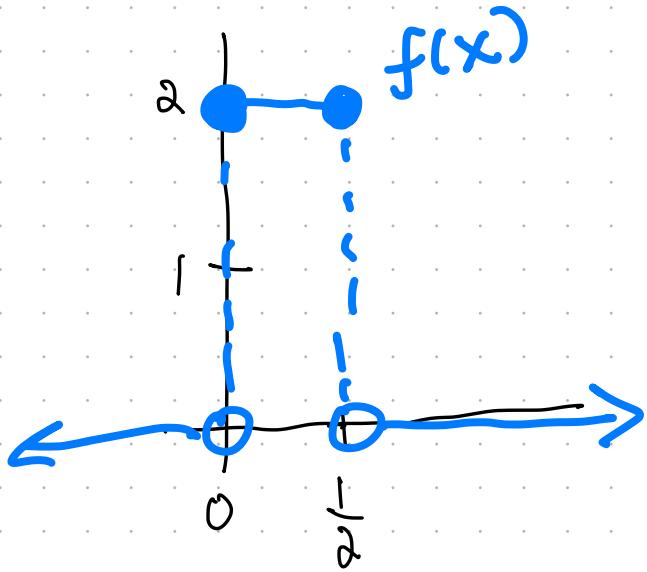
constant



ex Uniform  $(0, \frac{1}{2})$

$$\begin{array}{c} \downarrow \\ a \\ \downarrow \\ b \end{array}$$

$$\frac{1}{b-a} = \frac{1}{\frac{1}{2}-0} = 2$$



$$E[X] = \frac{a+b}{2} \text{ (midpoint)}$$

Derivation:

$$\begin{aligned}
 E[X] &= \int_{-\infty}^{\infty} x f(x) dx = \int_a^b x \cdot \frac{1}{b-a} dx = \frac{1}{b-a} \int_a^b x dx \\
 &= \frac{1}{b-a} \frac{x^2}{2} \Big|_a^b \\
 &= \frac{1}{b-a} \left( \frac{b^2}{2} - \frac{a^2}{2} \right) \quad \text{diff of 2 } \square \\
 &= \frac{1}{2(b-a)} (b^2 - a^2) = \frac{1}{2(b-a)} (b-a)(b+a) \\
 &= \frac{a+b}{2}
 \end{aligned}$$

$$\begin{aligned}
 E[X^2] &= \int_a^b x^2 \frac{1}{b-a} dx = \frac{1}{b-a} \left. \frac{x^3}{3} \right|_a^b \\
 &= \frac{1}{3(b-a)} (b^3 - a^3) \quad \text{diff of cubes} \\
 &= \frac{(b-a)(b^2 + ab + a^2)}{3(b-a)} \\
 &= \frac{b^2 + ab + a^2}{3}
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(x) &= E[X^2] - (E[X])^2 \\
 &= \frac{b^2 + ab + a^2}{3} - \left( \frac{a+b}{2} \right)^2 = \frac{(a-b)^2}{12} \text{ OR } \frac{(b-a)^2}{12}
 \end{aligned}$$

clean up algebra

## Uniform Distribution $(a, b)$ interval

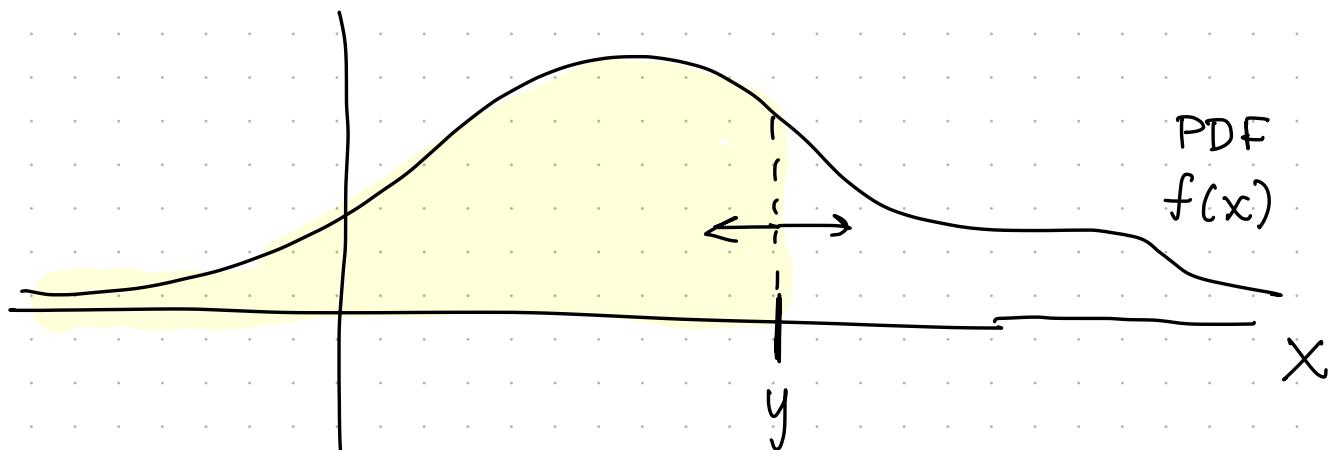
$$E[X] = \frac{a+b}{2} \quad (\text{midpt})$$

$$\text{Var}(X) = \frac{(b-a)^2}{12} \quad \frac{(\text{length of interval})^2}{12}$$

The Cumulative Distribution Function (CDF) of  $X$   
is defined as

$$F(y) = F_x(y) = P(-\infty < x \leq y) = \int_{-\infty}^y f(x) dx$$

pdf



integrate  
differentiate

PDF  $f(x)$

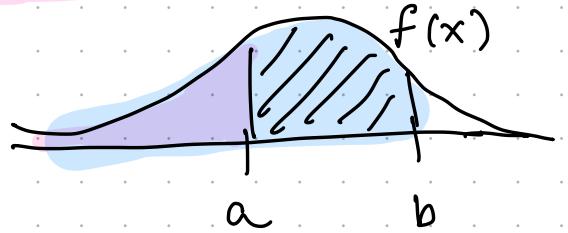
CDF  $F_x(y)$

Thm  $\frac{d}{dy} F_x(y) = f(x)$

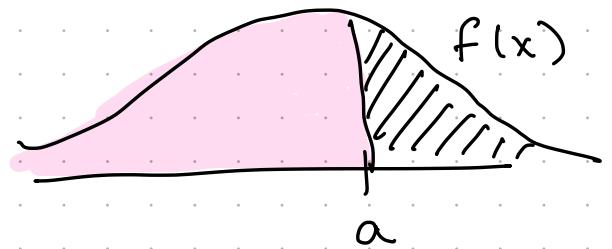
Prop  $\lim_{y \rightarrow \infty} F_x(y) = \lim_{y \rightarrow \infty} \int_{-\infty}^y f(x) dx = 1$

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

$$\circ P(a \leq X \leq b) = F_x(b) - F_x(a)$$

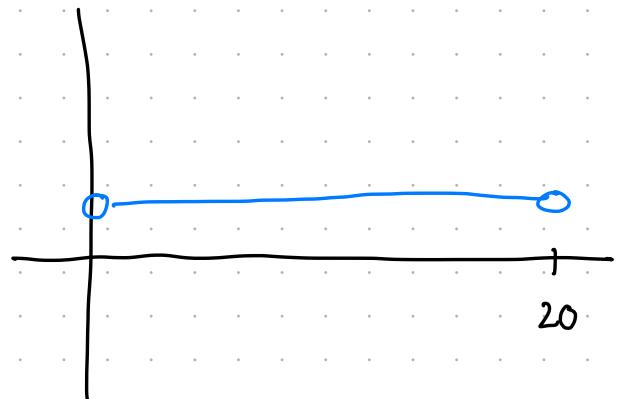


$$\circ P(X > a) = 1 - F_x(a)$$



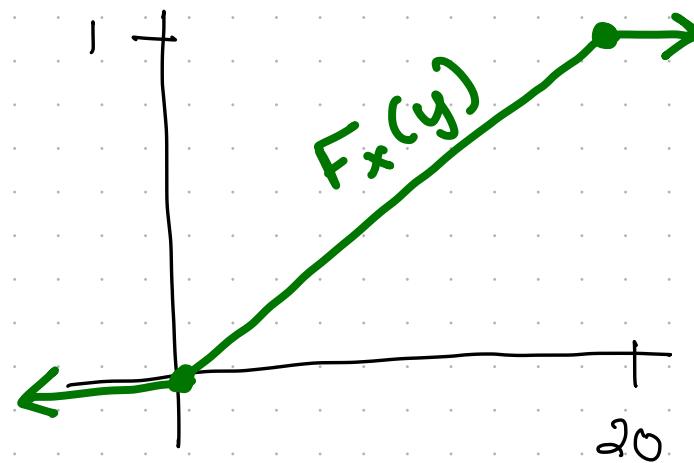
ex) Uniform(0,20)

$$f_x(y) = \begin{cases} \frac{1}{20} & \text{if } 0 < y < 20 \\ 0 & \text{Otherwise} \end{cases}$$



$$\text{(CDF: } P(X \leq y) = \int_{-\infty}^y f_x(x) dx = \int_0^y \frac{1}{20} dx = \left[ \frac{1}{20} x \right]_0^y = \frac{1}{20} y \text{)}$$

$$F_x(y) = \begin{cases} 0 & y \leq 0 \\ \frac{y}{20} & 0 < y < 20 \\ 1 & y \geq 20 \end{cases}$$



Given: PDF of  $X$  (distribution of  $X$ )

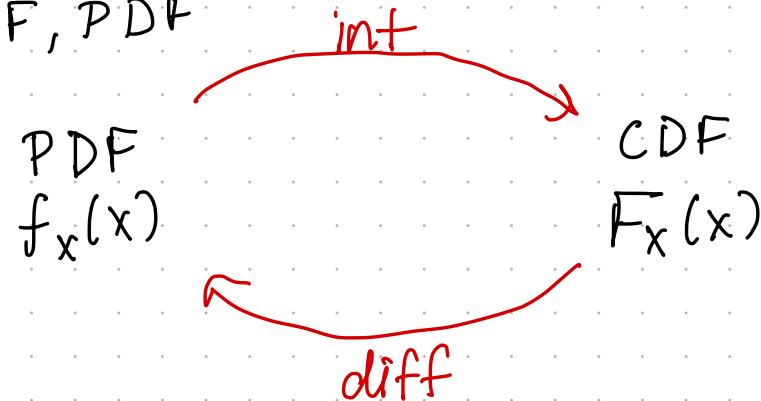
Want: PDF of function of  $X$  ( $g(X)$ )

ex)  $X = \text{Uniform}(0, 1) = \begin{cases} 1 & 0 \leq X \leq 1 \\ 0 & \text{otherwise} \end{cases}$

Want the distribution for  $Y = X^n$ .

$\hookrightarrow X^2, X^{10}$

trick: CDF, PDF



For  $0 \leq y \leq 1$ :

$$F_Y(y) = P(Y \leq y) = P(X^n \leq y) = P(X \leq y^{1/n}) = F_X(y^{1/n})$$

CDF for  $Y$   $\downarrow$  def'n of CDF

def'n of CDF

$$F_Y(y) = F_X(y^{1/n})$$

CDF for  $x$

plug into CDF

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

above (last page)

CDF  $F_Y(y) = y^{1/n}$

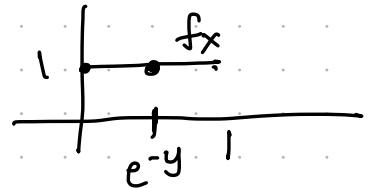
PDF  $f_y(y) = \frac{1}{n} y^{1/n-1}$  ↓ take der

$$\Rightarrow f_y(y) = \begin{cases} \frac{1}{n} y^{1/n-1} & 0 \leq y \leq 1 \\ 0 & \text{Otherwise} \end{cases}$$

ex  $X = \text{Uniform}(0, 1)$   $f_x(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{Otherwise} \end{cases}$

$$Y = -\log(X)$$

Want PDF of  $Y$ :



$$\begin{aligned} F_Y(y) &= P(Y \leq y) = P(-\log(X) \leq y) \\ &= P(\log(X) \geq -y) \\ &= P(X \geq e^{-y}) \\ &= 1 - P(X \leq e^{-y}) \\ &= 1 - F_X(e^{-y}) \end{aligned}$$

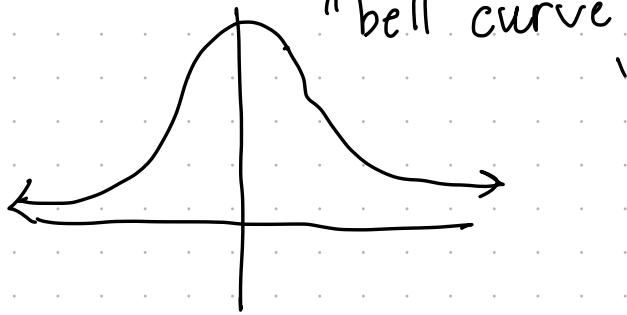
$$F_Y(y) = 1 - F_X(e^{-y})$$

↓ take der, chain rule

$$f_Y(y) = 0 - \underline{f_X(e^{-y})} \cdot (-e^{-y})$$

$$\begin{aligned}
 &= 0 - 1 \cdot (-e^{-y}) \\
 &= e^{-y} \quad \text{distribution of } Y \\
 &\quad \text{PDF}
 \end{aligned}$$

## Normal Random Variable



$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

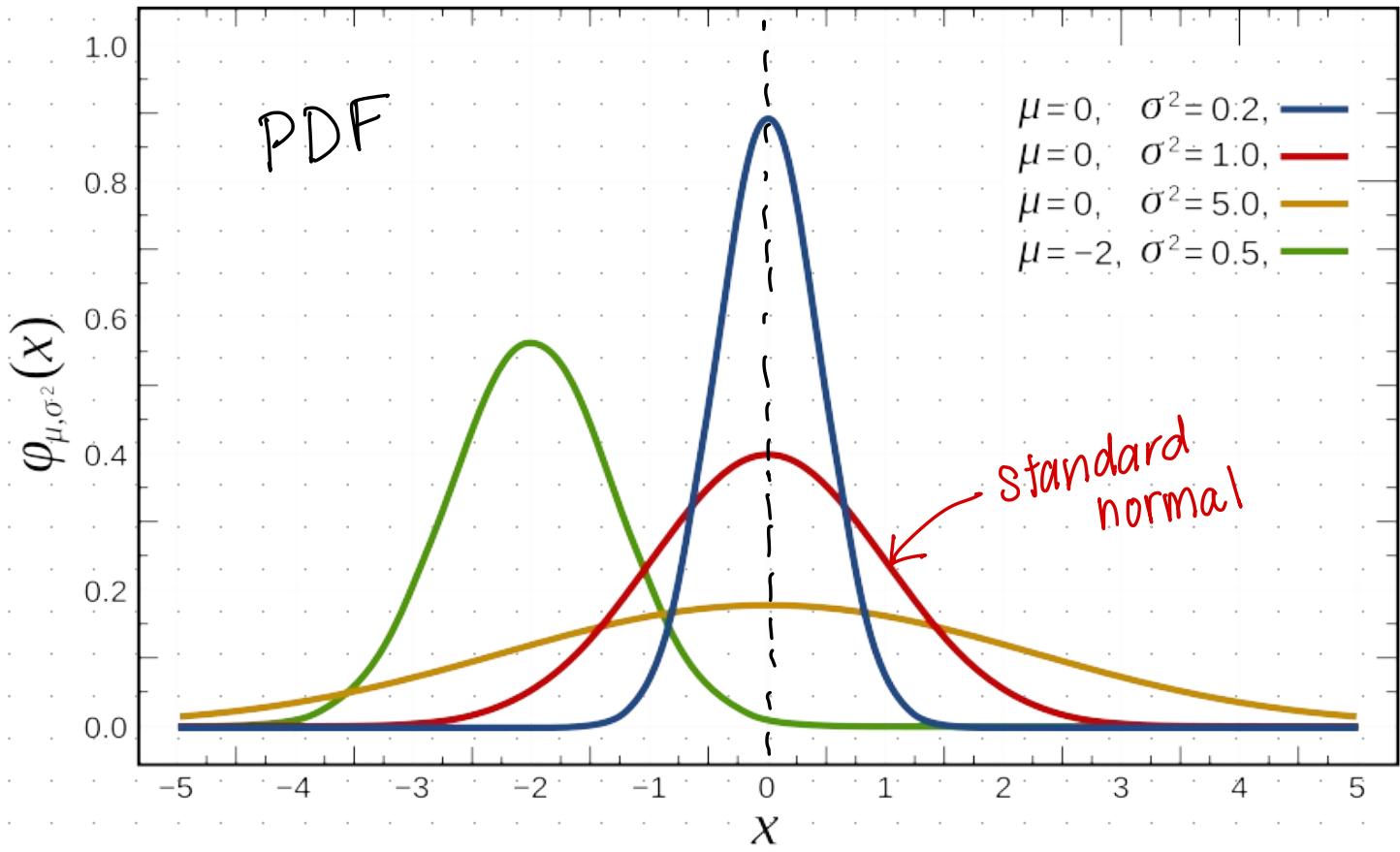
$$\frac{-(x-\mu)^2}{2\sigma^2}$$

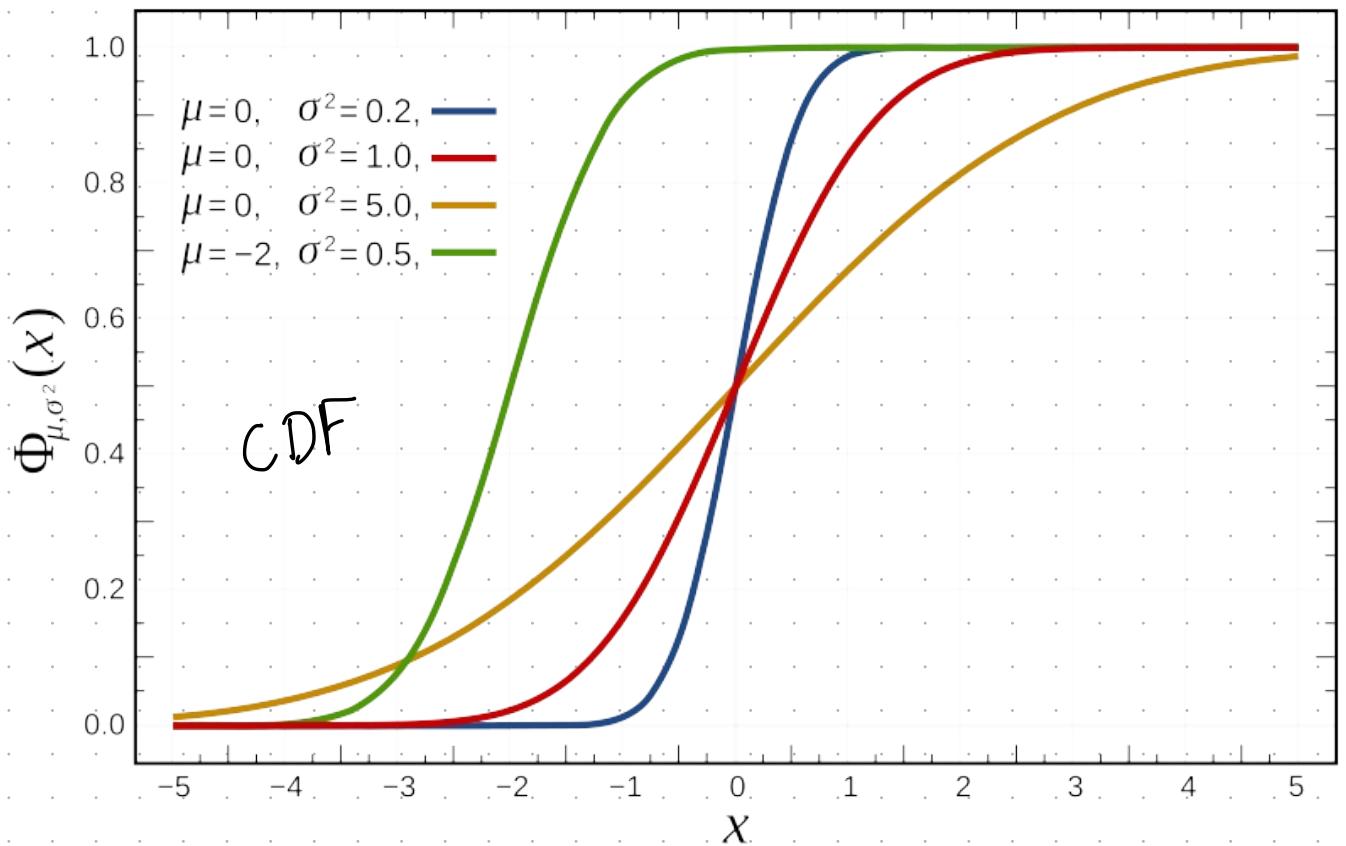
$\mu$  = mean = center

$\sigma^2$  = variance  $\rightarrow$  spread

The "standard" normal with  $\mu=0$  and  $\sigma^2=1$   
 $\sigma=1$

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





Normal  $f(x) \rightarrow$  tricky integral

$$\int_a^b \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

tricky to do

$$\sim \int_a^b e^{-x^2} dx$$

$\mu, \sigma, \pi$   
constants  
involves 2  
integrals w/  
polar coordinates

instead use a CDF chart!  
↑ standard normal

$X$ : any normal distribution "messy"  $\mu$  mean  
 $\sigma^2$  variance

$Z$ : Standard normal  $\mu=0$   
 $\sigma=1$

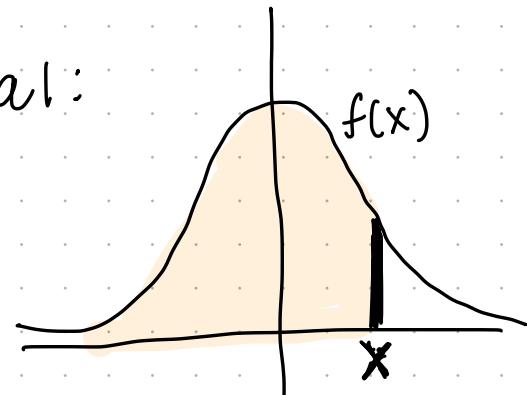
$$Z = \frac{X - \mu}{\sigma}$$

standardize



CDF of a standard normal:

$$\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-y^2/2} dy$$



\* for negative:  $\Phi(-x) = 1 - \Phi(x)$

\* CDF is only for standard normal "messy"

↳ standardize first!  $Z = \frac{X - \mu}{\sigma}$   
 ↑ Standard

## Area $\Phi(x)$ under the standard normal curve to the left of $X$

$X$	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
.0	.5000	.5040	.5080	.5120	.5160	.5199	.5239	.5279	.5319	.5359
.1	.5398	.5438	.5478	.5517	.5557	.5596	.5636	.5675	.5714	.5753
.2	.5793	.5832	.5871	.5910	.5948	.5987	.6026	.6064	.6103	.6141
.3	.6179	.6217	.6255	.6293	.6331	.6368	.6406	.6443	.6480	.6517
.4	.6554	.6591	.6628	.6664	.6700	.6736	.6772	.6808	.6844	.6879
.5	.6915	.6950	.6985	.7019	.7054	.7088	.7123	.7157	.7190	.7224
.6	.7257	.7291	.7324	.7357	.7389	.7422	.7454	.7486	.7517	.7549
.7	.7580	.7611	.7642	.7673	.7704	.7734	.7764	.7794	.7823	.7852
.8	.7881	.7910	.7939	.7967	.7995	.8023	.8051	.8078	.8106	.8133
.9	.8159	.8186	.8212	.8238	.8264	.8289	.8315	.8340	.8365	.8389
1.0	.8413	.8438	.8461	.8485	.8508	.8531	.8554	.8577	.8599	.8621
1.1	.8643	.8665	.8686	.8708	.8729	.8749	.8770	.8790	.8810	.8830
1.2	.8849	.8869	.8888	.8907	.8925	.8944	.8962	.8980	.8997	.9015
1.3	.9032	.9049	.9066	.9082	.9099	.9115	.9131	.9147	.9162	.9177
1.4	.9192	.9207	.9222	.9236	.9251	.9265	.9279	.9292	.9306	.9319
1.5	.9332	.9345	.9357	.9370	.9382	.9394	.9406	.9418	.9429	.9441
1.6	.9452	.9463	.9474	.9484	.9495	.9505	.9515	.9525	.9535	.9545
1.7	.9554	.9564	.9573	.9582	.9591	.9599	.9608	.9616	.9625	.9633
1.8	.9641	.9649	.9656	.9664	.9671	.9678	.9686	.9693	.9699	.9706
1.9	.9713	.9719	.9726	.9732	.9738	.9744	.9750	.9756	.9761	.9767
2.0	.9772	.9778	.9783	.9788	.9793	.9798	.9803	.9808	.9812	.9817
2.1	.9821	.9826	.9830	.9834	.9838	.9842	.9846	.9850	.9854	.9857
2.2	.9861	.9864	.9868	.9871	.9875	.9878	.9881	.9884	.9887	.9890
2.3	.9893	.9896	.9898	.9901	.9904	.9906	.9909	.9911	.9913	.9916
2.4	.9918	.9920	.9922	.9925	.9927	.9929	.9931	.9932	.9934	.9936
2.5	.9938	.9940	.9941	.9943	.9945	.9946	.9948	.9949	.9951	.9952
2.6	.9953	.9955	.9956	.9957	.9959	.9960	.9961	.9962	.9963	.9964
2.7	.9965	.9966	.9967	.9968	.9969	.9970	.9971	.9972	.9973	.9974
2.8	.9974	.9975	.9976	.9977	.9977	.9978	.9979	.9979	.9980	.9981
2.9	.9981	.9982	.9982	.9983	.9984	.9984	.9985	.9985	.9986	.9986
3.0	.9987	.9987	.9987	.9988	.9988	.9989	.9989	.9989	.9990	.9990
3.1	.9990	.9991	.9991	.9991	.9992	.9992	.9992	.9992	.9993	.9993
3.2	.9993	.9993	.9994	.9994	.9994	.9994	.9994	.9995	.9995	.9995
3.3	.9995	.9995	.9995	.9996	.9996	.9996	.9996	.9996	.9996	.9997
3.4	.9997	.9997	.9997	.9997	.9997	.9997	.9997	.9997	.9997	.9998

Φ(1.27)

Φ(.03)

Φ(3.35)

ex Find  $P(1 \leq X \leq 4)$  if  $X$  is  $\mathcal{N}(2, 25)$

$\downarrow \mu$   $\downarrow \sigma^2$   
 normal

not standard

First: standardize:  $Z = \frac{X - \mu}{\sigma} \rightarrow Z_0 = \frac{X - \mu}{\sigma}$   
 $Z_0 + \mu = X$

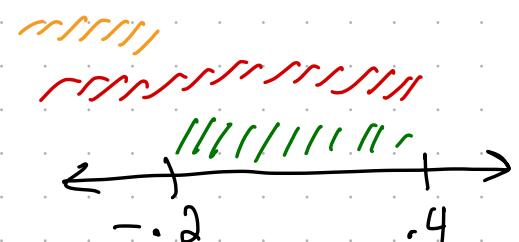
$$P(1 \leq X \leq 4) = P(1 \leq Z_0 + \mu \leq 4)$$

$$= P(1 \leq 5Z + 2 \leq 4)$$

$$= P(-1 \leq 5Z \leq 2)$$

$$= P\left(\frac{-1}{5} \leq Z \leq \frac{2}{5}\right)$$

$$= P(-.2 \leq Z \leq .4)$$



$$= P(Z \leq .4) - P(X \leq -.2)$$

$$= \Phi(.4) - \Phi(-.2)$$

$$= \Phi(.4) - (1 - \Phi(.2))$$

chart

chart

$$= 0.6554 - (1 - 0.5793)$$

$$= 0.2347$$

23.47%

also: Standardize

$$\begin{aligned} P(1 \leq X \leq 4) &= P(1-\mu \leq X-\mu \leq 4-\mu) \\ &= P\left(\frac{1-\mu}{\sigma} \leq \frac{X-\mu}{\sigma} \leq \frac{4-\mu}{\sigma}\right) \\ \mu = 2 & \\ \sigma = 5 & \\ &= P\left(\frac{1-2}{5} \leq Z \leq \frac{4-2}{5}\right) \end{aligned}$$

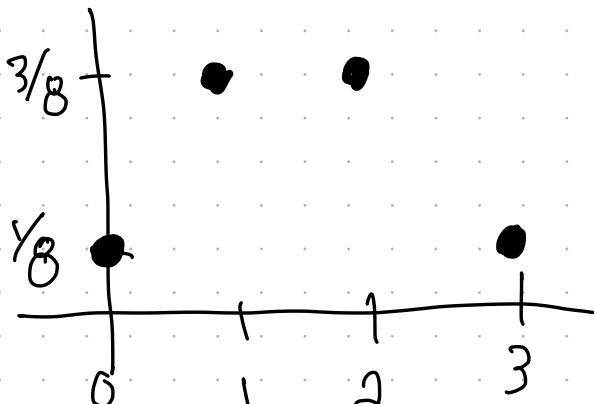
---

## The Normal Approximation to the Binomial Distribution

---

Recall Binomial distribution from discrete chapter

- n independent trials  $\rightarrow$  prob p
  - $X = \#$  of success after n trials
- \* n=3 (coin flips)



$$P(X=0) = 1/8$$

$$P(X=1) = 3/8$$

$$P(X=2) = 3/8$$

$$P(X=3) = 1/8$$

Symmetric?

Now if increase  $n$  (# of trials) the shape starts to look "normal"!

## The DeMoivre-Laplace Limit Thm

If  $S_n = \#$  of successes that can occur when  $n$  independent trials are performed (prob  $p$  each time)  $\rightarrow$  Binomial then for any  $a < b$ ,

$$P\left\{ a \leq \frac{S_n - np}{\sqrt{np(1-p)}} \leq b \right\} \xrightarrow{\mu = E[S_n]} \Phi(b) - \Phi(a)$$

$\underbrace{\sigma^2 = \text{var}(S_n)}$

Standardize

CDF of standard normal chart

\* this approximation is good if  $np(1-p) \geq 10$   
 (gets better as  $np(1-p)$  increases  $\rightarrow n$  increases)

\* this is a special case of the Central Limit Thm (we will see later)

ex Suppose a fair coin is tossed 100 times. What is the prob there will be more than 60 heads?

$$\text{Binomial: } P(X > 60) = \sum_{k=61}^{100} \binom{100}{k} \left(\frac{1}{2}\right)^k \left(\frac{1}{2}\right)^{100-k}$$

40 terms!

$$= 0.028$$

Normal? Can we approximate?

$$\text{check: } np(1-p) \geq 10?$$

$$100 \cdot \frac{1}{2} \left(1 - \frac{1}{2}\right) = 50 \cdot \frac{1}{2} = 25 \checkmark$$

$$P(S_n > 60) = P\left(\frac{S_n - np}{\sqrt{np(1-p)}} > \frac{60 - np}{\sqrt{np(1-p)}}\right)$$

Standardize

$$= P\left(\frac{X - 50}{5} > \frac{60 - 50}{5}\right)$$

$$= P(Z > 2)$$

$\downarrow$  standard

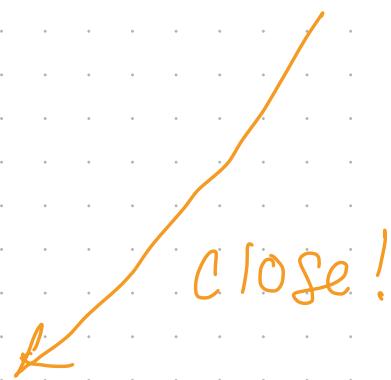
$$= 1 - P(Z \leq 2)$$

$$= 1 - \Phi(2)$$

Chart

$$= 1 - .9772$$

$$= 0.0228$$



→ next exponential R.V.