

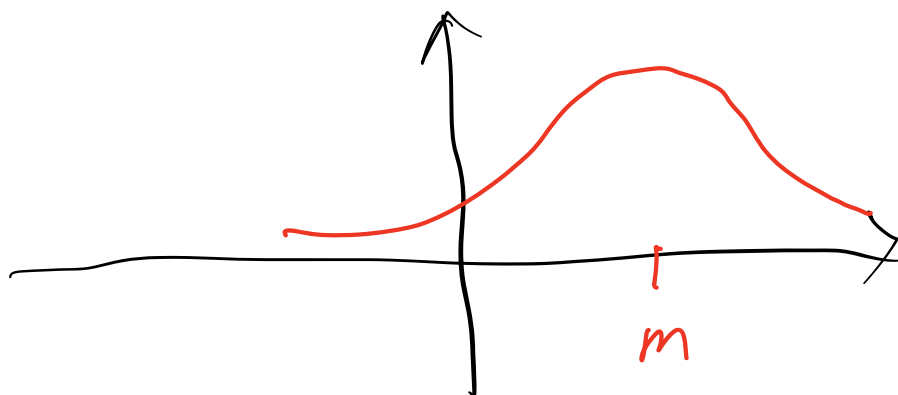
## Lec 10:

Gaussian pdfs

$$f_x(u) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u-m}{\sigma}\right)^2}$$

Notation

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



Special Case:  $m=0$ ,  $\sigma^2=1$

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

Is called a standard or unit Gaussian (or normal)

— Prove the constant in front is correct.

Show integrates to 1

$$\text{Let } f_x(u) = C e^{-\frac{1}{2}\left(\frac{u-m}{\sigma}\right)^2}$$

$$\text{Know } 1 = \int_{-\infty}^{\infty} f_x(u) du$$

$$= C \cdot \int_{-\infty}^{\infty} e^{-\frac{1}{2} \left( \frac{u-m}{\sigma} \right)^2} du$$

Substitution  $v = \frac{u-m}{\sigma} \quad dv = \frac{du}{\sigma}$

$$= C \int_{-\infty}^{\infty} e^{-v^2/2} (\sigma dv)$$

$$= \sigma C \cdot \int_{-\infty}^{\infty} e^{-v^2/2} dv$$

$$\frac{1}{\sigma C} = \int_{-\infty}^{\infty} e^{-v^2/2} dv$$

Let  $I = \int_{-\infty}^{\infty} e^{-v^2/2} dv$

Trick:

$$I^2 = \left( \int_{-\infty}^{\infty} e^{-v^2/2} dv \right)^2$$

$$= \left( \int_{-\infty}^{\infty} e^{-x^2/2} dx \right) \left( \int_{-\infty}^{\infty} e^{-y^2/2} dy \right)$$

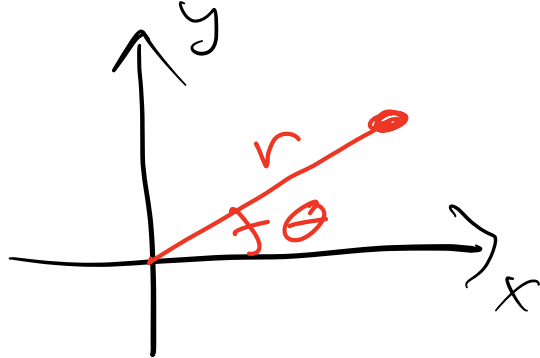
$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-(x^2+y^2)/2} dx dy$$

Convert to polar coordinates

$$r^2 = x^2 + y^2$$

$$\int_0^{2\pi} \int_0^{\infty} e^{-r^2/2} r \, dr \, d\theta$$

Do  $\theta$  1st.



$$= \int_0^{\infty} r e^{-r^2/2} \left( \int_0^{2\pi} d\theta \right) dr$$

$$= 2\pi \int_0^{\infty} r e^{-r^2/2} dr$$

$$= 2\pi \left( -e^{-r^2/2} \right) \Big|_0^{\infty}$$

$$= 2\pi (1 - 0) = 2\pi$$

$$\therefore I^2 = 2\pi \Rightarrow I = \sqrt{2\pi}$$

$$\therefore \frac{1}{\sqrt{c}} = I = \sqrt{2\pi}$$

$$C = \frac{1}{\sqrt{2\pi}} \quad \underline{\text{Done}}$$

Expectation (expected value, average, mean)

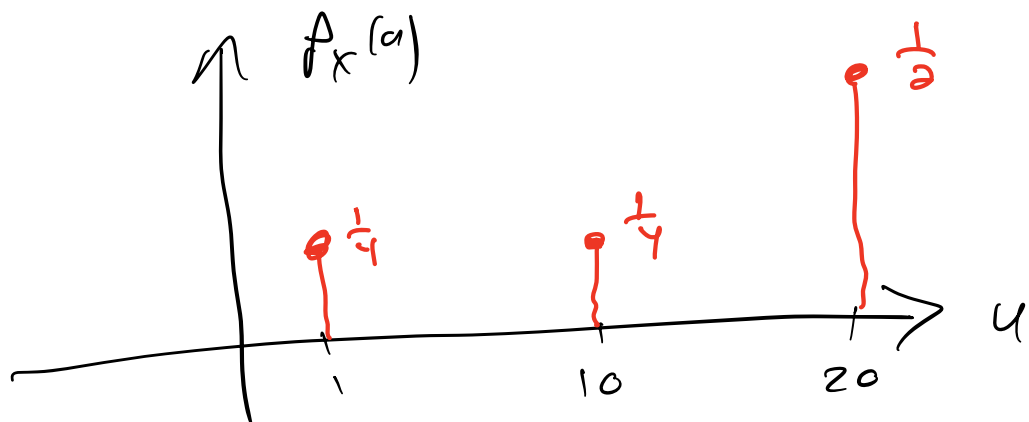
Ex: Given a set of numbers

$$a_1, a_2, a_3, \dots, a_n$$

Their "average" is:

$$\frac{a_1 + a_2 + \dots + a_n}{n}$$

ex: Suppose  $X$  is a discrete r.v.  
with pmf:  $P_X(1) = \frac{1}{4}$ ,  $P_X(10) = \frac{1}{4}$   
 $P_X(20) = \frac{1}{2}$



If we observe  $n$  independent values of  $X$  (for large  $n$ ), you would expect roughly:

$\frac{n}{4}$  1's,  $\frac{n}{4}$  10's,  $\frac{n}{2}$  20's

So the average would be about:

$$\frac{\left(\frac{n}{4} \cdot 1\right) + \left(\frac{n}{4} \cdot 10\right) + \left(\frac{n}{2} \cdot 20\right)}{n}$$

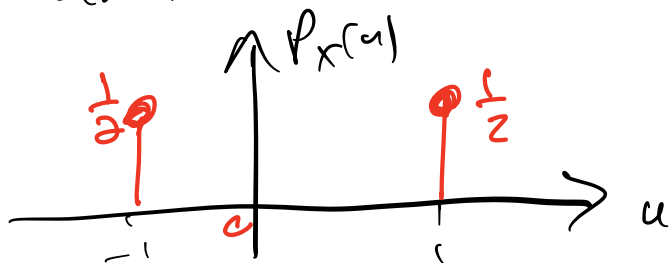
$$= \left(\frac{1}{4} \cdot 1\right) + \left(\frac{1}{4} \cdot 10\right) + \left(\frac{1}{2} \cdot 20\right) = 12.75$$

Def : If  $X$  is a discrete r.v.  
then its expected value (or mean)  
is:

$$E[X] = \sum_u u \cdot p_X(u)$$

Sum is really over only those  $u$   
for which  $p_X(u) \neq 0$ .

Ex :  $X$  is uniform discrete on  $\{-1, 1\}$

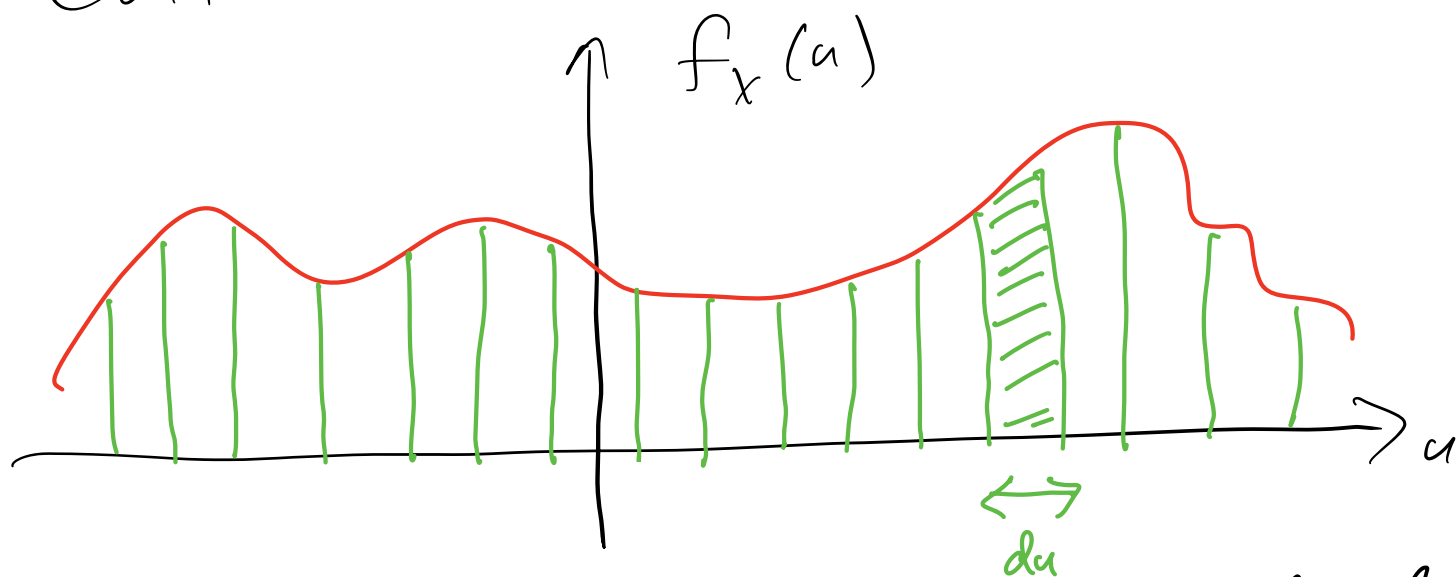


The expected value of  $X$  is:

$$E[X] = (1 \cdot \frac{1}{2}) + (-1) \cdot (\frac{1}{2}) = 0$$

The mean is 0.

Continuous r.v.'s:



As  $du \rightarrow 0$ , the prob of the shaded region is about  $f_X(u) du$

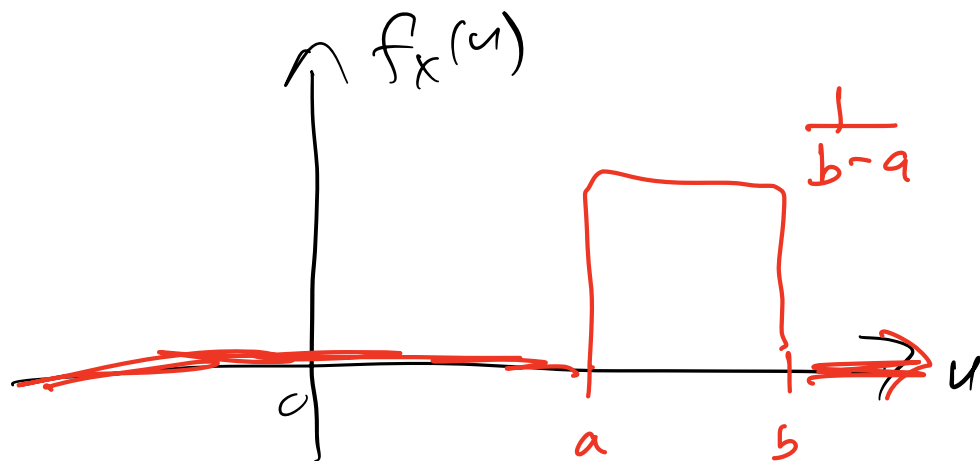
Thus, the mean is about:

$$\sum_u u f_X(u) du \approx \int_{-\infty}^{\infty} u f_X(u) du$$

Def: The expected value of a continuous r.v. is

$$E[X] = \int_{-\infty}^{\infty} u f_X(u) du$$

Ex: Find the mean of a uniform r.v. on  $[a, b]$ .



$$\begin{aligned} E[X] &= \int_{-\infty}^{\infty} u f_X(u) du \\ &= \int_a^b u \cdot \frac{1}{b-a} du = \frac{1}{b-a} \cdot \frac{u^2}{2} \Big|_a^b \\ &= \frac{1}{b-a} \cdot \frac{b^2 - a^2}{2} = \frac{(b-a)(b+a)}{2(b-a)} = \frac{a+b}{2} \\ &= \text{midpoint of interval } [a, b] \end{aligned}$$

Ex: Find mean of a binomial r.v.

Recall pmf:

$$P_X(k) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$k = 0, 1, \dots, n$$

$$E[X] = \sum_u u p_X(u)$$

$$= \sum_{k=0}^n k \cdot \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \sum_{k=1}^n k \cdot \frac{n!}{k! (n-k)!} p^k (1-p)^{n-k}$$

$$= (np) \sum_{k=1}^n \frac{(n-1)!}{(k-1)! (n-1-(k-1))!} p^{k-1} (1-p)^{n-1-(k-1)}$$

Now let  $j = k-1$  and  $m = n-1$   
and substitute

$$= (np) \sum_{j=0}^m \underbrace{\frac{m!}{j! (m-j)!} p^j (1-p)^{m-j}}_{\text{binomial pmf}}$$

$$= (np) \cdot 1$$

$$= np$$

$$\therefore E[X] = np$$

i.e. if you flip a biased coin  $n$  times, the average # of Heads is  $np$ .

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EX: Find the mean of a Poisson r.v.  $X$   
The pmf is-



$$P_X(k) = \frac{e^{-\lambda} \lambda^k}{k!} \text{ for } k=0, 1, 2, \dots$$

$$E[X] = \sum_{k=0}^{\infty} k P_X(k) = \sum_{k=0}^{\infty} k \cdot \frac{e^{-\lambda} \lambda^k}{k!}$$

$$= e^{-\lambda} \cdot \sum_{k=1}^{\infty} \frac{\lambda^k}{(k-1)!} = \lambda e^{-\lambda} \cdot \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!}$$

Substitute  $n = k-1$

$$= \lambda e^{-\lambda} \cdot \sum_{n=0}^{\infty} \frac{\lambda^n}{n!}$$

Taylor Series for  $e^{\lambda}$

$$= \lambda e^{-\lambda} \cdot e^{\lambda} = \lambda$$

$$\therefore \boxed{E[X] = \lambda}$$

Ex : Find mean of a geometric r.v.

The pmf:  $P_X(k) = p(1-p)^{k-1}$

for  $k=1, 2, 3, \dots$

$$E[X] = \sum_{k=1}^{\infty} k p_X(k) = p \cdot \sum_{k=1}^{\infty} k (1-p)^{k-1}$$

Let  $q = 1-p$

$$= (1-q) \cdot \sum_{k=1}^{\infty} k q^{k-1} = (1-q) \cdot \frac{1}{(1-q)^2}$$

$$= \frac{1}{1-q} = \frac{1}{p}$$

$$E[X] = \frac{1}{p}$$

derive from  
derivative of  
geometric sum

Ex: Find mean of Gaussian  
 $X \sim (m, \sigma^2)$

ie. show  $E[X] = m$ .

pdf is  $f_X(u) = \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{1}{2}\left(\frac{u-m}{\sigma}\right)^2}$

$$E[X] = \int_{-\infty}^{\infty} u f_X(u) du$$

$$= \int_{-\infty}^{\infty} u \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u-m}{\sigma}\right)^2} du$$

Let  $y = \frac{u-m}{\sigma}$   $dy = \frac{du}{\sigma}$

Substitute into integral

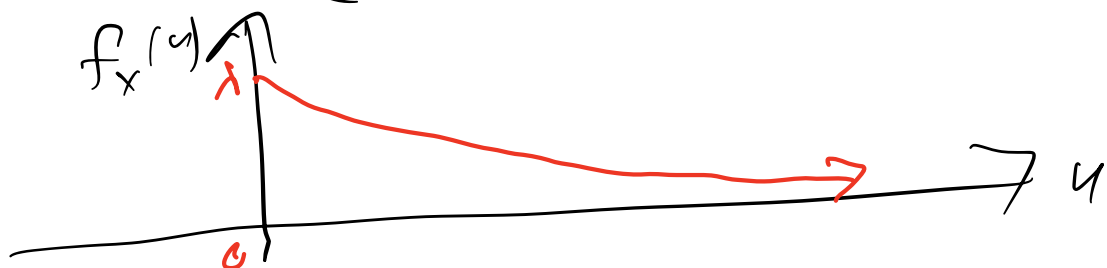
$$\begin{aligned}
 &= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} (\underbrace{\sigma y + m}_u) e^{-y^2/2} (\underbrace{\sigma dy}_{du}) \\
 &= \frac{1}{\sigma\sqrt{2\pi}} \left( \underbrace{\sigma \int_{-\infty}^{\infty} y e^{-y^2/2} dy}_{\substack{= 0 \\ ye^{-y^2/2} \text{ is odd} \\ \int_{-\infty}^0 = -\int_0^{\infty}}} + m \int_{-\infty}^{\infty} e^{-y^2/2} dy \right)
 \end{aligned}$$

$$= m \int_{-\infty}^{\infty} \underbrace{\frac{1}{\sigma\sqrt{2\pi}} e^{-y^2/2}}_{\substack{\text{pdf of} \\ N(0,1) \text{ r.v.}}} dy = m \cdot 1 = m$$

$$E[X] = m$$

Ex : Find mean of an exponential r.v.  
with pdf

$$f_X(u) = \begin{cases} \lambda e^{-\lambda u} & \text{if } u > 0 \\ 0 & \text{else} \end{cases}$$



$$E[X] = \int_{-\infty}^{\infty} z f_X(z) dz$$

$$= \int_0^{\infty} z \cdot \lambda e^{-\lambda z} dz = \int_0^{\infty} u dv$$

Integrate by parts:

Let  $u = z$   
 $du = dz$

$v = -e^{-\lambda z}$   
 $dv = \lambda e^{-\lambda z} dz$

$$= uv \Big|_0^{\infty} - \int_0^{\infty} v du$$

$$= - \underbrace{z e^{-\lambda z} \Big|_0^{\infty}}_{(0-0)} - \int_0^{\infty} -e^{-\lambda z} dz$$

$$= (0-0) + \underbrace{-\frac{1}{\lambda} e^{-\lambda z} \Big|_0^{\infty}}$$

$$= \frac{1}{\lambda}$$

$$E[X] = \frac{1}{\lambda}$$