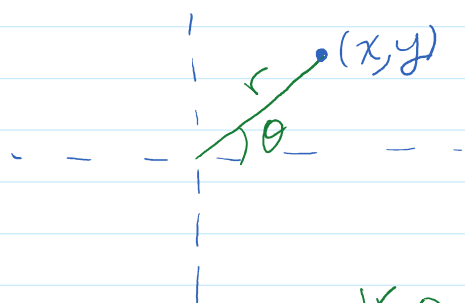
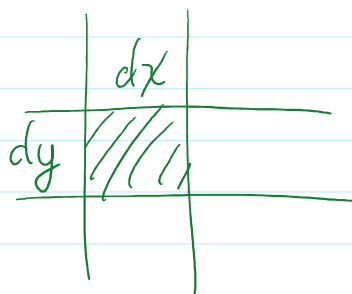
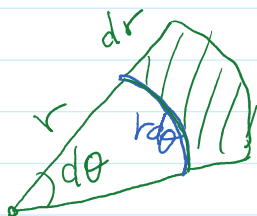


Polar coordinates



$$\begin{cases} x = r \cos \theta \\ y = r \sin \theta \\ x^2 + y^2 = r^2 \end{cases}$$



$$dx dy = r dr d\theta$$

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2}(x^2 + y^2)\right) dx dy = \frac{1}{2\pi} \int_{\theta=0}^{\theta=2\pi} \int_{r=0}^{r=\infty} \exp\left(-\frac{1}{2}r^2\right) r dr d\theta$$

u-substitution

$$u = \frac{1}{2}r^2 \quad \text{then} \quad du = r dr$$

$$\frac{1}{2\pi} \int_0^{2\pi} \int_0^{\infty} \exp(-u) du d\theta$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \left(-e^{-u} \Big|_0^{\infty} \right) d\theta$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \left(-e^{-\infty} - (-e^0) \right) d\theta$$

$$= \frac{1}{2\pi} \int_0^{2\pi} \underbrace{(-e^{-u})}_0 - \underbrace{(-1)}_{-(-1)} d\theta$$

$$-e = (-e)$$

$$= -0 + 1$$

$$= \frac{1}{2\pi} \int_0^{2\pi} d\theta = \frac{1}{2\pi} 2\pi = 1 = I^2$$

so $I = 1$.

Expected Value

If X is a RV then the mean or expected value of X

denoted $E[X]$

is defined as

① discrete

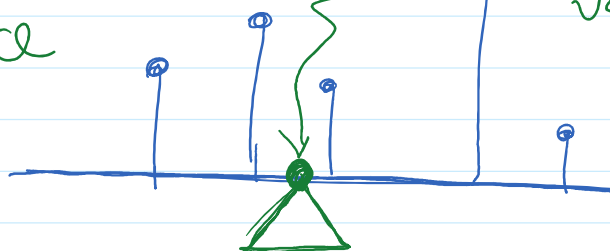
$$E[X] = \sum_{x \in R} x f(x)$$

weighted average
or sum
of the values
in the support
weighted by their
prob. of occurrence

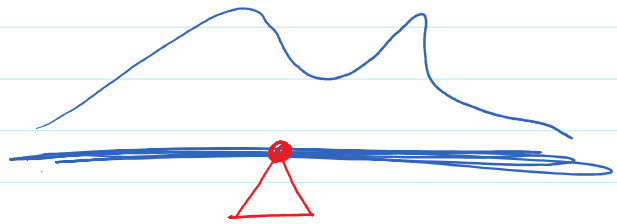
$$= \sum_{x \in \text{Support}(X)} x f(x)$$

PMF

mean/
expected
value



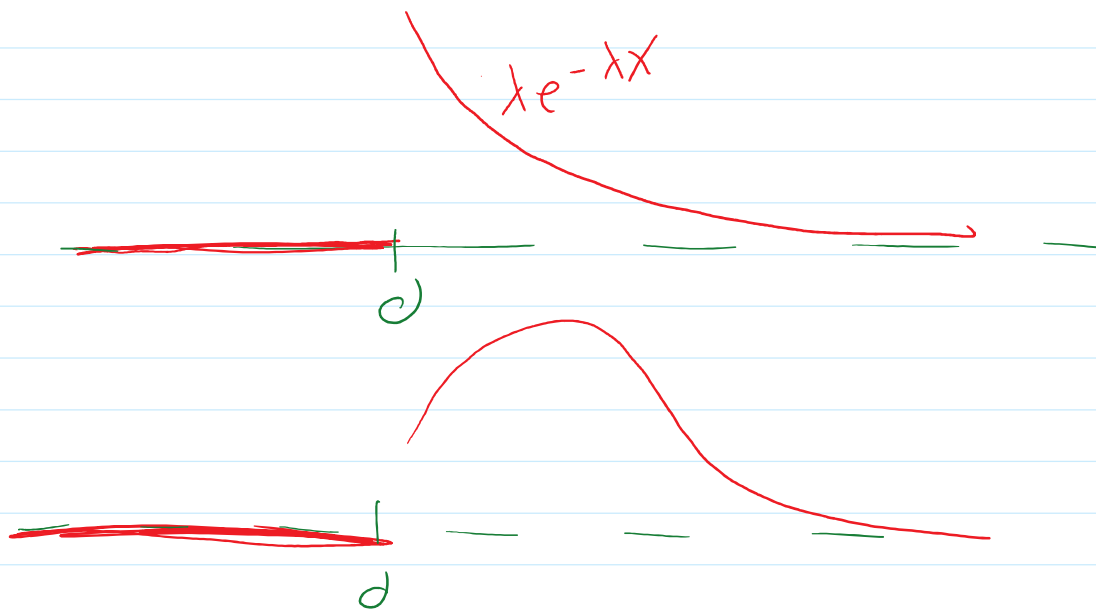
② continuous $E[X] = \int_{\mathbb{R}} x f(x) dx$



Ex.

Let $X \sim \text{Exp}(\lambda)$
 means $\lambda > 0$
 Exponential distribution

$$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$$



Q: what is $E[X]$?

$$E[X] = \int x f(x) dx = \int_0^{\infty} x \lambda e^{-\lambda x} dx$$

$$\mathbb{E}[X] = \int_{\mathbb{R}} x f(x) dx = \int_0^{\infty} \underbrace{x}_u \underbrace{\lambda e^{-\lambda x}}_{dv} dx$$

integration by parts: $u = x$ $dv = \lambda e^{-\lambda x} dx$
 $du = dx$ $v = -e^{-\lambda x}$

$$\int u dv = uv - \int v du$$

$$\rightarrow -xe^{-\lambda x} \Big|_0^{\infty} + \int_0^{\infty} e^{-\lambda x} dx$$

$$= 0 + \left(-\frac{1}{\lambda} e^{-\lambda x} \right) \Big|_0^{\infty}$$

$$= 0 - \left(-\frac{1}{\lambda} \right)$$

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

Ex.

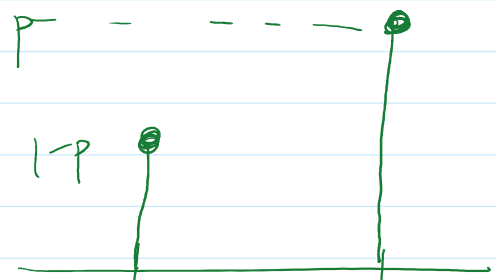
$X \sim \text{Bernoulli}(p)$

$$0 \leq p \leq 1$$

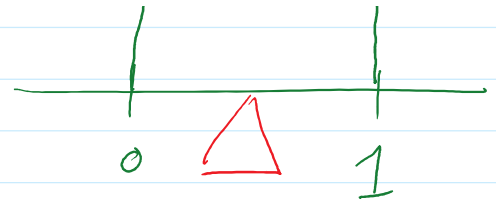
Flip a coin w/ prob. p of H

$X = \# \text{ heads.}$

$$f(x) = \begin{cases} p, & x=1 \\ 1-p, & x=0 \end{cases}$$



$$1-p, x=0$$



$$\begin{aligned} E[X] &= \sum x f(x) = (0)f(0) + (1)f(1) \\ &= (0)(1-p) + (1)p \\ &= p. \end{aligned}$$

Ex. Binomial

$$X \sim \text{Bin}(n, p) \quad \begin{array}{l} 0 \leq p \leq 1 \\ \text{integer } n > 0 \end{array}$$

X = # heads among n independent coin flips w/ a prob. p of H on each

$$f(x) = \binom{n}{x} p^x (1-p)^{n-x} \quad \text{for } x=0, \dots, n$$

Binomial Theorem:

$$(x+y)^n = \sum_{i=0}^n \binom{n}{i} x^i y^{n-i}$$

Justify:

$$\sum_{x=0}^n f(x) = 1$$

$x=p$ and $y=1-p$

$$E[X] = \sum_{x=0}^n x f(x) = \sum_{x=1}^n x \binom{n}{x} p^x (1-p)^{n-x}$$

$$\begin{aligned} x \binom{n}{x} &= \cancel{x} \frac{n!}{\cancel{x!} (n-x)!} \\ &= \frac{n(n-1)!}{(x-1)! ((n-1)-(x-1))!} \\ &= n \binom{n-1}{x-1} \end{aligned}$$

$$\begin{aligned} &= \sum_{x=1}^n n \binom{n-1}{x-1} p^x (1-p)^{n-x} \\ &\quad y = x-1 \Leftrightarrow x = y+1 \\ &= n \sum_{y=0}^{n-1} \binom{n-1}{y} p^{y+1} (1-p)^{n-y-1} \\ &= np \left(\sum_{y=0}^{n-1} \binom{n-1}{y} p^y (1-p)^{(n-1)-y} \right) \end{aligned}$$

PMF of a
Bin(n-1, p)

Sum of PMF of Bin(n-1, p)
over its support
= 1

$$E[X] = np$$

Ex. $n=10$ w/ $p=1/2$

$$E[X] = 5$$

General trick: PMF/PDF trick

Often I can recognize in a calculation

some term is either

$$\sum f(x) \quad \text{or} \quad \int f(x) dx$$

over the support — I can replace this w/ 1.

Functions of RVs.

Note: a function of a RV is also a RV.

Ex. If I have a RV X then

X^2 or $\log X$ is a RV.

Theorem: Law of the Unconscious Statistician

If $g: \mathbb{R} \rightarrow \mathbb{R}$ and X is a RV then

$$E[g(X)] = \begin{cases} \sum_x g(x) f(x) & (\text{discrete}) \\ \int_{\mathbb{R}} g(x) f(x) dx & (\text{cts}) \end{cases}$$

Ex. $X \sim \text{Exp}(\lambda)$

$$f(x) = \lambda e^{-\lambda x} \text{ for } x > 0$$

$$E[X] = 1/\lambda$$

$$E[X^2] = \int_{\mathbb{R}} x^2 f(x) dx = \int_0^{\infty} \underbrace{x^2}_u \underbrace{\lambda e^{-\lambda x} dx}_{dv}$$

$$\begin{aligned} u &= x^2 & dv &= \lambda e^{-\lambda x} dx \\ du &= 2x dx & v &= -e^{-\lambda x} \end{aligned}$$

$$uv - \int v du = -x^2 e^{-\lambda x} \Big|_0^{\infty} + 2 \int_0^{\infty} e^{-\lambda x} x dx$$

$$= \frac{2}{\lambda} \int_0^{\infty} \lambda e^{-\lambda x} x dx$$

$\underbrace{\hspace{10em}}_{E[X] = 1/\lambda}$

$E[X] = 2/\lambda^2$

 $\neq (E[X])^2 = 1/\lambda^2$