

DISTRIBUTIONS (The General ones)

C.D.F (cumulative distribution function)

def. $F(x) = P(X \leq x)$
↑ ↑ ↑
the probability random variable
function

↓
often
used
for discrete
variables
↗ continuous

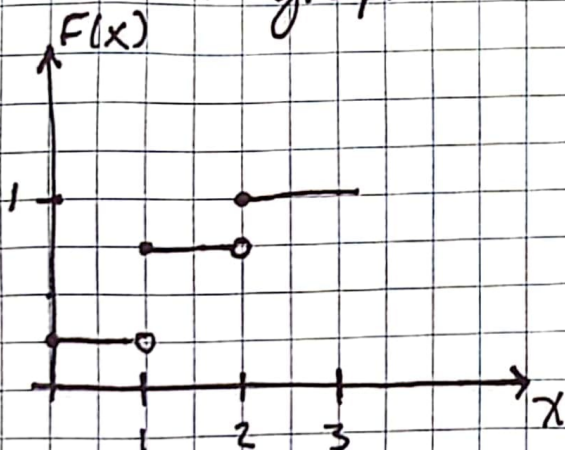
this function is non-decreasing and goes from 0 to 1

Example w/ coin tosses

If we flip 2 coins:

X (# of tails)	0	1	2
$P(X)$	$1/4$	$1/2$	$1/4$

then on graph:

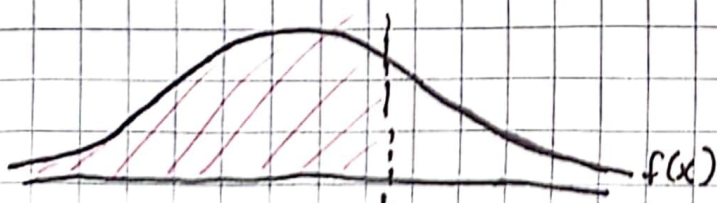


P.D.F (probability density function) → ~~term~~

for continuous
functions

If X is a continuous variable
then $f(x)$ is the pdf if:

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



$f(x)$ is not a probability
the area under $f(x)$
is!

Binomial Distribution

This distribution is used when there are two outcomes of an experiment/trial.
→ often called the success & Failure

The probability of success:

$$P(X=i) = \binom{n}{i} p^i (1-p)^{n-i}$$

or expanded:

$$P(X=x | n, p) = \left(\frac{n!}{x!(n-x)!} \right) p^x (1-p)^{n-x}$$

↑ ↑ ↑
of success # of trials probability of success

Poisson Distribution

- events occur independently
- prob. that an event occurs in a given length of time does not change

Random & Independent

$$P(X=x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

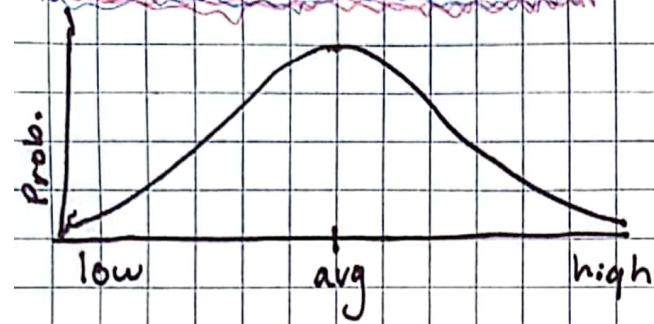
λ is mean of x s

$$E[X] = \text{Var}(X) = \lambda$$



the expected value
& the variance are
the same

Normal Distribution



much more likely to make an observation around the mean

the center is always the avg.

The curve can be narrower or wider based on the std. deviation

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

E.D.F Empirical Distribution

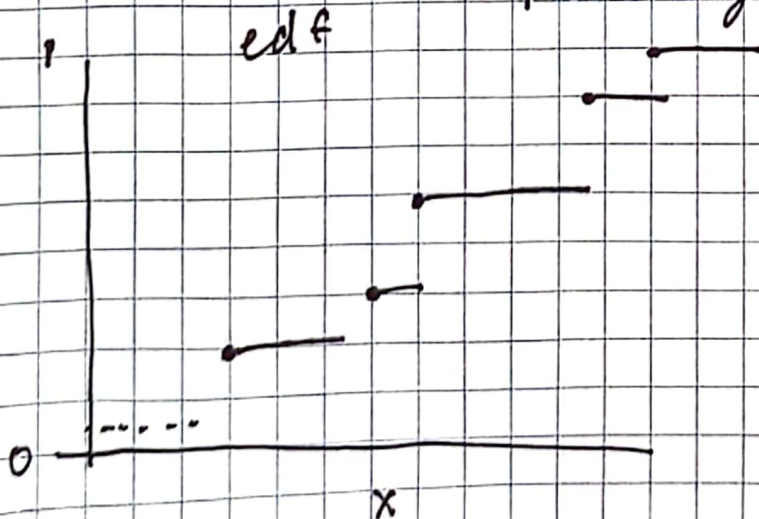
$$\hat{F}_n(x_0) = \frac{\text{number of } X_i \leq x_0}{\text{total observations}} = \sum_{i=1}^n \frac{I(X_i \leq x_0)}{n}$$

$$\frac{1}{n} \sum I(X_i \leq x_0)$$

$$\hat{F}(x_0) = P(X_i \leq x_0)$$

An estimate of this is the ratio that is occurred in the sample

probability of the event $X_i \leq x_0$



Looks "block-y" for small samples.