

## BASICS of PROBABILITY...

def: RANDOM VARIABLE: object  $X$  defined by:

- 1) sample space - set of possible values (or states)
- 2) probability distribution defined over sample space

def: DISCRETE RANDOM VARIABLE: Random variable with DISCRETE sample space.  $\{S_1, S_2, \dots, S_n\}$

We will take  $S_j$  to have NUMERICAL values. [ So expect. val  $\hat{=}$  etc make sense ]

def: Realization: random assignment of  $X$  to one of its states, with specified probabilities

INTERPRETATION: For discrete random var.,  
proba.  $P(S_j) = \text{fraction of realizations when } X=S_j$

### • COIN TOSsing AND BINARY-VALUED RANDOM VARIABLES:

$X$  : <sup>binary</sup> random variable.

Sample space

DIFF NOTATION:  $= \{S_1, S_2\}$   
 $= \{1, 0\}$   
EQUIVALENT.

PROBABILITIES:

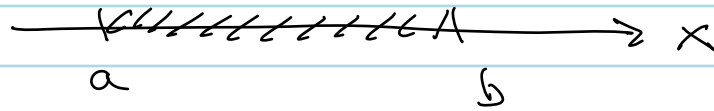
$P(X=S_1) \quad \vdots \quad \dots$   
 $P(X=1)=H \quad ; \quad P(X=0)=T \quad \uparrow$

FAIR coin:  $H=1/2$ ,  $T=1/2$

UNFAIR coin:  $H \in [0, 1]$  (any value),  $T=1-H$ .

..  $\rightarrow$  continuous-valued random variables.  $\leftarrow$

- Sample space =  $[a, b]$ , continuous range



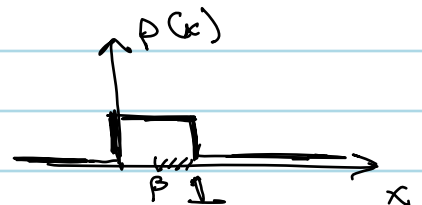
- proba. distribution given by probability density  $p(x)$ :

$$P(x \in [l, r]) = \int_l^r p(x) dx$$

Ex 1. Uniformly distributed random variable with range,  $[0, 1]$ . (sample space) ✓

$$p(x) = \begin{cases} 1 & x \in [0, 1] \\ 0 & \text{otherwise.} \end{cases}$$

$$P(X > \beta) = 1 - \beta$$





$$p(x) = \frac{1}{\sqrt{2\pi} \sigma} \exp\left(-\frac{(x-m)^2}{2\sigma^2}\right)$$

- average  $m$  (or mean, expectation - later!)
- "standard deviation"  $\sigma$

Generating random numbers (realizations of random variables) computationally:

(NOTE about pseudorandom #s from D.H. Lehmer)

Idea: create a sequence of integers  $\hat{x}_k$  b/w 1 and  $m-1 \dots$   
in "random" order

• Multiplicative congruential algorithm:

$$x_{k+1} = \text{mod}(a x_k, m)$$

$\text{mod}(y, m) = \text{remainder}$  after division by  $m$

Ex 1

$$a = 13$$

$$m = 31$$

$$x_0 = 1$$

$$\rightarrow x_1 = 13, x_2 = \text{mod}(13 \cdot 13, 31) = 14, x_3 = 14, \dots$$

$$27, 10, 6, 16, \dots$$

Get a nonrepeating sequence of integers in range  $\{1, \dots, m-1\}$

Then, sequence repeats.

To get uniform distribution on  $[0, 1]$ , let realizations

$$\underline{\bar{x}_k = \frac{x_k}{m}}$$

Ex 2

$$a = 7^5 = 16807$$

[Park & Miller, 1998]

$$m = 2^{31} \approx 2 \times 10^9 \dots$$

get nonrepeating seq. in range  $\{1, \dots, m-1\}$  again

Sequence of length  $m-1$ , then repeats -  
improved!

Say "period" is  $m-1$ .

In MATLAB, rand implements more-sophisticated version —  
period  $\approx 2^{1492}$  ?

Notes :

- Need to specify different starting point (like  $x_0$ ) through random seq. :

\* Then, SYNTAX :  $x = \text{sand}$

or---  $xlist = \text{rand}(1, n)$   
 $\uparrow \quad \uparrow$  # columns.  
 # rows

—, —

Generating...

Generating ...  
 $\rightarrow$  Binary-valued random variable.  $\begin{cases} P(X=1) = t \\ P(X=0) = (1-t) \end{cases}$

for FALR can't.

$$(H = \frac{1}{2})$$

$\left\{ \begin{array}{l} \gg x = \text{round}(\text{rand}) \quad , \quad \text{or} \\ \gg x_{\text{list}} = \text{round}(\text{rand}(1, n)) \end{array} \right.$

Normally distributed random variable  $\dots$  mean 0, standard-dev. 1  
More-complex transformation of rand  $\dots$  implemented  
by

$$x_i \sim \text{randn}$$

```
>> xlist = random (1, n)
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Mean  $\mu$ , std. dev.  $\sigma$ :  $\gg X = \mu + \sigma \cdot \text{randn}$

To generate binary-valued random variables w/  $H \neq 1/2$ , use  
logical operators:

$>$ ,  $<$ ,  $==$ ,  $!=$ ,  $>=$ ,  $<=$ ,  
(and there are others:  $\&$ ,  $|$ ) ...

Syntax:  $\underbrace{a > b}_{\text{logical statement}}$  returns  $\begin{cases} 1 & \text{if logical statement is true} \\ 0 & \text{if false} \end{cases}$

Therefore, general syntax for binary random  
var w/  $P(X=1) = H$ :

$$X = (\text{rand} > (1-H))$$

Another, simpler way:

$$X = \text{round} \left( \text{rand} + (H - 1/2) \right)$$

↑  
Make up more-than-fair proba-

## Variance and standard deviation of random variables:

LET  $X$  be any random variable with any probability distribution

### Recall:

- Expectation, <sup>mean</sup> average, of discrete random variable  $X$  is
$$\bar{X} = E(X) = \sum_{j=1}^N s_j P(s_j)$$

$s_j$  are numerical values of  $X$ .

then

$$E(X) + b E(Y)$$

- Note, exist similar definitions for cts-valued r.v.

- Variance of  $X$  measures fluctuations of  $X$  around  $\bar{X}$

$$\text{var}(X) = E\left((X - \bar{X})^2\right)$$

Average squared variation of  $X$  from mean

For discrete random variable :

$$= \sum_{j=1}^N P(S_j) (S_j - \bar{X})^2$$

Ex 1. Let  $X$  be binary random variable,  $P(X=1) = H$   
 $P(X=0) = (1-H)$

$$\bar{X} = H$$

$$\begin{aligned} \text{var}(X) &= H \cdot (1 - \bar{X})^2 + (1-H) (0 - \bar{X})^2 \\ &= H \cdot (1-H)^2 + (1-H) H^2 \\ &= H (1 - 2H + H^2) + H^2 - H^3 \\ &= H - 2H^2 + H^3 + H^2 - H^3 \\ &= H - H^2 = \underline{\underline{H(1-H)}} \end{aligned}$$

skip

Ex 2 | Let  $Y$  be binomial r.v. with parameters  $(N, H)$ .

Fact:  $\text{var}(Y) = N H (1-H)$



Getting mean and variance from data (collection of samples, or realizations, of  $X$ )

Approximate from  $M$  samples: Say  $X = S_k$  on  $k$ th sample

$$E(X) = \bar{X} \sim \frac{1}{M} \sum_{k=1}^M S_k$$

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ASIDE ABOUT VARIANCE:

$$\begin{aligned} \text{var}(X) &= E(X - \bar{X})^2 = E(X^2 - 2X\bar{X} + \bar{X}^2) = \\ &= E(X^2) - E(2X\bar{X}) + E(\bar{X}^2) \\ &= E(X^2) - 2\bar{X}E(X) + E(\bar{X}^2) \\ &= E(X^2) - 2\bar{X}^2 + \bar{X}^2 \\ &= \underline{E(X^2) - \bar{X}^2} \end{aligned}$$

Exercise: check steps using definition of  $E(X)$

$$\text{var}(X) \sim \frac{1}{M} \sum_{k=1}^M \left[ S_k - \left( \frac{1}{M} \sum_{k=1}^M S_k \right) \right]^2$$

• Implementation in MATLAB:

- Say have sample-list =  $(x_1, x_2, \dots, x_M)$

mean(sample-list)

var(sample-list)

Start with a list of samples. (Don't know - continuous/discrete / what states are, etc...)

How to plot a proba. distribution from sample-list - - -

IDEA: ~~xxx | xxx | xxx~~ Break states into RANGES - "bins"

MATLAB:  $[nlist, centerlist] = hist(sample-list, numbins)$

number of  
samples that  
fall into  
each bin.

bin centers

$[c_1, c_2, c_3, \dots]$

uses THAT  
many bins.

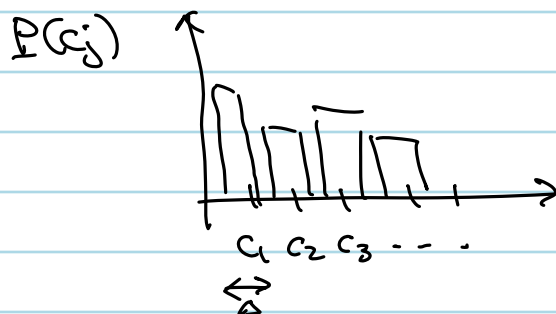
$[n_1, n_2, n_3, \dots]$

1) Discrete random variable interpretation.

Let each  $c_j$  correspond to one of the possible states  
with which we describe r.v.

$P(c_j) = n_j / M$  ← frequency of occurrence, over  
the  $M$  samples

To plot:  $\gg bar(centerlist, nlist/M)$



See MATLAB CODE:  
hist-demo.m

2) Continuous random variable interpretation:

Approximate Proba. density:  $p(x)$ .

Then, if spacing  $c_{j+1} - c_j = \Delta$ ,  $P(c_j) \approx p(c_j) \Delta$

→ To plot PDF,  $\gg bar(centerlist, nlist / (M \cdot \Delta))$