

14-03 Discrete RVs.

21-03 Special discrete distributions

28-03 Revision

30-03 Start of midterm

Probability and Statistics

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Outline

Part 2 – Random variables

2.1 Introduction

2.2 Discrete random variables

- Probability Mass Function (PMF)
- Cumulative Distribution Function (CDF)
- Expectation
- Variance and Standard deviation

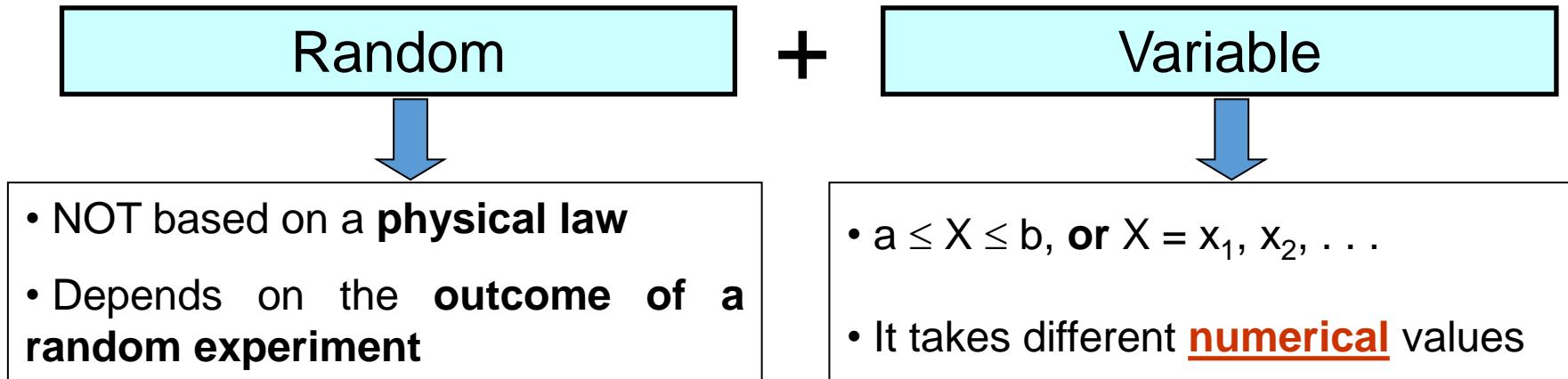
2.3 Some important discrete PMF's

- Bernoulli PMF

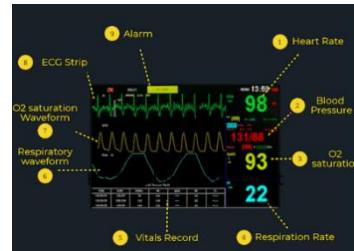
2.1 Introduction

2.1 Introduction

What is a random variable (RV)



Examples



Heart rate

Discrete



Queue length



Call duration

Continuous



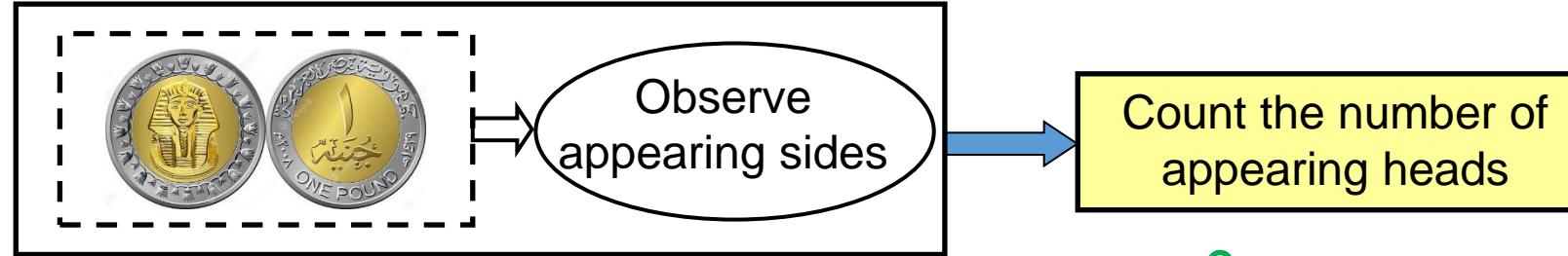
Life time

From random experiment to RV's

Example

A coin is tossed twice and the appearing sides are observed. Let X be the number of appearing heads.

Find the possible values of X and the probability of each value.



$\begin{array}{c} H \quad < \quad T \\ | \quad \quad \quad | \\ T \quad < \quad H \end{array}$

$$S = \{ \underbrace{TT}_0, \underbrace{TH}_1, \underbrace{HT}_1, \underbrace{HH}_2 \}$$

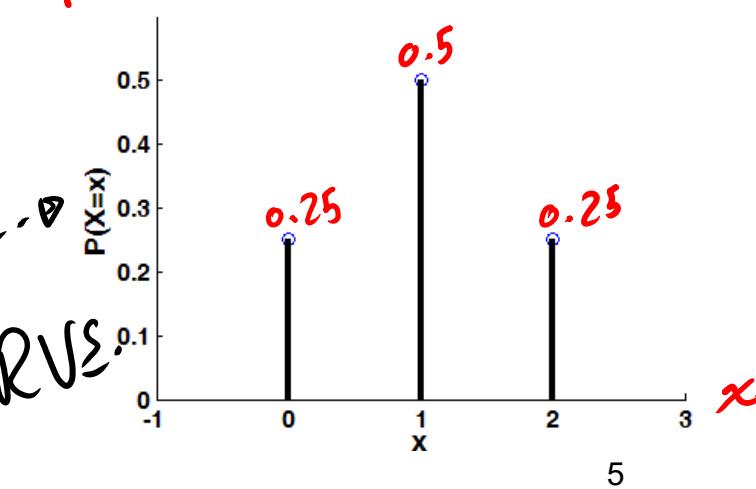
$\sum \text{Prob.} = 1$

x	0	1	2
P(X = x)	1/4	1/2	1/4

probability mass function
"pmf"

→ Discrete RVs

Capital "RV"
 $P(X=x)$
small "value taken by the RV"



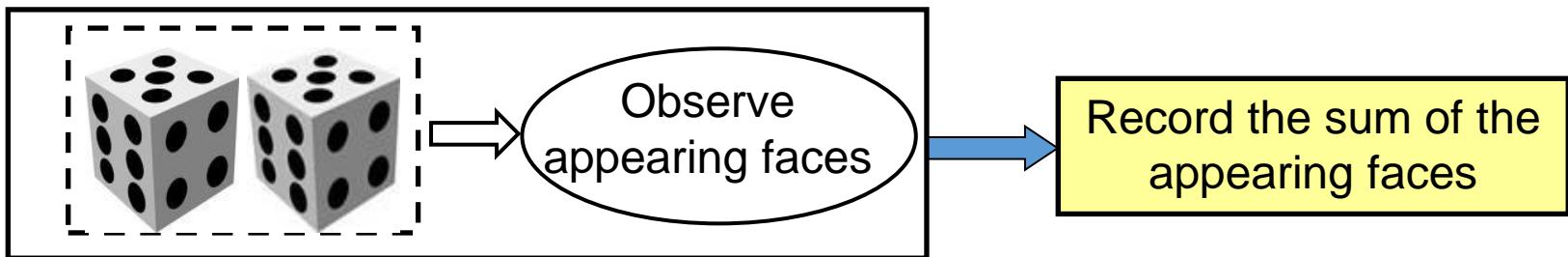
From random experiment to RV's (Cont'd)

Example

A die is thrown twice.

Let X denotes the sum of the appearing numbers.

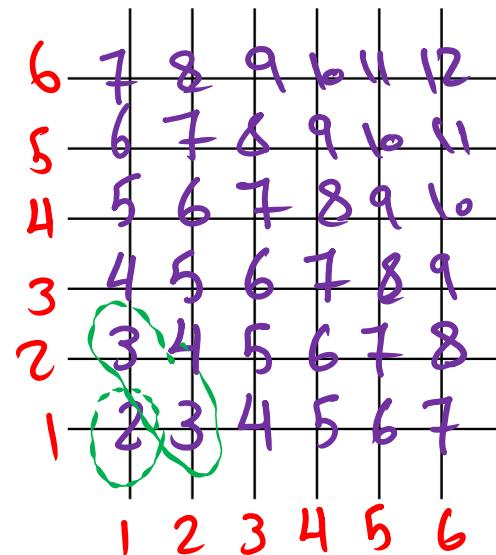
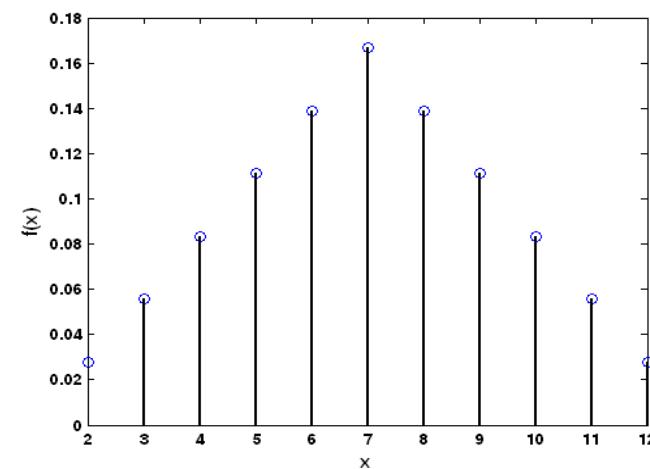
Find the possible values of X and the probability of each value.



each element is a subproblem

X	2	3	...	12
$P(X = x)$	1/36	2/36	...	1/36

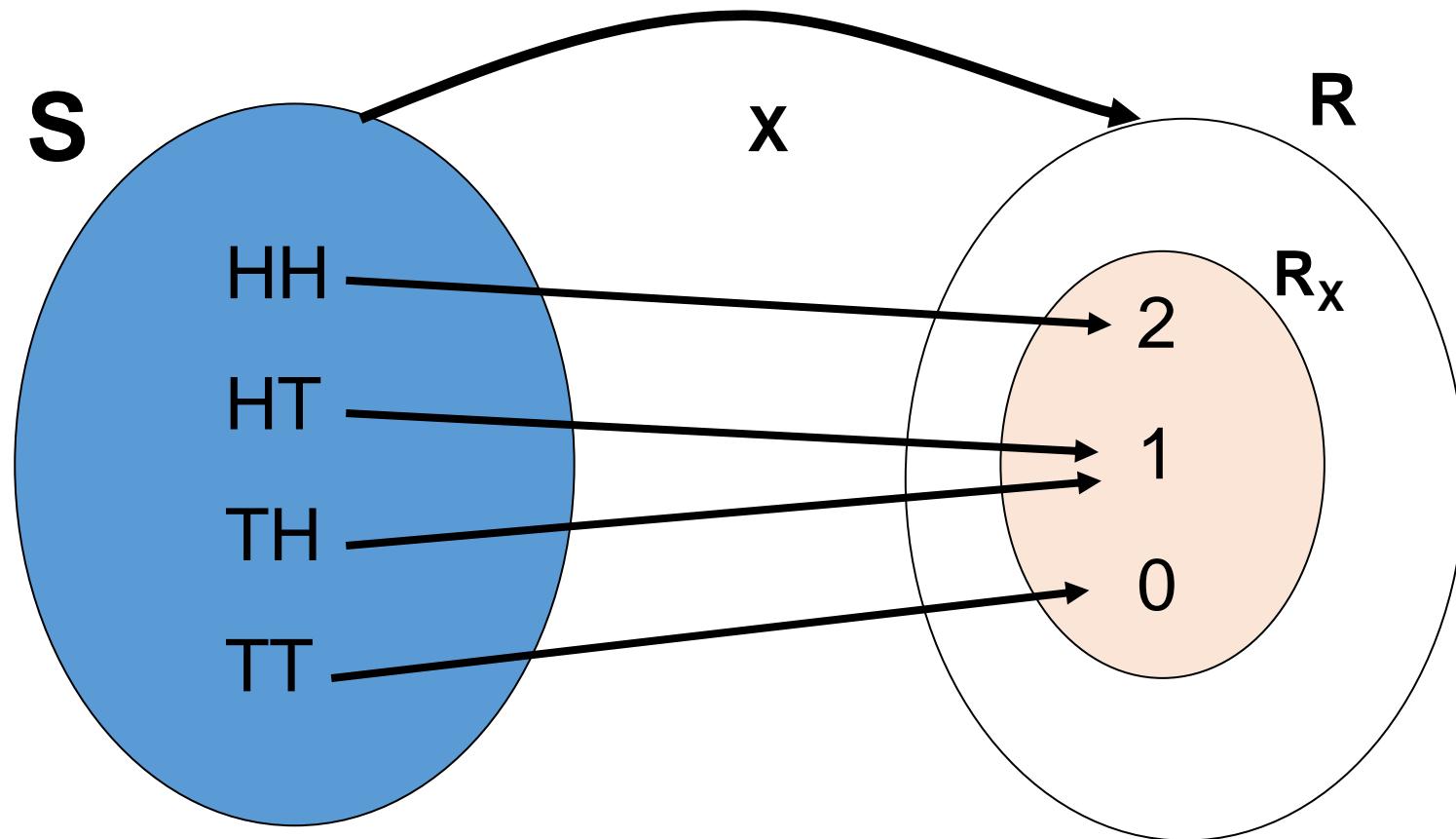
$$\sum \text{prob.} = 1$$



pmf of X

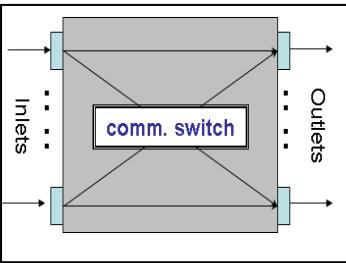
Mathematical definition of a RV

A *random variable* is a real function defined on the sample space.



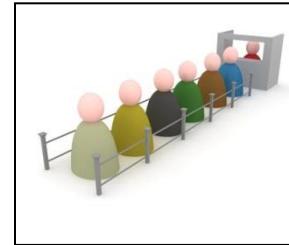
RV's classification

Discrete
RV



X: Number of packets/min

$$R_X = \{0, 1, 2, \dots, N\}$$



X: Queue length

$$R_X = \{0, 1, 2, \dots\}$$

✓
Before
midterm

Continuous
RV



X: Call duration

$$R_X = [0, \infty]$$



X: Life time

$$R_X = [0, T]$$

2.2 Discrete random variables

2.2.1 Probability Mass Function (PMF)

2.2 Discrete random variables

Probability mass function (PMF)

Example

An arriving call is equally likely to be voice or data.

Observe three calls which are assumed to be independent.

Let X be the number of voice calls. Find the PMF of X.



$$S = \{ \text{DDD}, \text{DDV}, \dots, \text{VVV} \}$$

$$N_S = 2^3 = 8$$

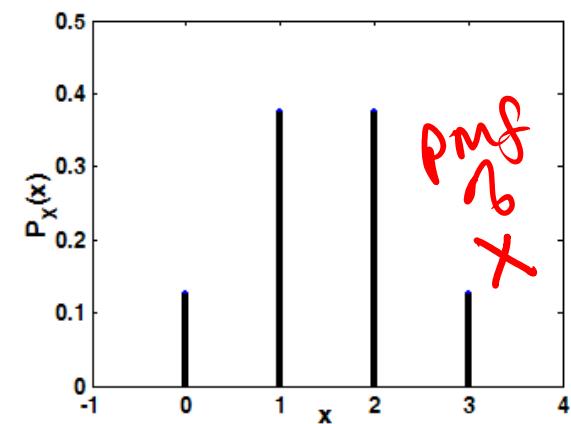
x	0	1	2	3
P _X (x)	1/8	3/8	3/8	1/8

DDD VDD DVV VVV
DVD VDV VVD VVV
DDV VVD VVV
 3C_1 3C_2

Definition
 $P_X(x) = P(X = x)$

General properties

- (1) $P_X(x) \geq 0$, (2) $\sum_x P_X(x) = 1$



Example

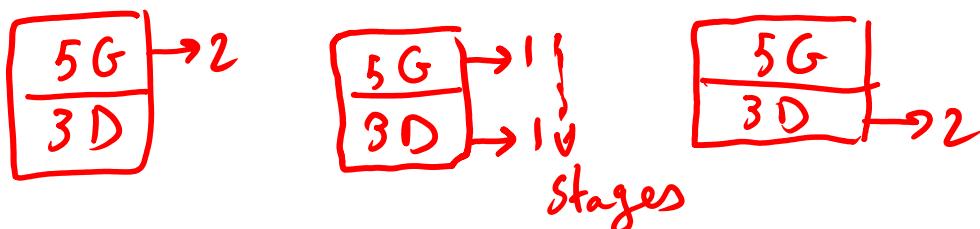
A shipment of ^{Total} 8 microcomputers to retail outlet contains 3 that are defective.
If a school makes a random purchase of two of these computers. Find the probability distribution for the number of defectives.

8

$$\begin{array}{c} \boxed{5G} \\ \hline 3D \end{array} \xrightarrow{\text{Select}} 2 \Rightarrow N_s = {}^8C_2$$

X: no. of defectives

X	0	1	2
P _X (x)	$\frac{{}^5C_2}{{}^8C_2}$	$\frac{{}^5C_1 * {}^3C_1}{{}^8C_2}$	$\frac{{}^3C_2}{{}^8C_2}$



Example

The probability of a successful connection to a call center is 0.8.

You decided to try at most four times. Trials are independent.

Let X be the number of trials. Find $P_X(x)$.

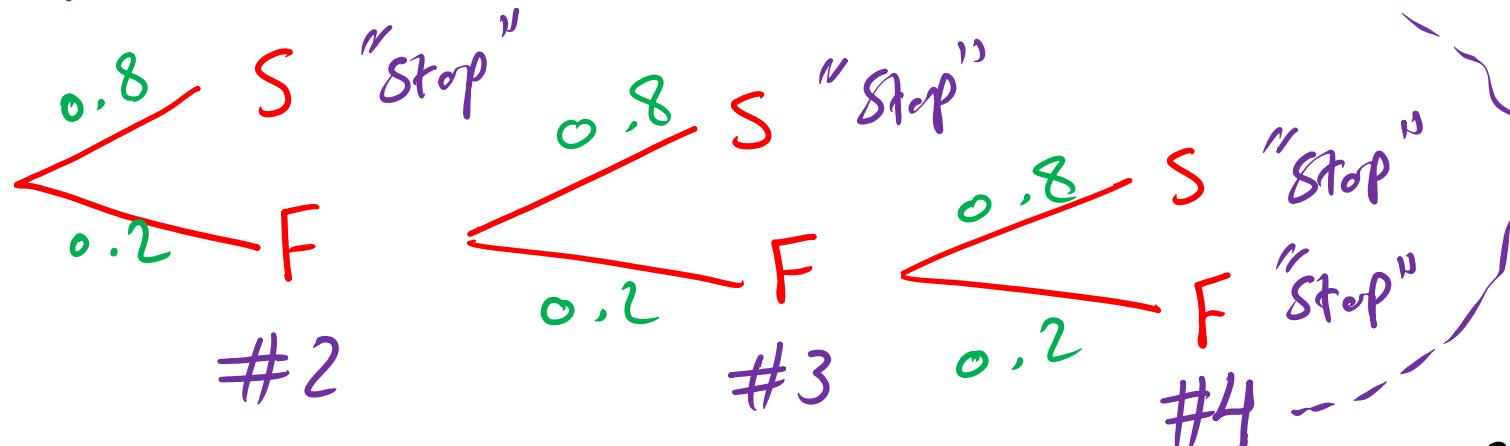
Default :
independent
trials

0.8 S "success" "stop"
0.2 F "fail"

#1

X :
number of
trials

x	1	2	3	4
$P_X(x)$	0.8	0.2×0.8	$(0.2)^2 \times 0.8$	
	S	FS Stages	FFS	FFFF or FFF



Alternatives $(0.2)^3 \times 0.8 + (0.2)^4$

$\sum \text{probs} = 1$

$1 - [0.8 + 0.2 \times 0.8 + (0.2)^2 \times 0.8]$

Example

The probability of a successful connection to a call center is 0.8.

You decided to try at most four times. Trials are independent.

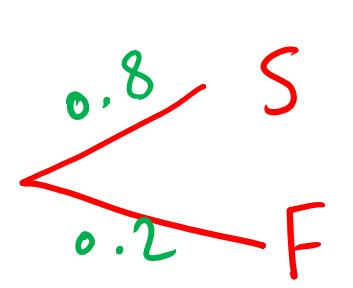
	# Success	# F
S	1	0
FS	1	1
FFS	1	2
FFFS	1	3
FFFF	0	4

Let X be the number of successes.

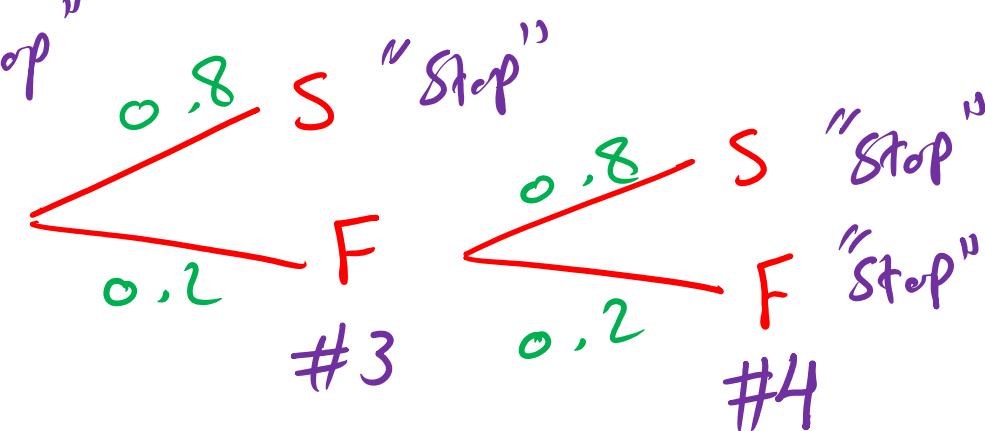
Find $P_X(x)$.



#1



#2



#3

#4

x	0	1
$P_X(x)$	$(0.2)^4$	

FFFF

$$\rightarrow 1 - (0.2)^4$$

x	0	1	2	3	4
$P_X(x)$	0.8	0.2×0.8	$(0.2)^2 \times 0.8$	$(0.2)^3 \times \frac{3}{0.8}$	$(0.2)^4$

Example

Let $P_X(x) = k(0.2)^x$, $x = 1, 2, 3, \dots$

(a) Find k .

$$\sum_{\forall x} P_X(x) = 1$$

Hint:

$$\sum_{n=0}^{\infty} a^n = \frac{1}{1-a} = 1 + a + a^2 + a^3 + \dots$$

$$P(X=1) + P(X=2) + P(X=3) + \dots = 1$$

$$\underbrace{k(0.2)^1 + k(0.2)^2 + k(0.2)^3 + \dots}_\text{Common} = 1$$

$$k(0.2) \left[1 + (0.2) + (0.2)^2 + \dots \right] = 1$$
$$k(0.2) \cdot \frac{1}{1-0.2} = 1$$

$$k \frac{0.2}{0.8} = 1$$

$$\boxed{k=4}$$

Example

Let $P_X(x) = k(0.2)^x$, $x = 1, 2, 3, \dots$

(a) Find k .

(b) Compute

$$P(X = 3),$$

$$P(X \geq 2),$$

$$P(X=3) = P_X(3) = 4(0.2)^3 = \checkmark$$

$$P(X \geq 2) = P(X=2) + P(X=3) + \dots$$

$$\text{Ans} = 1 - P(X < 2) = 1 - P(X=1) = 1 - 4(0.2)^1 \\ = 0.2$$

Example

Let $P_X(x) = k(0.2)^x$, $x = 1, 2, 3, \dots$

(a) Find k .

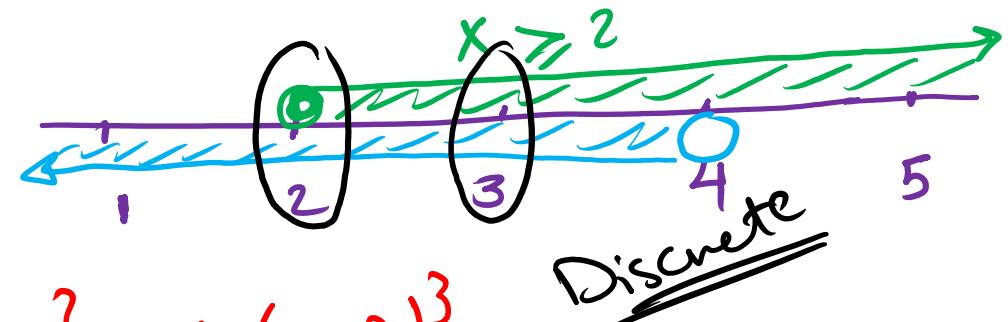
(b) Compute

$$P(X < 4 | X \geq 2).$$

$$\hookrightarrow = \frac{P(X < 4 \cap X \geq 2)}{P(X \geq 2)}$$

$$= \frac{P(X=2) + P(X=3)}{1 - P(X=1)} = \frac{4(0.2)^2 + 4(0.2)^3}{1 - 4(0.2)^1}$$

$$P(A|B) = \frac{\text{intersection}}{\text{condy.}} = \frac{P(A \cap B)}{P(B)}$$



2.2 Discrete random variables

2.2.2 Cumulative Distribution Function (CDF)

$$P(X \leq 5) \Rightarrow F_X(5)$$

$$P(X \leq 100) \Rightarrow F_X(100)$$

$$P(X > 20) \Rightarrow 1 - P(X \leq 20) = 1 - F_X(20)$$

$$P(10 < X \leq 100) \dots \dots$$

$$F_X(x) = P(X \leq x)$$

Easier calculations
for
the probabilities

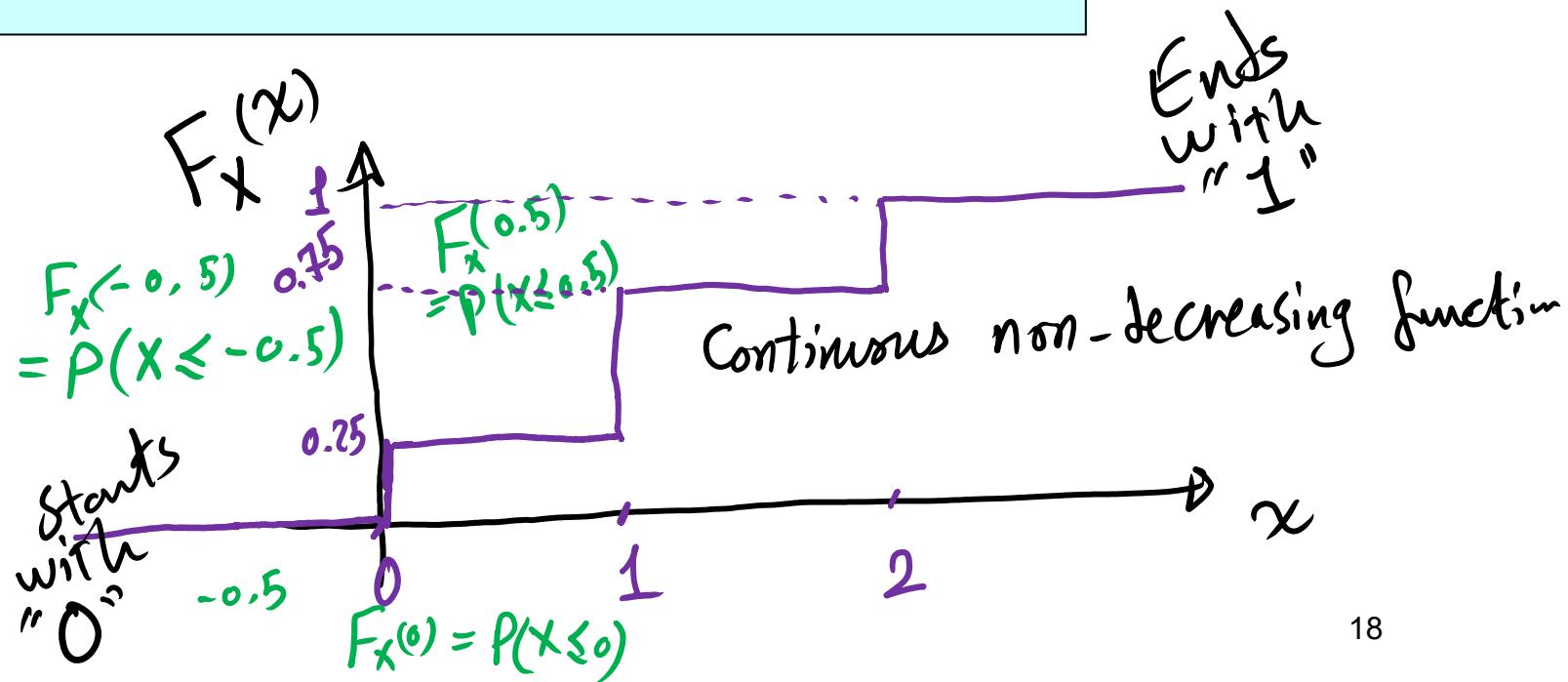
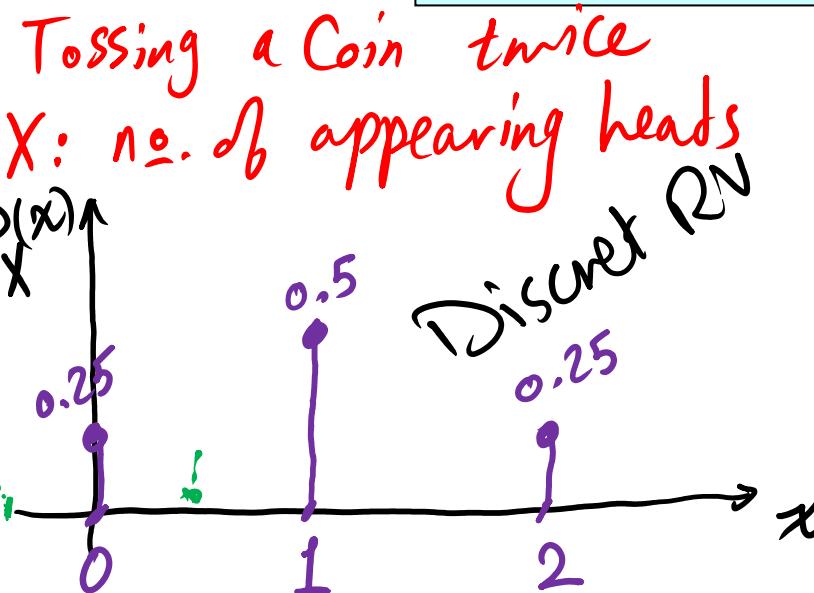
Cumulative distribution function (CDF)

Definition

$$F_X(x) = P(X \leq x)$$

General properties of $F_X(x)$

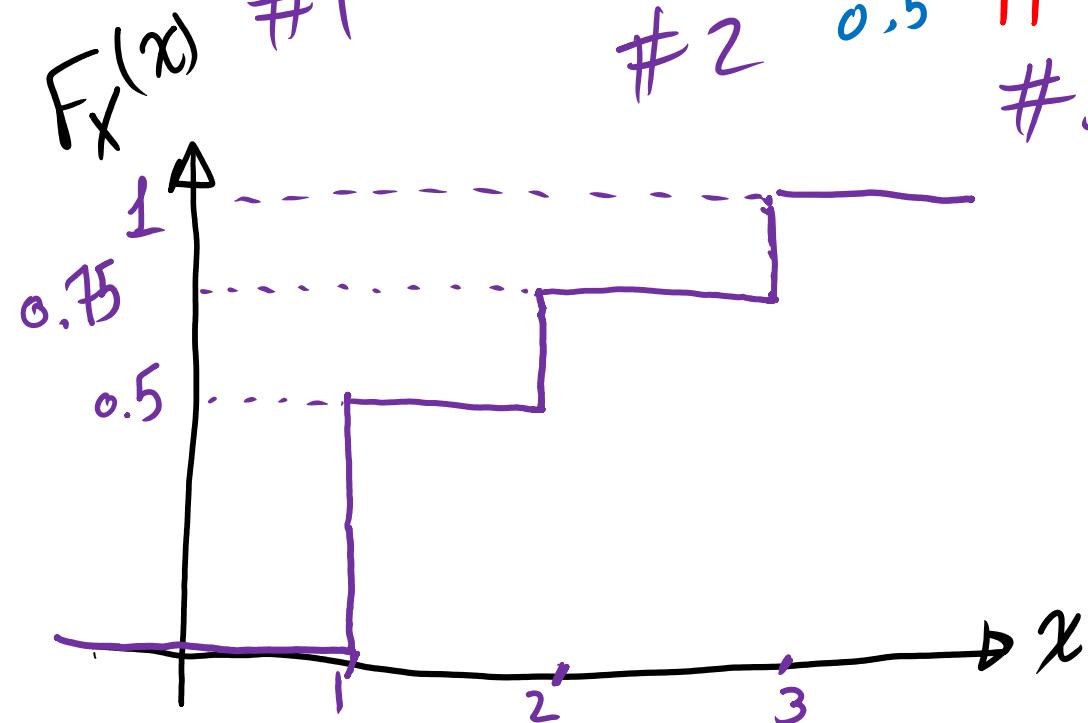
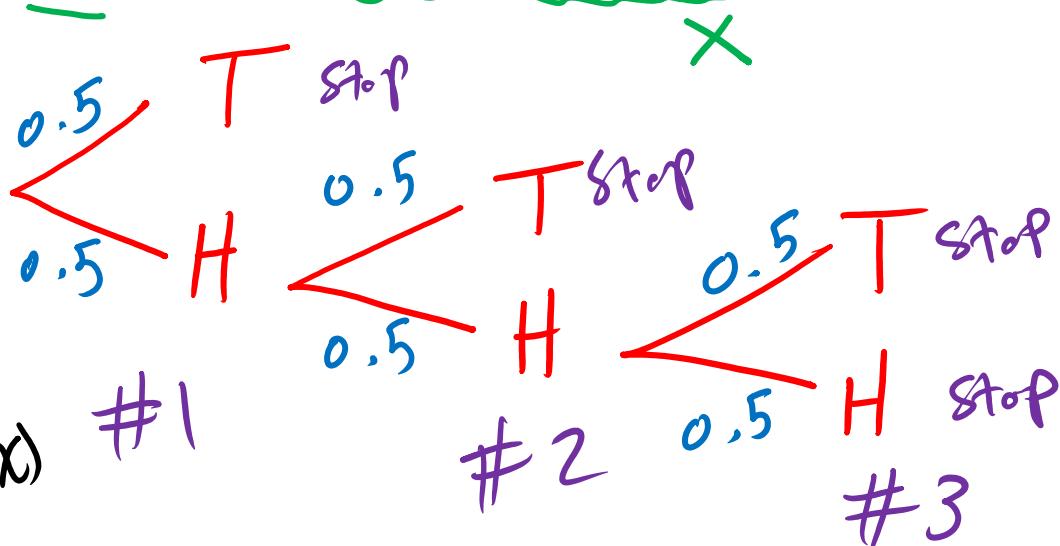
- (1) Non-decreasing function
- (2) $F_X(-\infty) = 0$
- (3) $F_X(\infty) = 1$



Example

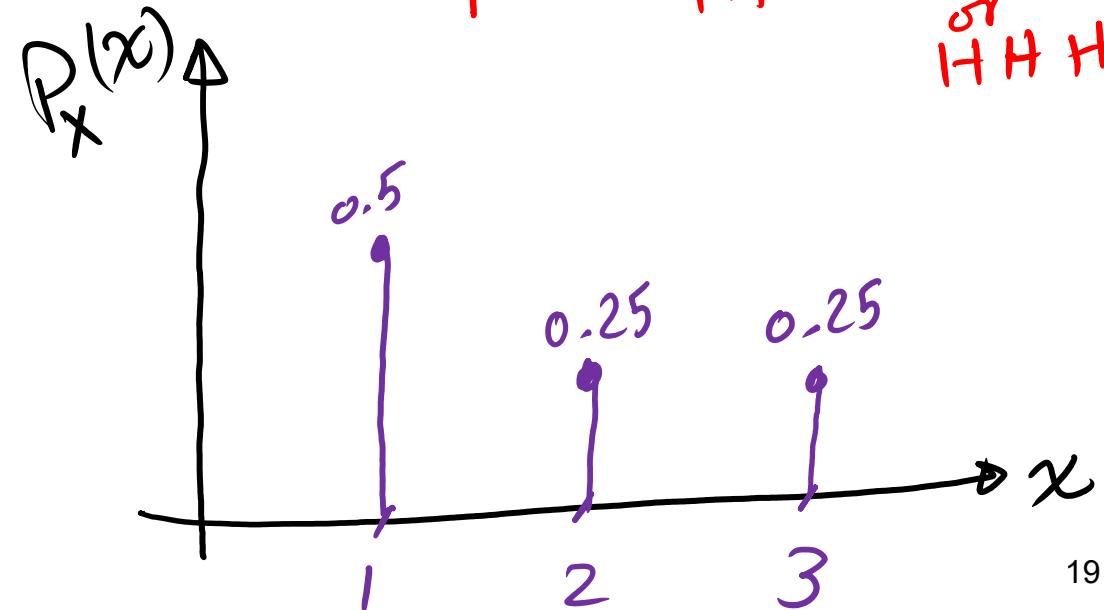
A coin is tossed until a tail appears or after three attempts have been made. Find the PMF and CDF of the number of trials made in this experiment and sketch their graphs

Default: fair coin



X : no. of trials

x	1	2	3
$P_X(x)$	0.5	0.25	$1 - [0.5 + 0.25] = 0.25$
	T	HT	HTT or HHH



Example

Find the CDF of the RV X where

$$P_X(x) = 4(0.2)^x, \quad x = 1, 2, 3, \dots$$

Hint:

$$\sum_{n=0}^{\infty} a^n = \frac{1}{1-a} = 1 + a + a^2 + \dots$$

$$P(X \leq \cancel{x}) = 1 - P(X > \cancel{x})$$

$$= 1 - [P(X = \cancel{1}) + P(X = \cancel{2}) + P(X = \cancel{3}) + \dots]$$

$$= 1 - [4(0.2)^{\cancel{1}} + 4(0.2)^{\cancel{2}} + 4(0.2)^{\cancel{3}} + \dots]$$

$$= 1 - 4(0.2)^{\cancel{1}} \underbrace{[1 + 0.2 + (0.2)^2 + \dots]}_{a=0.2}$$

$$= 1 - 4(0.2)^{\cancel{1}} \frac{1}{1-0.2} = 1 - 5(0.2)^{\cancel{1}} = F_X(x)$$

Example

Find the CDF of the RV X where

$$P_X(x) = 4(0.2)^x, \quad x = 1, 2, 3, \dots$$

then find: $P(X \leq 5) \cdot P(X \leq 100) \cdot P(X > 5) \cdot P(X \geq 5) \cdot P(4 \leq X \leq 20)$

$$P(X \leq 5) = F_X(5) = 1 - 5(0.2)^{5+1}$$

$$P(X \leq 100) = F_X(100) = 1 - 5(0.2)^{100+1}$$

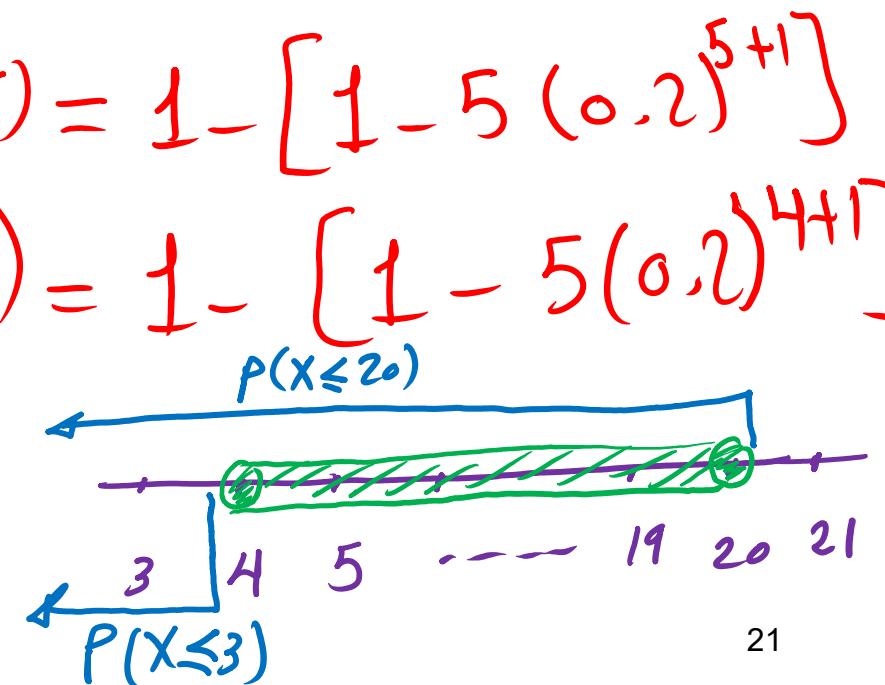
$$P(X > 5) = 1 - P(X \leq 5) = 1 - F_X(5) = 1 - [1 - 5(0.2)^{5+1}]$$

$$P(X \geq 5) = 1 - P(X < 5) = 1 - P(X \leq 4) = 1 - [1 - 5(0.2)^{4+1}]$$

$$\begin{aligned} P(4 \leq X \leq 20) &= P(X \leq 20) - P(X \leq 3) \\ &= F_X(20) - F_X(3) \end{aligned}$$

$F_X(x) = P(X \leq x) = 1 - 5(0.2)^{x+1}$

Note: equal is very important especially in discrete RV's!



Exercise in previous example
 $P(4 < X \leq 20)$

2.2 Discrete random variables

2.2.3 Expectation

Introduction to expectation

Example

A class of 100 students with ages ranging from 20 to 24 years old.

A record of the ages of students is introduced in the following table.

$$\text{Average age} = \frac{21 + 23 + 22}{3}$$

مُجموع الأعمار
عدد الطالب

Age (x)	20	21	22	23	24
Number of students	5	20	50	15	10
$P_x(x)$	5/100	20/100	50/100	15/100	10/100

Q: How to average these numbers?

Motivation

Sum the ages of all students and divide by the number of students

$$\text{Average age} = \frac{20 * 5 + 21 * 20 + 22 * 50 + 23 * 15 + 24 * 10}{100}$$

مُجموع الأعمار
عدد الطالب

The average of X may be computed as follows:

$$20 \times \underbrace{(5/100)}_{x(1)} + 21 \times \underbrace{(20/100)}_{x(2)} + 22 \times \underbrace{(50/100)}_{x(3)} + 23 \times \underbrace{(15/100)}_{x(4)} + 24 \times \underbrace{(10/100)}_{x(5)} = 22.05$$

$$\text{Average mean value} = \text{Expectation} = E(X) = \sum_{\text{All } x} x P_x(x)$$

Expectation or mean ($E(X)$, μ_x)

Definition

$$E(X) = \sum_x x P_X(x)$$

What does $E(X)$ represent?

- A *weighted sum* of the RV values.
- The *average value* of the RV over the *long run*.
- The *balancing point* of the PMF.

Example

Let X be the number of breakdowns/week of a certain machine. Find $E(X)$

x	0	1	2	3
$P_X(x)$	0.6	0.2	0.15	0.05

→ breakdowns/week

$$E(X) = \sum x P_X(x)$$

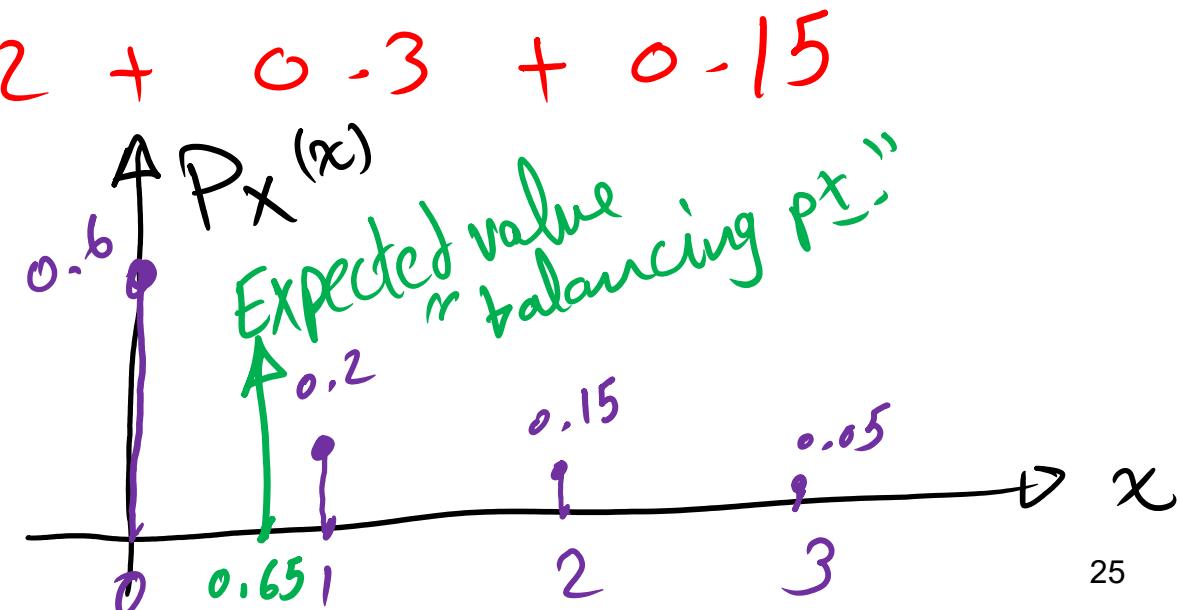
$\forall x$

$$E(X) = 0 * 0.6 + 1 * 0.2 + 2 * 0.15 + 3 * 0.05$$

$$= 0 + 0.2 + 0.3 + 0.15$$

$$= 0.65$$

breakdowns/week



$E(X)$ properties

$E(\text{const.}) =$
The same const.

$$E(aX + b) = aE(X) + b$$

$$E[g(X)] = \sum_x g(x)P_X(x)$$

$$E(X) = \sum_x x P_X(x)$$

$$E(ax+b) = \sum_x (ax+b) P_X(x)$$

$$= \underbrace{\sum_{\forall x} ax P_X(x)}_{aE(X)} + \underbrace{\sum_{\forall x} b P_X(x)}_1$$

$$= a \underbrace{\sum_{\forall x} x P_X(x)}_{E(X)} + b \underbrace{\sum_{\forall x} P_X(x)}_1 = aE(X) + b$$

$E(X)$ properties

Example

For the following PMF, if $Y = 100 X$. Find $E(Y)$.

If $W = Y + 500$. Find $E(W)$.

x	0	1	2	3
$P_X(x)$	0.6	0.2	0.15	0.05

$y = 100x$	0	100	200	300
$P_Y(y)$	0.6	0.2	0.15	0.05

$$E(aX + b) = aE(X) + b$$

$$E(X) = 0.65$$

$$E(Y) = \sum y P_Y(y) = 0 * 0.6 + 100 * 0.2 + 200 * 0.15 + 300 * 0.05 = 65$$

$$\hookrightarrow E(Y) = E(100X) = 100E(X) = 100 * 0.65 = 65$$

$$E(W) = E(Y + 500) = E(100X + 500) = 100E(X) + 500 = 100 * 0.65 + 500 = 565$$

Example

For the given PMF $Y = X^2$.

Find μ_Y .

x	1	2	3	4
$P_X(x)$	0.8	0.16	0.032	0.008

$$E[g(X)] = \sum_x g(x) P_X(x)$$

✓ $E(X^2)$ is called the *second moment* of X .

$$E(Y) = E(X^2) = \sum_{\forall x} x^2 P_X(x)$$

$$= (1)^2 * 0.8 + (2)^2 * 0.16 + (3)^2 * 0.032 + (4)^2 * 0.008$$

Exercise

x	-2	-1	0	1	2
$P_X(x)$	0.05	0.2	0.5	0.2	0.05

$Y = X^2$ find pmf of Y & $E(Y)$

$E(X)$: 1st moment
 $E(X^2)$: 2nd moment
;
 $E(X^n)$: nth moment

} "Moment generating function"

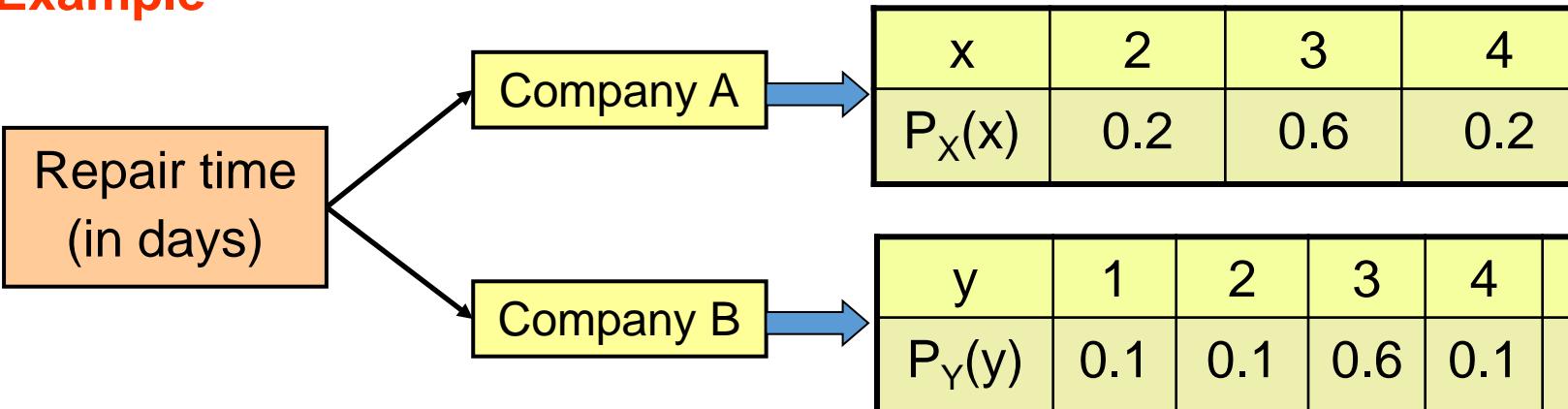
Statistics
Variance
2nd moment - (1st moment)²

2.2 Discrete random variables

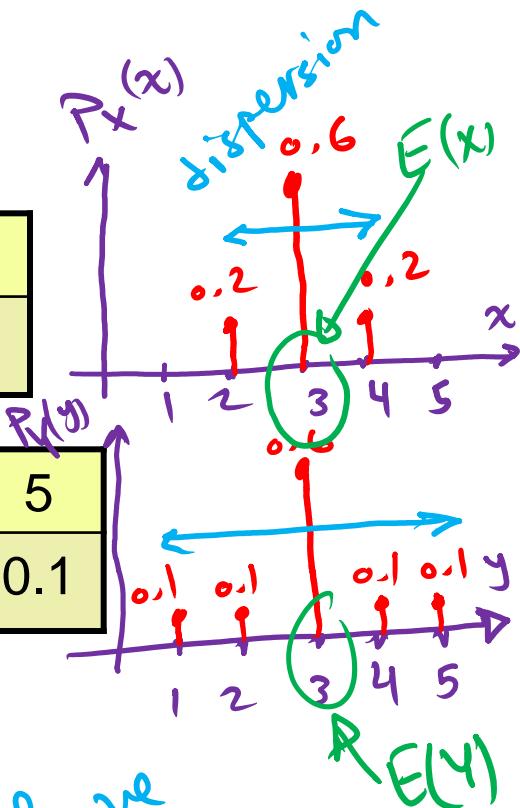
2.2.4 Variance and Standard deviation

Introduction to variance ($\text{Var}(X)$)

Example



Q: Which is better?



Definition

$$\text{Var}(X) = E(X - \mu_x)^2$$

$$= \sum_{\forall x} (x - \mu_x)^2 P_X(x)$$

to avoid adding true & ve dispersion

$E(X) = 3,$

$\text{Var}(X) = 0.4,$

$E(Y) = 3$

$\text{Var}(Y) = 1.0$

Variance is a measure of dispersion

$$\text{Var}(X) = E(X - \mu_x)^2 = \sum_{\forall x} (x - \mu_x)^2 P_X(x) = (2 - 3)^2 * 0.2 + (3 - 3)^2 * 0.6 + (4 - 3)^2 * 0.2 \\ = 0.4$$

More about variance

$$E(\text{Const.}) = \text{The same constant.}$$

$$\text{Var}(X) = E(X - \mu_X)^2$$

X : Random variable
 M_X : Constant

Computational formula

$$\text{Var}(X) = E(X^2) - (\mu_X)^2 = \text{2nd moment} - (\text{1st moment})^2$$

$$\text{Var}(X) = E(X - M_X)^2 = \sum_{\forall x} (x - M_X)^2 P_X(x)$$

$$= \sum_{\forall x} (x^2 - 2M_X x + M_X^2) P_X(x)$$

$$= \underbrace{\sum_{\forall x} x^2 P_X(x)}_{E(X^2)} - 2M_X \underbrace{\sum_{\forall x} x P_X(x)}_{E(X) = \mu_X} + M_X^2 \underbrace{\sum_{\forall x} P_X(x)}_1$$

$$E(X^2) - 2M_X^2 + M_X^2$$

$$= E(X^2) - M_X^2$$

More about variance

$$\text{Var}(X) = E(X - \mu_X)^2$$

$E(\text{const.}) = \text{The same const.}$

$$\text{Var}(\text{const}) = 0$$

"No dispersion"

Computational formula

$$\text{Var}(X) = \underline{\underline{E(X^2) - (\mu_X)^2}}$$

General property

$$\text{Var}(aX + b) = a^2 \text{Var}(X) \quad \#$$

$$\begin{aligned}\text{Var}(aX + b) &= E(aX + b)^2 - (E(aX + b))^2 \\&= E[a^2 X^2 + 2abX + b^2] - [a\mu_X + b]^2 \\&= \cancel{a^2 E(X^2)} + \cancel{2ab\mu_X + b^2} - \cancel{a^2 \mu_X^2} - \cancel{2ab\mu_X} - \cancel{b^2} \\&= a^2 [E(X^2) - \mu_X^2] = a^2 \text{Var}(X)\end{aligned}$$

More about variance

$$\text{Var}(X) = E(X - \mu_X)^2$$

Computational formula

$$\text{Var}(X) = E(X^2) - (\mu_X)^2$$

General property

$$\text{Var}(aX + b) = a^2 \text{Var}(X)$$

Standard deviation

$$\sigma_X = \sqrt{\text{Var}(X)}$$

X : m
↳ water level in atoms
unit

$$\begin{aligned}E(X) &: m \\ \text{Var}(X) &: m^2 \\ \widetilde{\sigma_X} = \sqrt{\text{Var}(X)} &: m^{\frac{1}{2}}\end{aligned}$$

Example

For the given PMF,

Find $\text{Var}(X)$, $\text{Var}(2X)$

x	0	1	2	3
$P_X(x)$	0.6	0.2	0.15	0.05

$$E(X) = 0.65$$

$$\begin{aligned} E(X^2) &= (0)^2 * 0.6 + (1)^2 * 0.2 + (2)^2 * 0.15 + (3)^2 * 0.05 \\ &= \square \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &= E(X^2) - (E(X))^2 \\ &= \square - (0.65)^2 = \checkmark \end{aligned}$$

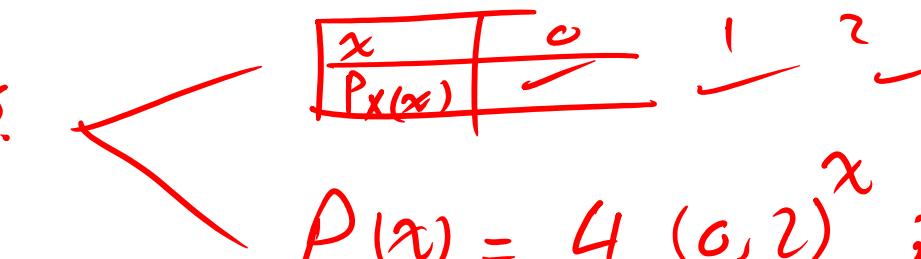
Example

Two new product designs are to be compared on the basis of revenue potential on the long run. Marketing estimates that the revenue from design A can be predicted accurately to be 3 M \$. The revenue potential of design B is more difficult to assess. Marketing concludes that there is a probability of 0.3 that the revenue from design B will be 7 M \$. However, there is a probability of 0.7 that the revenue will be 2 M \$. Calculate the mean and standard deviation of the revenue from each design.

Exercise

In previous
lectures

→ General PMFs.



$$P_X(x) = 4(0.2)^x ; x = 1, 2, \dots$$

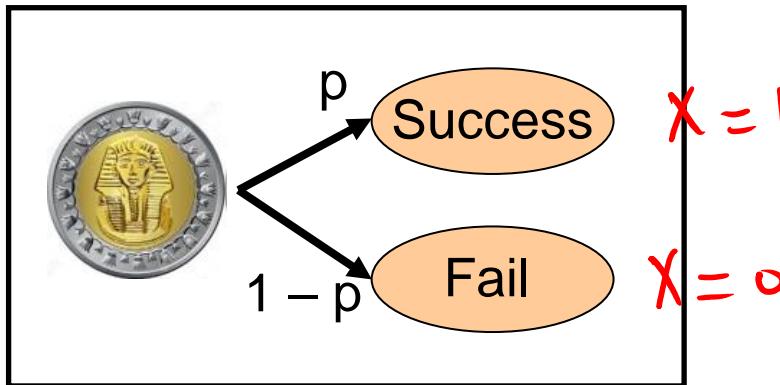
2.3 Some important discrete PMFs

2.3.1 Bernoulli PMF

Bernoulli Trial

A random experiment with two possible outcomes that may be success or failure.

Bernoulli trial



x	0	1
$P_X(x)$	$1 - P$	P

$$P_X(x) = \begin{cases} P, X = 1 \\ 1 - P, X = 0 \end{cases}$$

$$E(X) = \sum_{x \in \{0, 1\}} x P_X(x) = 0(1 - P) + 1(P) = P$$

$$E(X^2) = (0)^2(1 - P) + (1)^2P = P$$

$$V(X) = E(X^2) - (E(X))^2 = P - P^2 = P(1 - P)$$

$$E(X) = p,$$

$$\text{Var}(X) = p(1 - p)$$

success: Event of interest

Random experiment could have more than 2 outcomes
& still Bernoulli

Example

Rolling a die, where a six is "success" and everything else a "failure". Find the probability of success.

$$X \stackrel{\text{D has}}{\sim} \text{Bernoulli} \left(P_{\text{success}} \right)$$

6 appears

$$\frac{1}{6}$$