

# Probability Distribution



# Frequency distribution

## Quantitative data

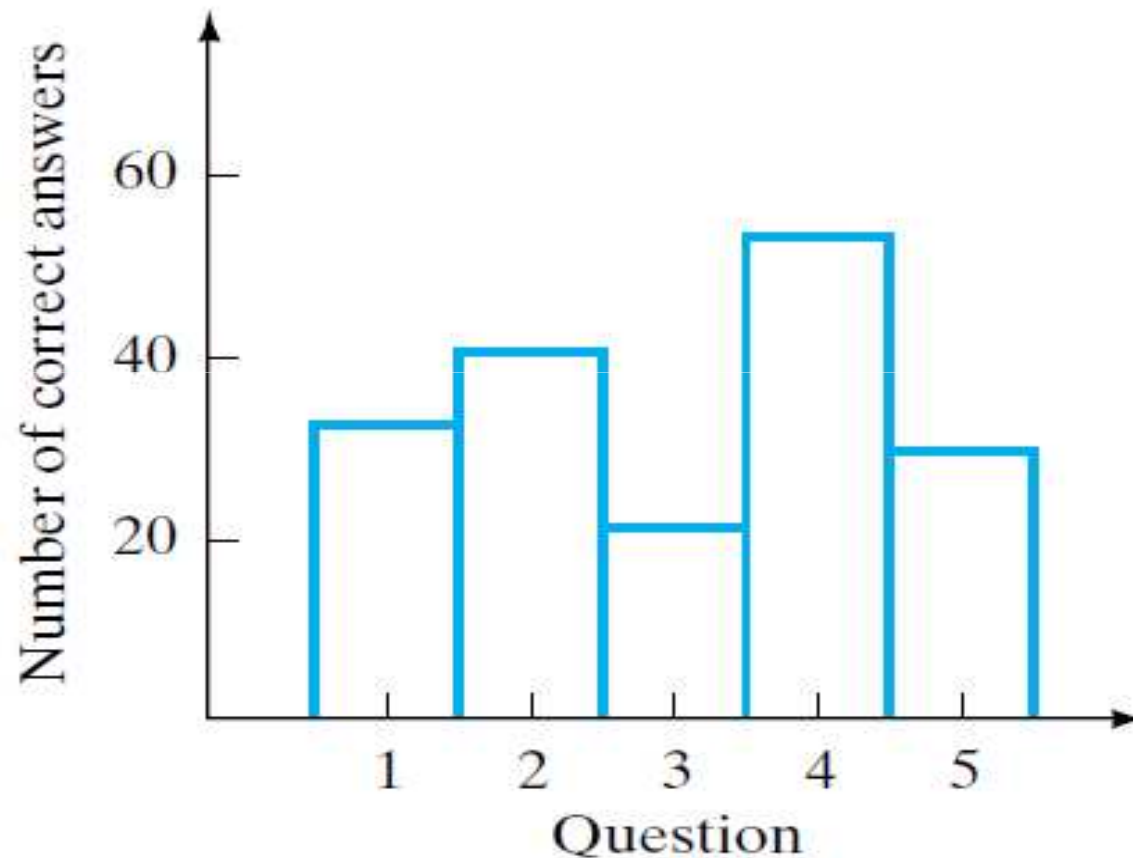
Score	Number of Students Making Score
0–20	0
21–40	18
41–60	36
61–80	83
81–100	110
101–120	121
121–140	73
141–160	<u>16</u>
Total	457

## Qualitative data

Major	Number of Students
Science	429
Arts	132
Languages	41
Social sciences	631
Engineering	<u>344</u>
Total	1577



# Histogram – no. of correct answers for 5 questions in an exam



# Relative frequency

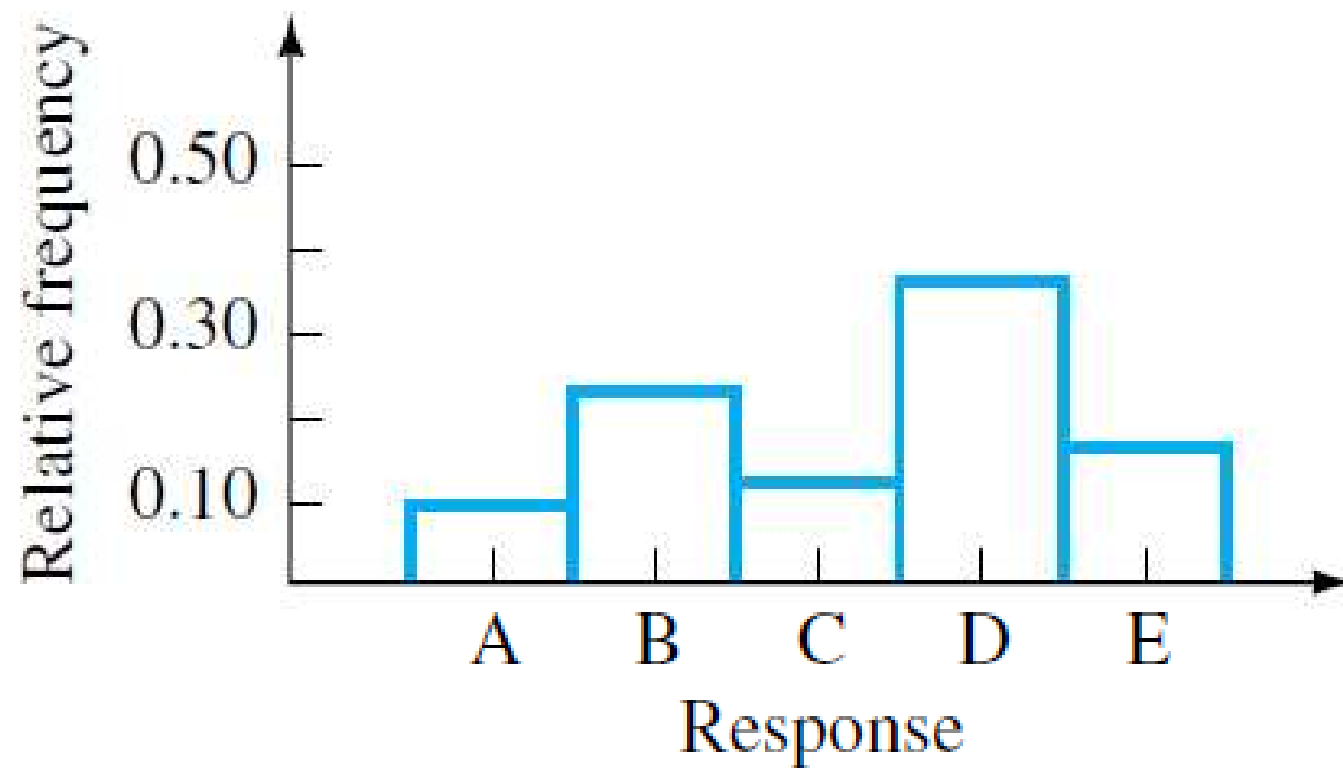
- Relative frequency counts the fractional part of the data that belong to a category



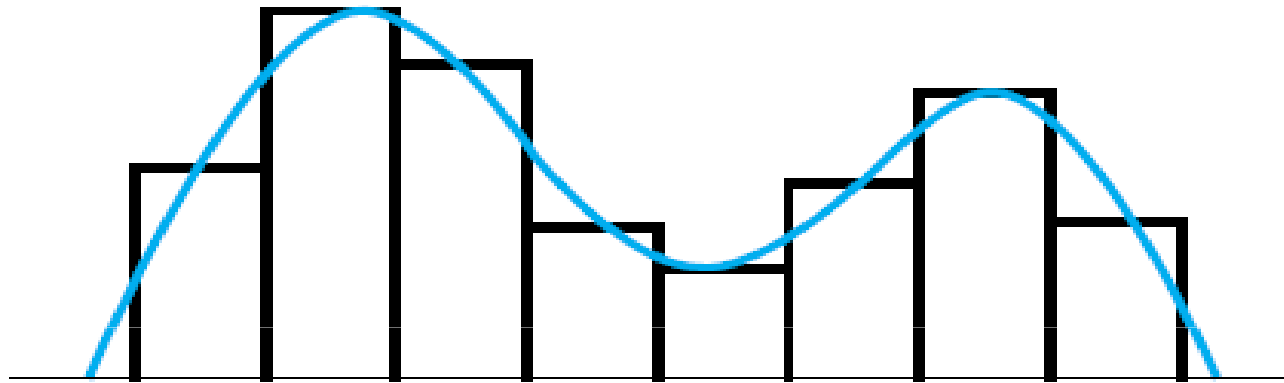
An exam has five possible grades:  
A, B, C, D, and E

Response	Frequency	Relative frequency
A	6	$6/60=0.1$
B	14	$14/60=0.23$
C	8	$8/60=0.13$
D	22	$22/60=0.37$
E	10	$10/60=0.17$
Total = 60		Total = 1



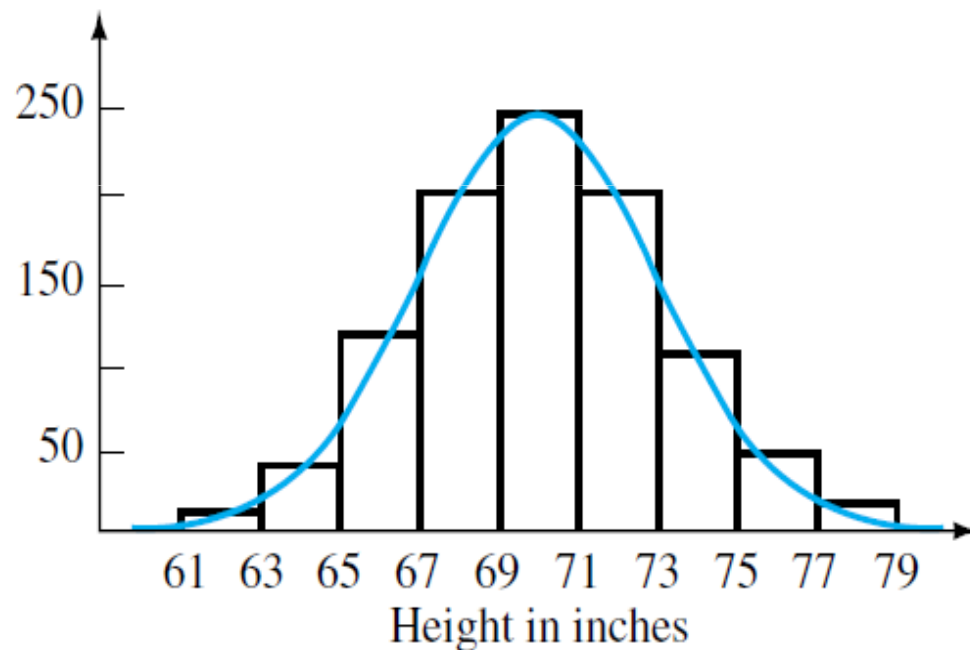


# When dealing with continuous data...



- A smooth curve conveys the impression of continuous data better than a histogram.
- Sketch a smooth curve based on a histogram by drawing it through the midpoints at the top of the bars

# Continuous data – example

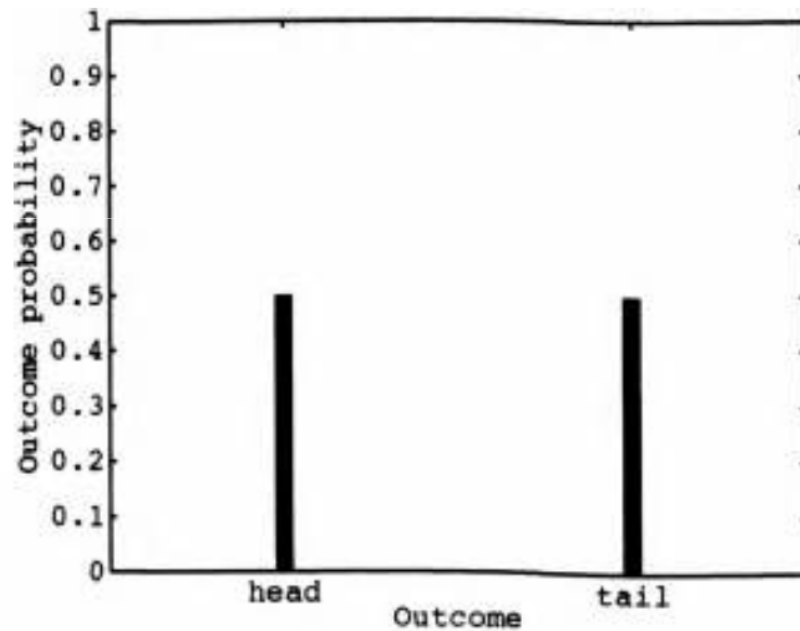


Height (Inches)	Frequency
61–62.9	10
63–64.9	51
65–66.9	115
67–68.9	200
69–70.9	240
71–72.9	195
73–74.9	104
75–76.9	42
77–78.9	15



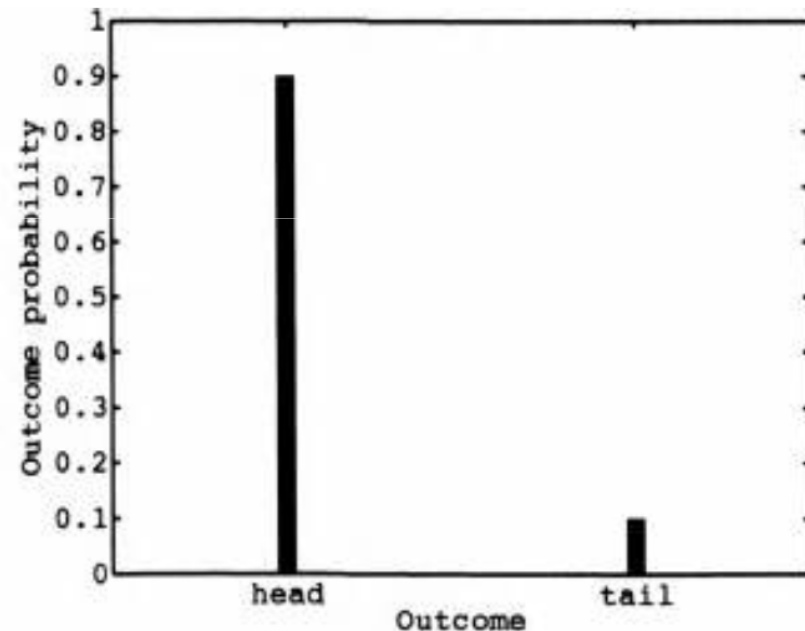
# Probability distributions of two coins

$p(\text{head})=0.5$



Unbiased coin

$p(\text{head})=0.9$



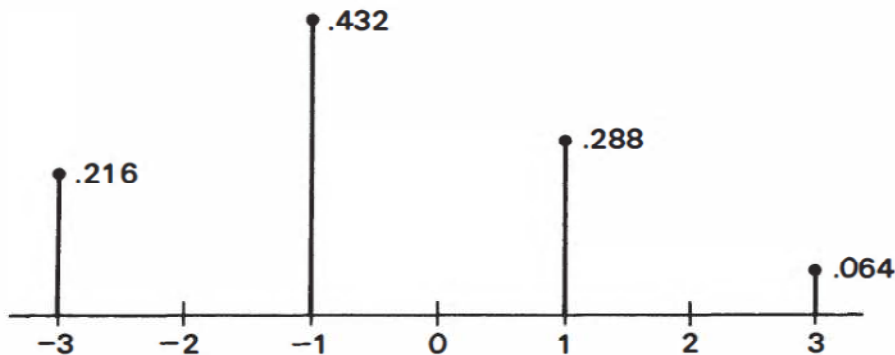
Biased coin



## Example - RV

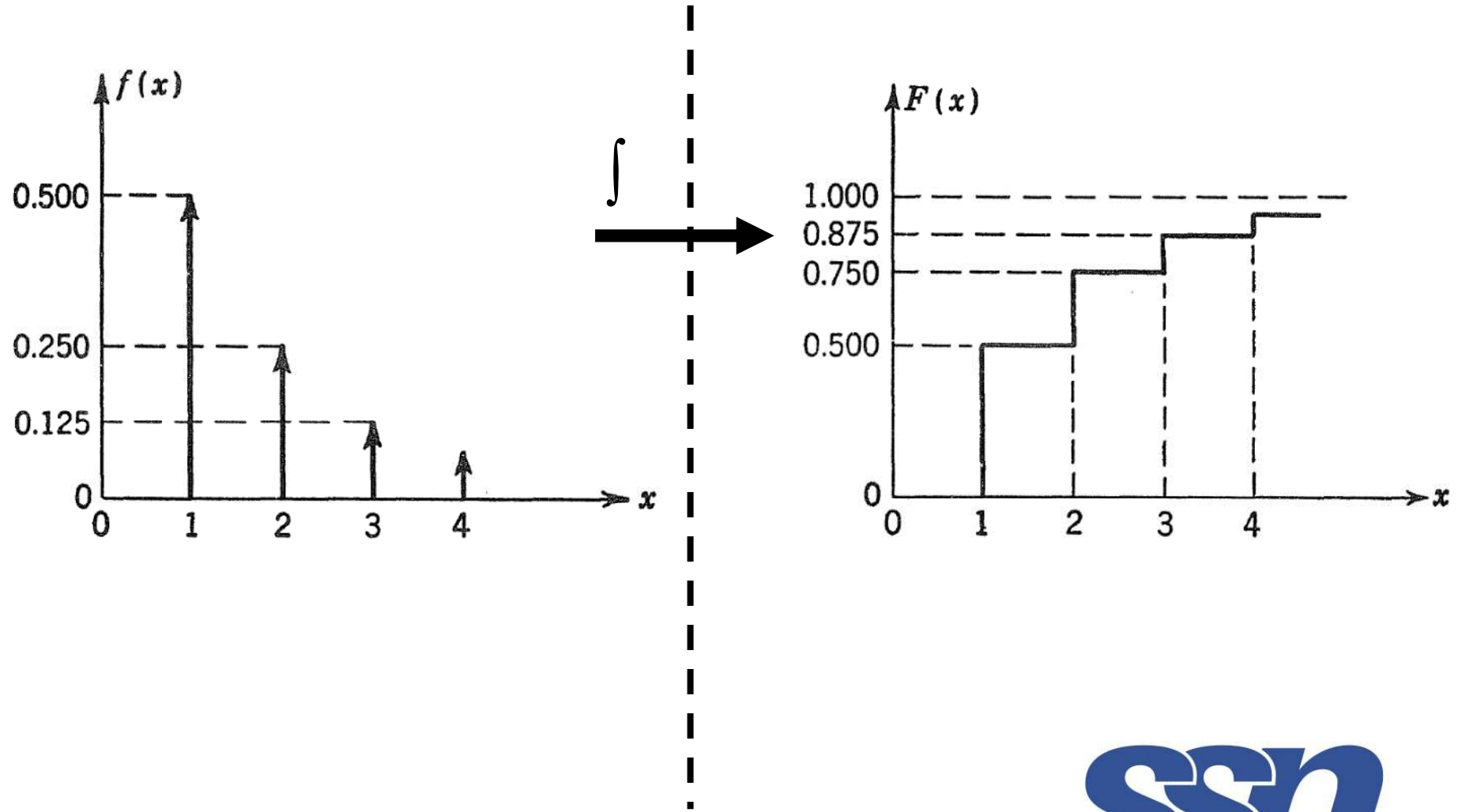
A coin is flipped 3 times  
3 heads earn 3 points  
2 heads earn 1 point  
3 tails earn -3 points  
2 tails earn -1 point

If  $p = 0.4$ ,



$\omega$	$X(\omega)$	$P\{\omega\}$
HHH	3	$p^3$
HHT	1	$p^2(1 - p)$
HTH	1	$p^2(1 - p)$
THH	1	$p^2(1 - p)$
HTT	-1	$p(1 - p)^2$
THT	-1	$p(1 - p)^2$
TTH	-1	$p(1 - p)^2$
TTT	-3	$(1 - p)^3$

# Probability distribution function – another look



# Example

- 15 cards with 5 colored red, 3 colored black, and 7 colored green
- The cards are shuffled, and 2 cards selected
  - No points if both cards different colors
  - Five points if both cards are green
  - 10 points are if both cards are red
  - 5 points are if both cards are black



# Random variable

<b>Outcome</b>	<b>Value of <math>X</math> (points)</b>
Red and black	0
Red and green	0
Red and red	10
Black and black	15
Black and green	0
Green and green	5

The probability of drawing two greens is  $\frac{C(7, 2)}{C(15, 2)} = 0.200$

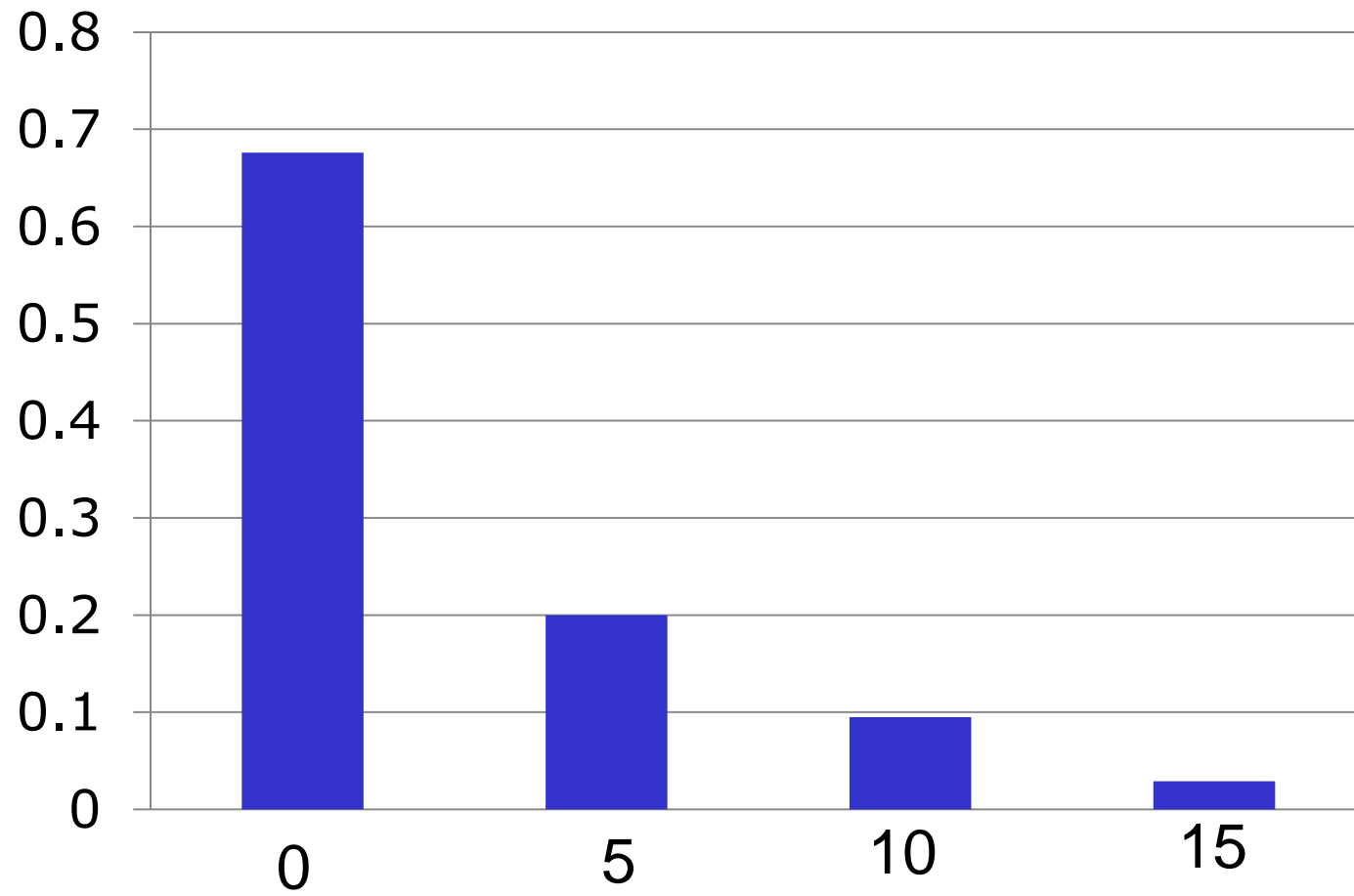
of two reds  $\frac{C(5, 2)}{C(15, 2)} = 0.095$

of two blacks  $\frac{C(3, 2)}{C(15, 2)} = 0.029$

<i><b>X</b></i>	<i><b>P(X)</b></i>
0	0.676
5	0.200
10	0.095
15	0.029

→  $1 - (0.2 + 0.095 + 0.029)$





**ssn**

## One more example

- Two people are selected from a group of five men and four women
- A random variable  $X$  is the number of women selected
- What is the range of RV?
- $X = \{0, 1, 2\}$
- Range is 0 to 2





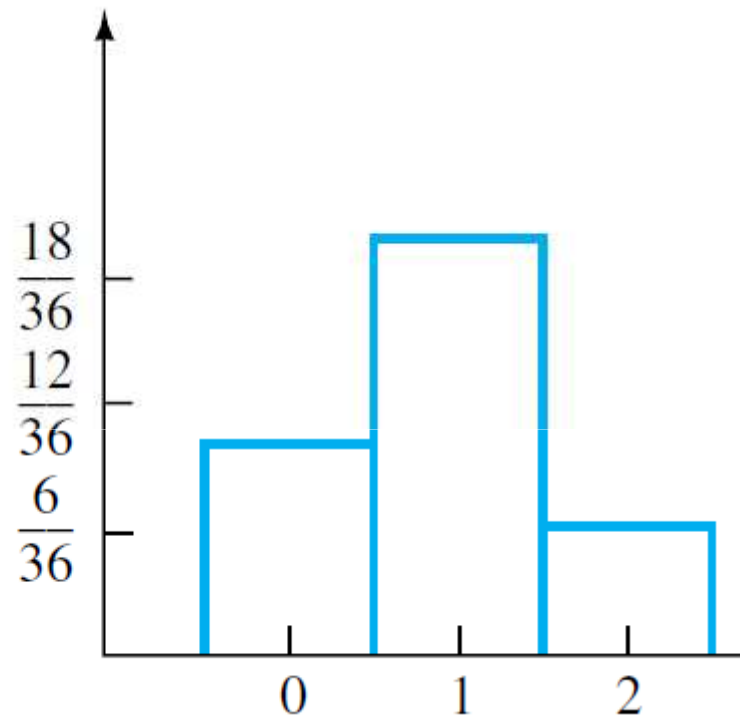
Outcome	X
Three men	0
One man and one woman	1
Two women	2

Outcome	X	Possible ways
Two men	0	$C(5,2) = 5!/(3! \times 2!) = 10$
One man and one woman	1	$C(5,1) \cdot C(4,1) = 5!/(4! \times 1!) \times 4!/(3! \times 1!) = 20$
Two women	2	$C(4,2) = 4!/(2! \times 2!) = 6$

Two people from nine:  $C(9,2) = 9!/(7! \times 2!) = 36$

## Probability for $\{0, 1, 2\}$

$X$	$P(X)$
0	$\frac{10}{36}$
1	$\frac{20}{36}$
2	$\frac{6}{36}$



# Average

- N numbers are given
- Add all the numbers and divide by N

# Expectation

- Kind of average for random variable
- Dealing with probabilities



# Difference between average and expectation - mathematically

## **Average of N numbers:**

- Weighting factor is  $(1/N)$

## **Expectation of random variable with N outcomes**

- Weighting factor is probability



# Average

$$\begin{aligned}\text{Average} &= (x_1 + x_2 + \dots + x_n)/n \\ &= (1/n) \cdot x_1 + (1/n) \cdot x_2 + \dots + (1/n) \cdot x_n\end{aligned}$$

- All the data are weighed equally with  $(1/n)$
- $(1/n)$  can also be called as weighing function

# Expectation

- Let  $X$  represent the outcome of a roll of an unbiased six-sided die
- Possible values for  $X$  are 1, 2, 3, 4, 5, & 6
- Each having the probability of occurrence of  $1/6$

## Expectation

$$E(X) = (1/6)*1 + (1/6)*2 + (1/6)*3 + (1/6)*4 + (1/6)*5 + (1/6)*6$$

$$E(X) = 3.5$$



# Average

- Suppose that in a sequence of ten rolls of the die, if the outcomes are 5, 2, 6, 2, 2, 1, 2, 3, 6, 1,
- Average (arithmetic mean) of the results is given by  
$$(5+2+6+2+2+1+2+3+6+1)/10=3.0$$



# Average and expectation

- Expectation: Outcome is not sure i.e. in the form of probability
- Average: Deterministic data values

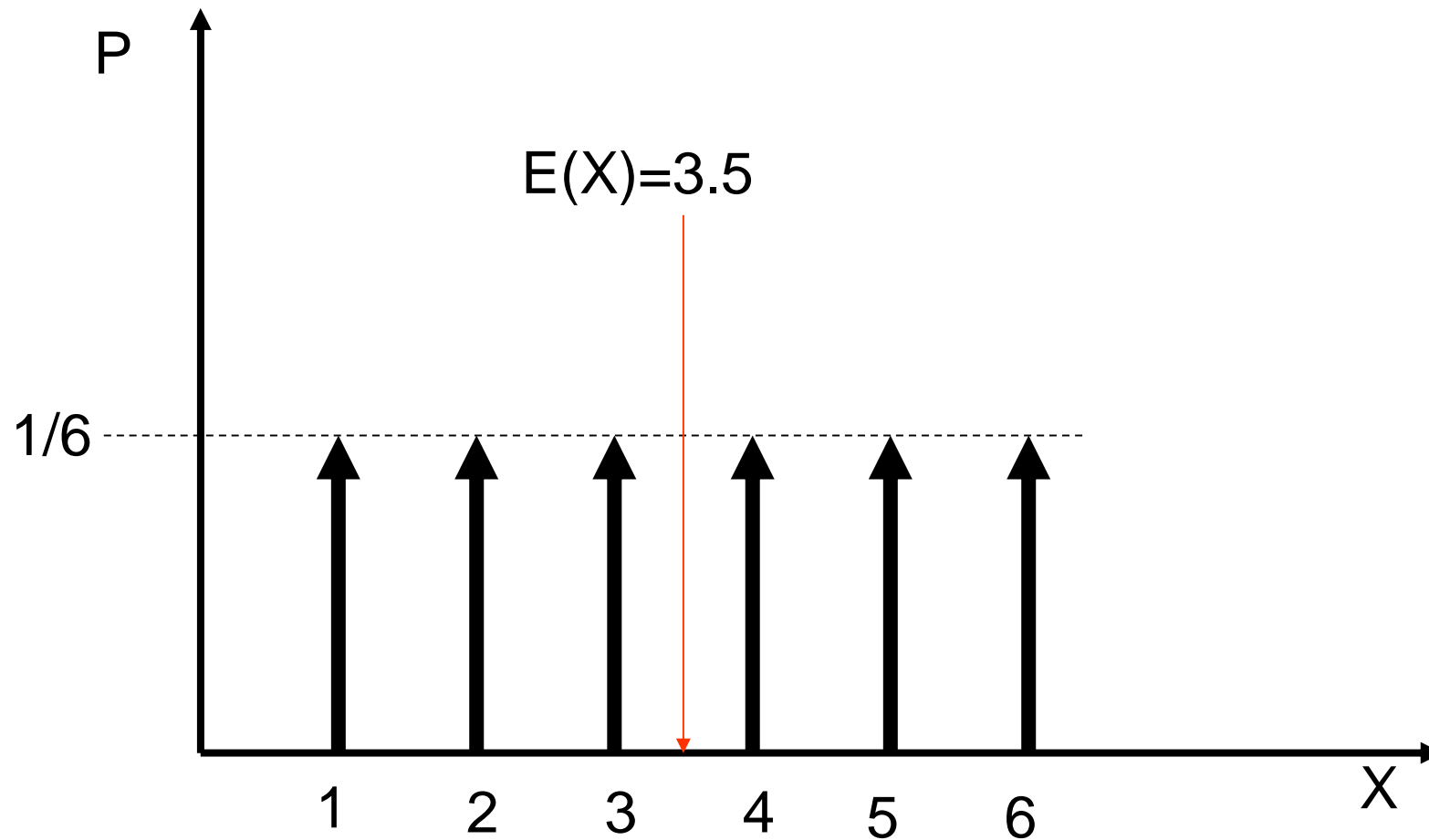
## **Previous E.g.**

- Average value is 3.0, with the distance of 0.5 from the expectation value of 3.5
- Roll the die N times, N is very large
- Then the average will converge to the expected value,

$$\text{Average (X)} \rightarrow \text{Expected(X)}$$



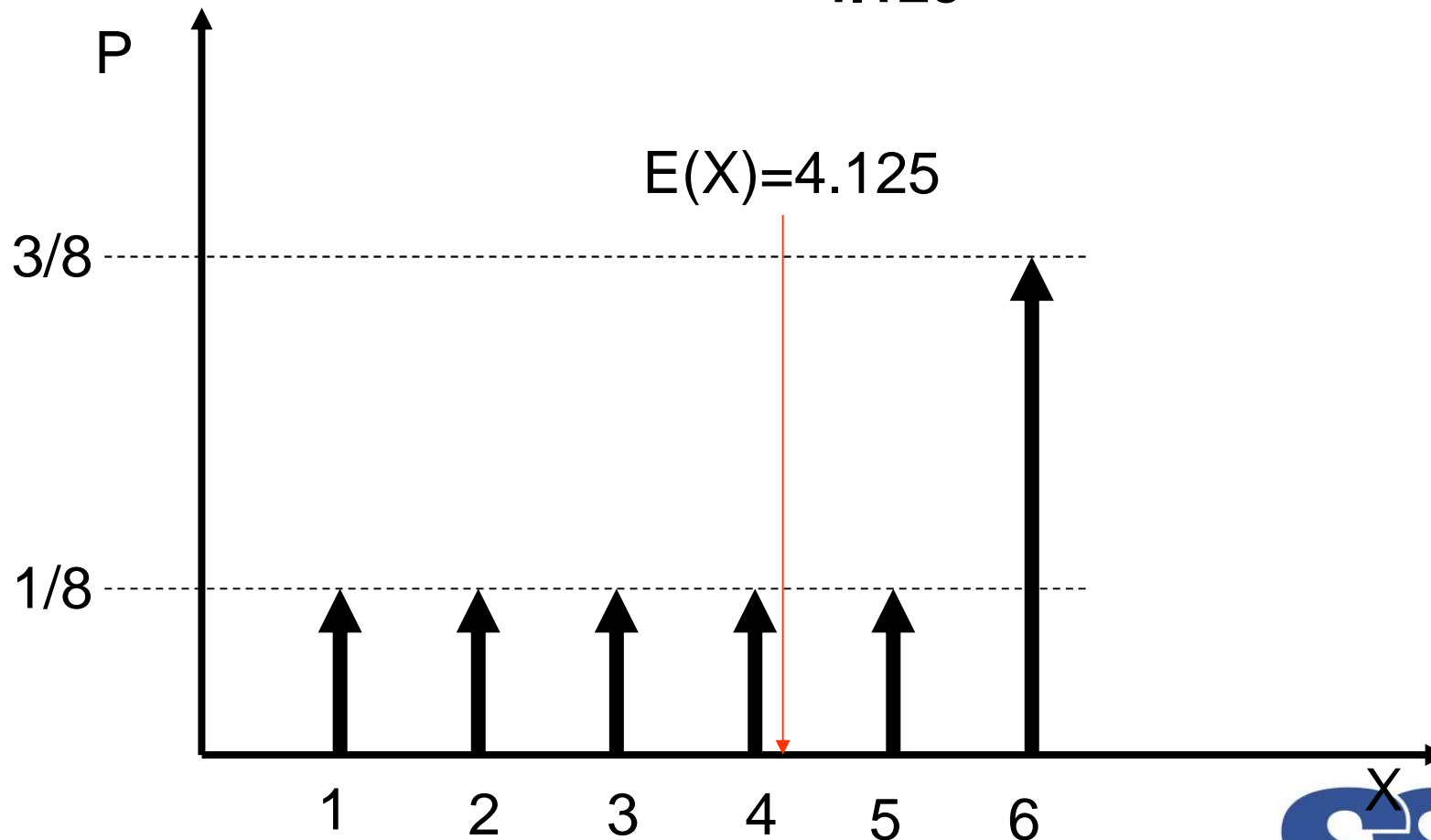
## Expectation – rolling fair dice



## Expectation – rolling biased dice

### Expectation

$$E(X) = (1/8)*1 + (1/8)*2 + (1/8)*3 + (1/8)*4 + (1/8)*5 + (3/8)*6 = 4.125$$



# Expectation

- Consider a random variable

- **Probability function**

$[x_1, x_2, \dots, x_n]$

$[p_1, p_2, \dots, p_n]$

- Average of  $X$

$$\overline{X} = \sum_{k=1}^n p_k x_k$$

## Expectation - example

$x_i$	2	5	9	24
$p_i$	0.4	0.2	0.3	0.1

$$\begin{aligned} E(X) &= 0.4 \times 2 + 0.2 \times 5 + 0.3 \times 9 + 0.1 \times 24 \\ &= 0.8 + 1.0 + 2.7 + 2.4 \\ &= 6.9 \end{aligned}$$

## One more example

- A tray of fruits contains nine good and three rotten.
- Two fruits selected at random
- What is the expected number of rotten?
- Let the random variable  $X$  be the number of defective
- $X$  can have the value 0, 1, or 2



## Solution

$P(0)$  = probability of no defective (both good)

$$= \frac{C(9, 2)}{C(12, 2)} = \frac{36}{66} = \frac{12}{22}$$

$P(1)$  = probability of one good and one defective

$$= \frac{C(9, 1)C(3, 1)}{C(12, 2)} = \frac{27}{66} = \frac{9}{22}$$

$P(2)$  = probability of two defective

$$= \frac{C(3, 2)}{C(12, 2)} = \frac{3}{66} = \frac{1}{22}$$



$$E(X) = \frac{12}{22} (0) + \frac{9}{22} (1) + \frac{1}{22} (2) = \frac{11}{22} = \frac{1}{2}$$

Interpretation:

If the experiment is repeated many times

- You expect to get no defectives a little less than half the time
- Either one or two the rest of the time





# Standard deviation of a RV

$x_i$	4	7	10	8
$p_i$	0.2	0.2	0.5	0.1

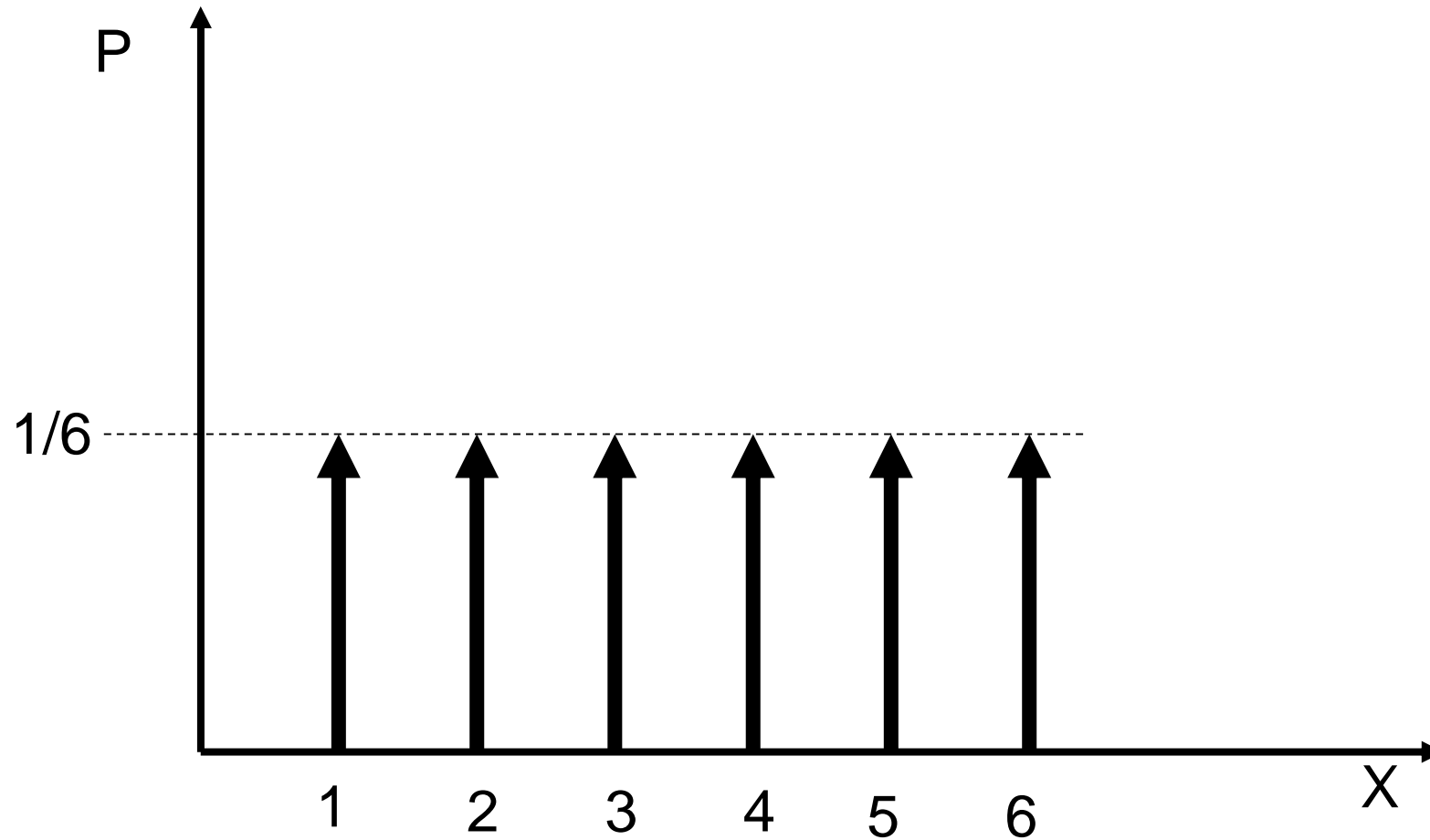
$x_i$	$p_i$	$p_i x_i$	$x_i - \mu$	$(x_i - \mu)^2$	$p_i(x_i - \mu)^2$
4	0.2	0.8	-4	16	3.2
7	0.2	1.4	-1	1	0.2
10	0.5	5.0	2	4	2.0
8	0.1	<u>0.8</u>	0	0	<u>0</u>
$\mu = 8.0$			$\sigma^2(X) = 5.4$		

The variance of  $\sigma^2(X) = 5.4$ . So the standard deviation is

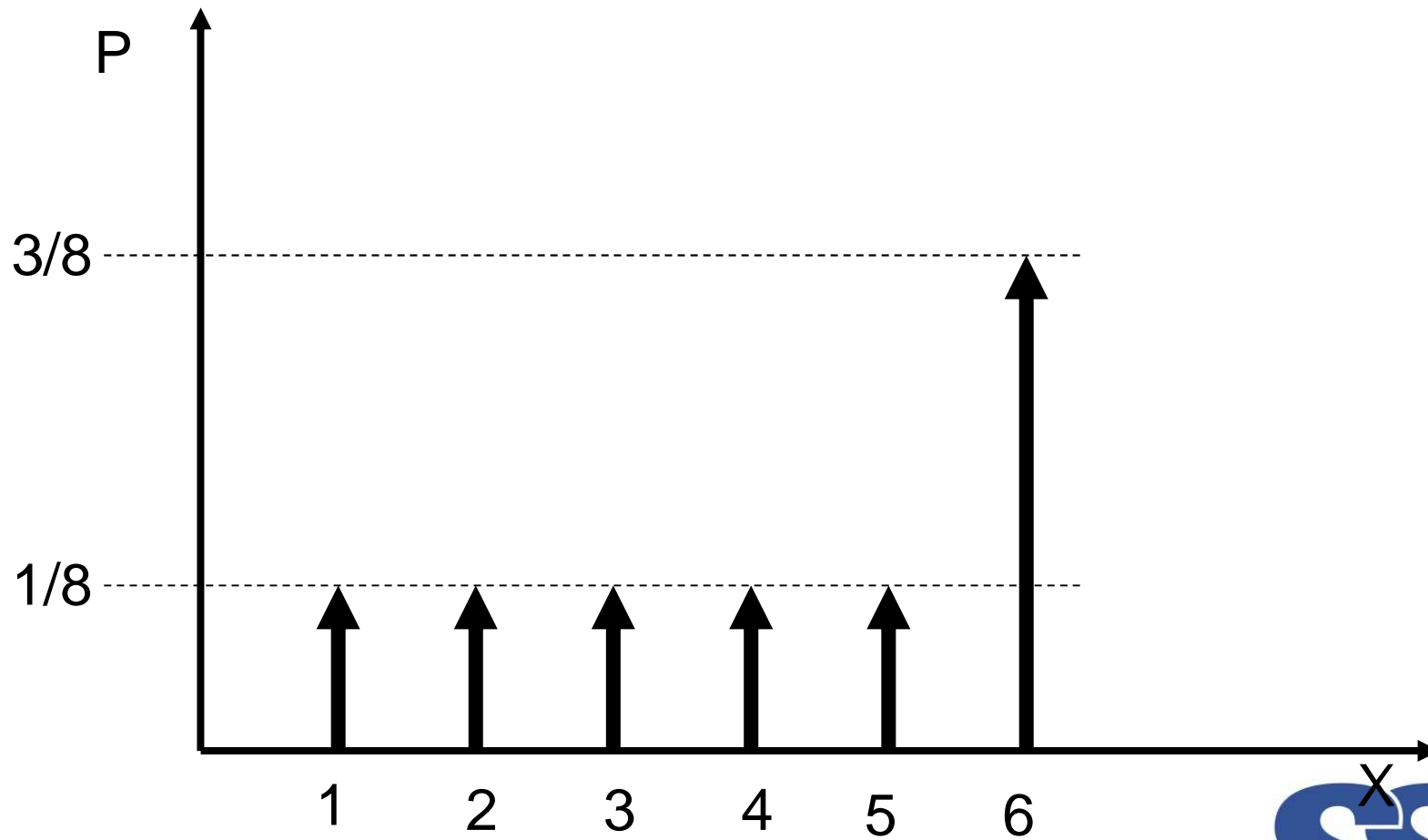
$$\sigma(X) = \sqrt{5.4} = 2.32$$

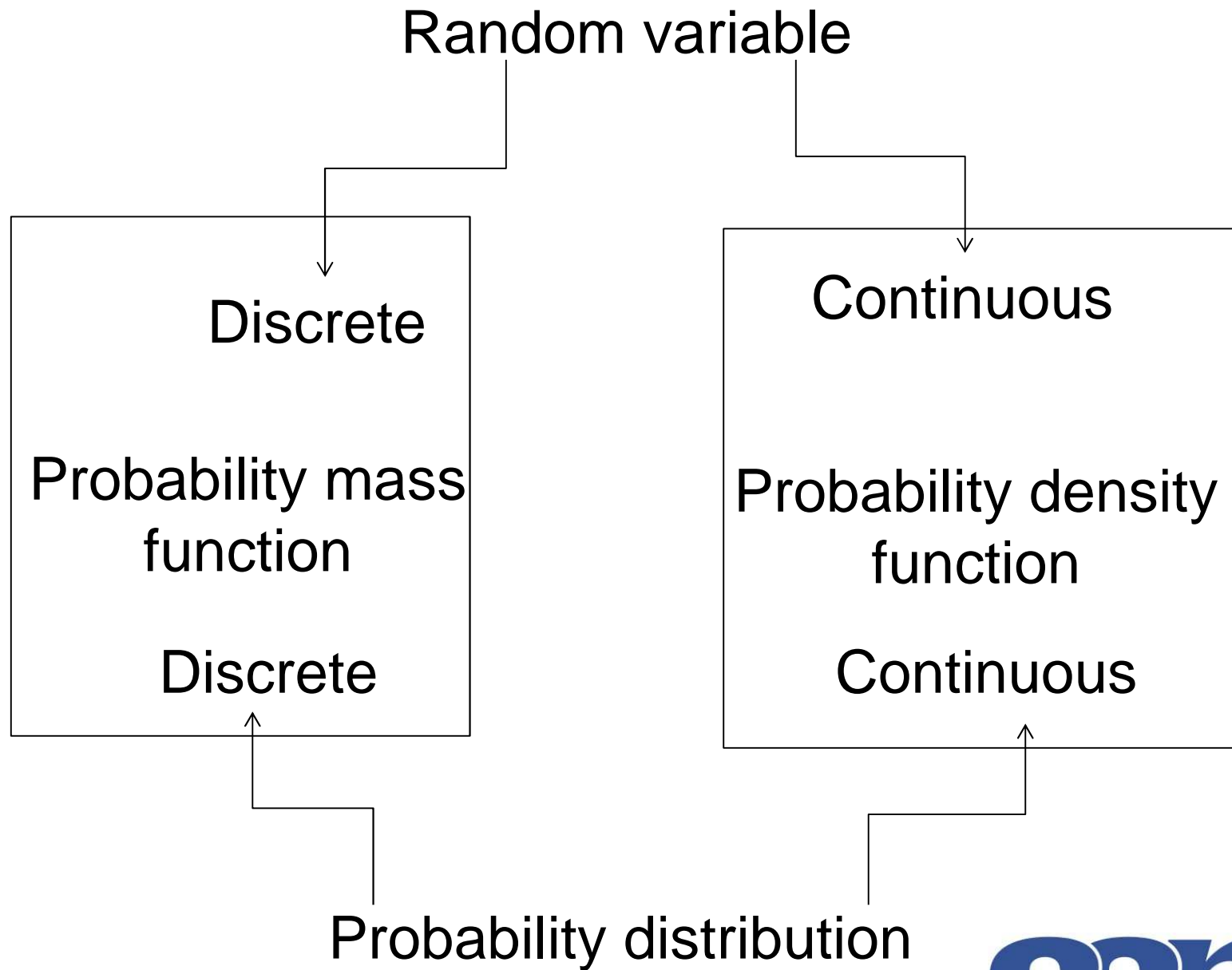


# Uniform distribution

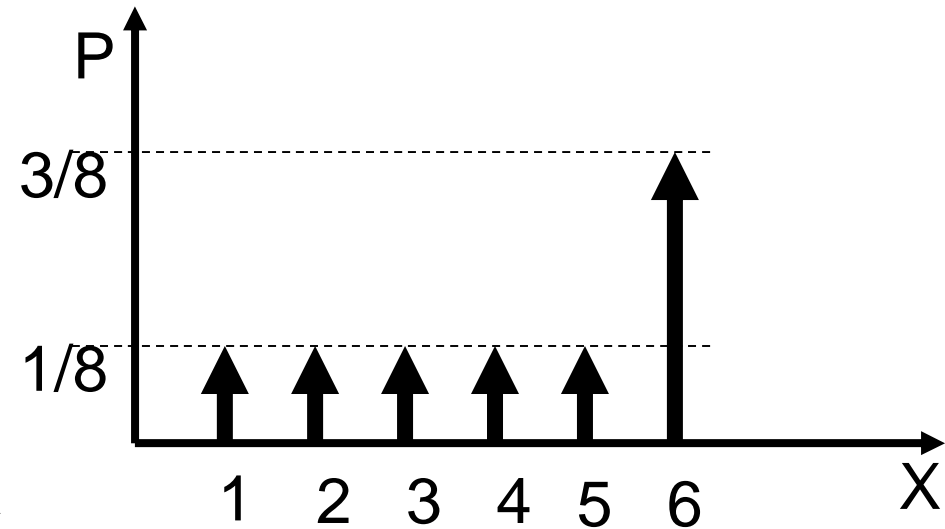
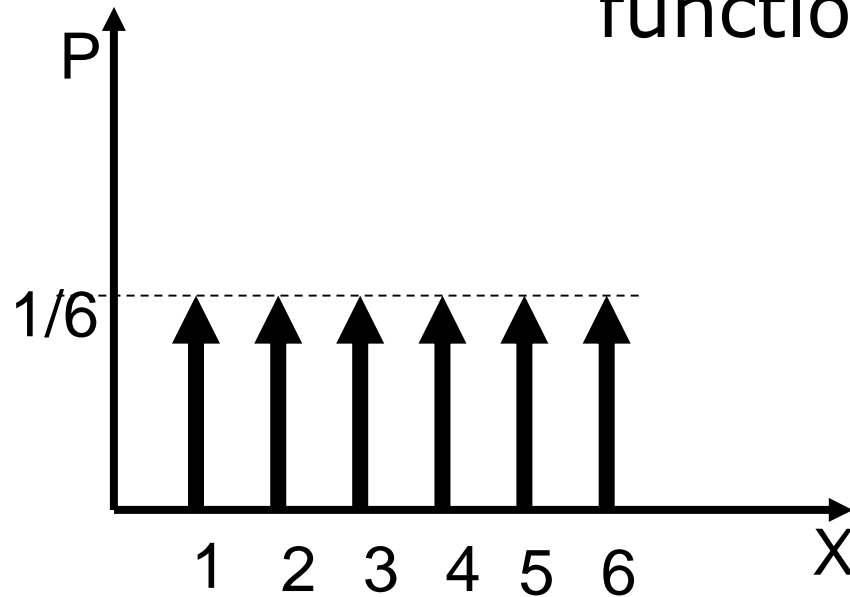


## Non-uniform distribution





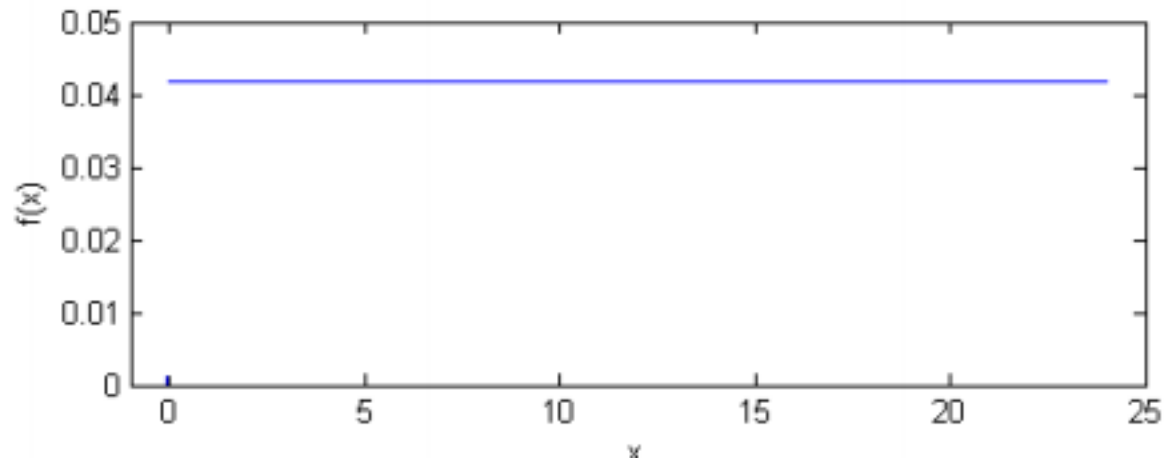
Discrete random variable:  
Probability distribution – Probability Mass  
function - PMF



$$\Sigma p(i) = 1$$

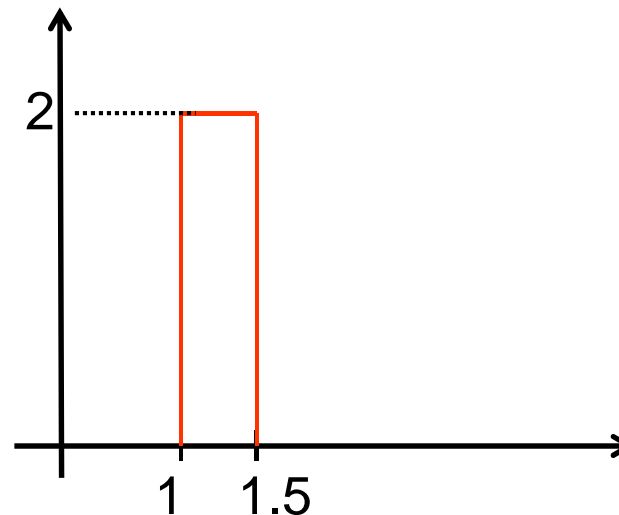
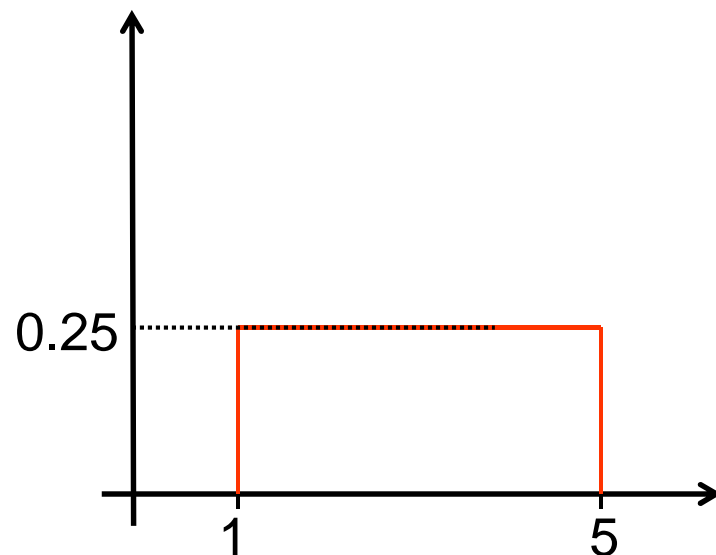
# Continuous random variable: Probability distribution - Probability Density function - PDF

- In PDF – what should not be asked?
- What is the probability @ some value of  $x$ ?
- Ask – what is the probability for  $x=x_1$  to  $x=x_2$  i.e. interval?



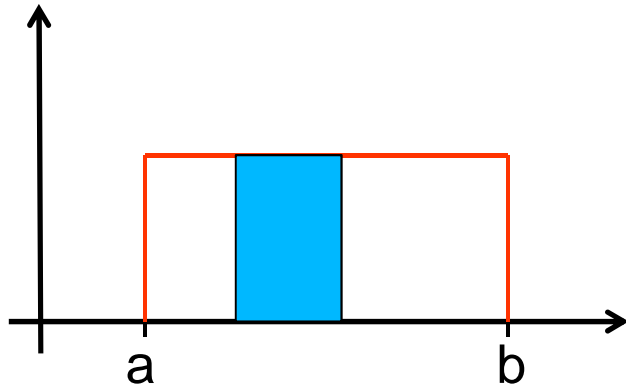
## Some more points on PDF

- PMF values cannot be greater than 1
- But PDF values can be greater than 1
- Area under PDF = 1 (only condition)
- Consider following uniform distributions



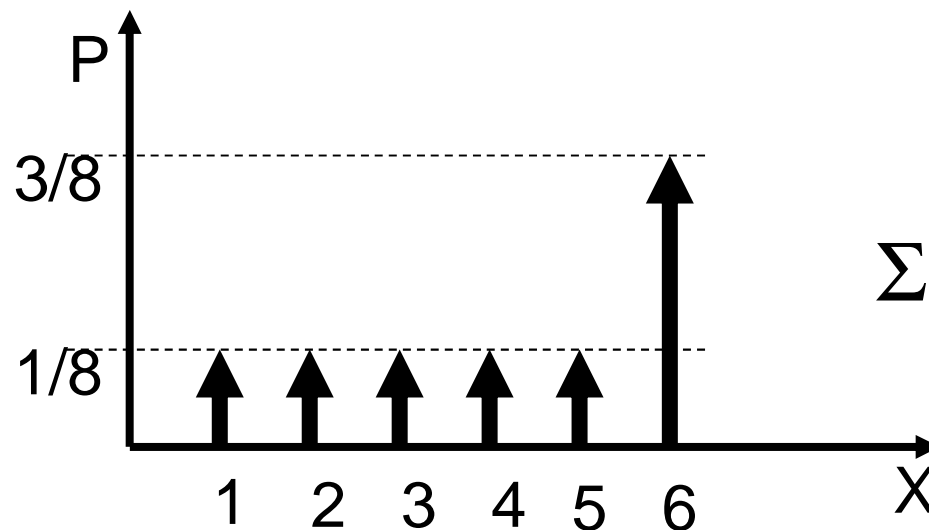
## PDF - Area gives probability

$$p(x_1 < x < x_2) = \int_{x_1}^{x_2} p(x) dx$$



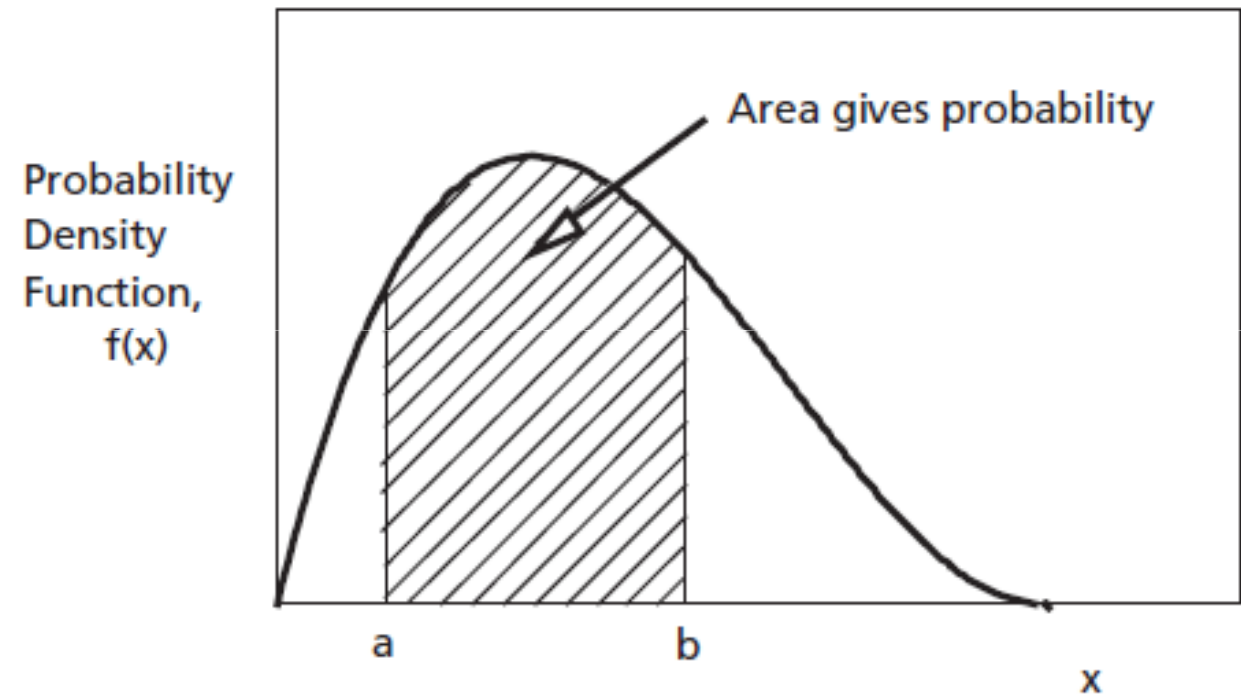
$$\int_a^b p(x) dx = 1$$

PMF – probability values can be directly read

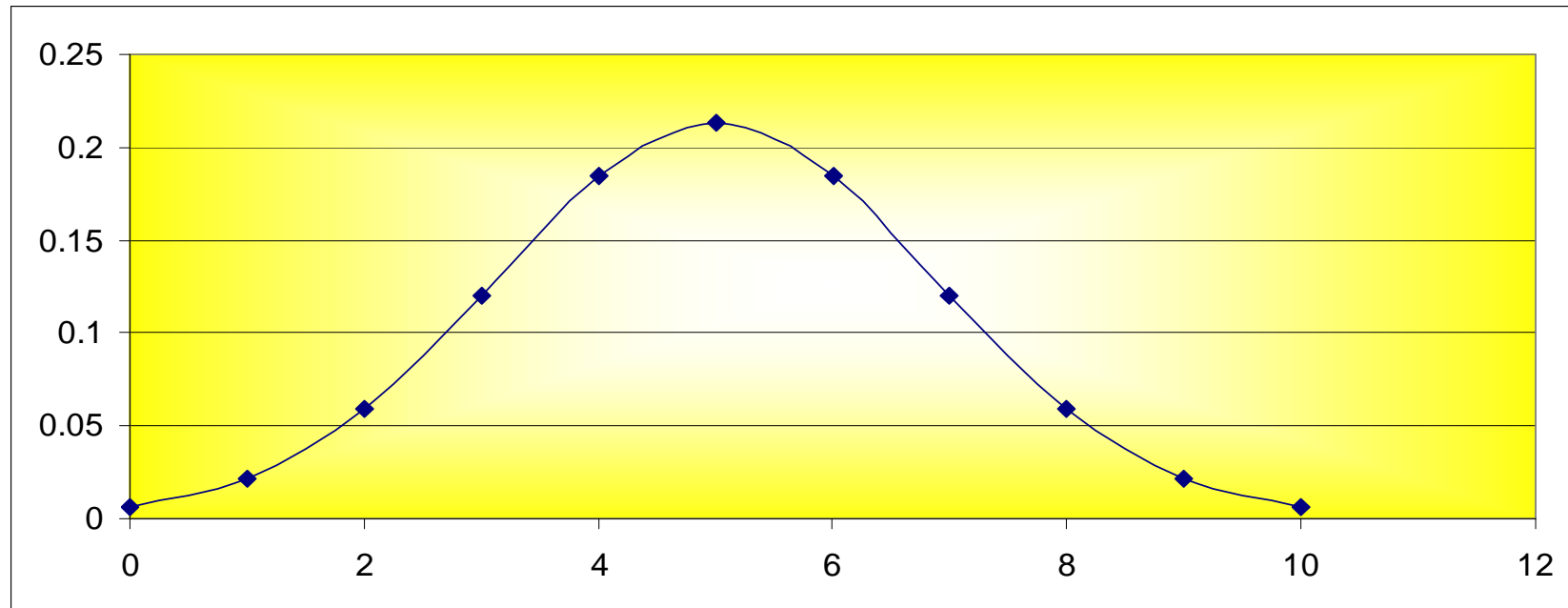


$$\sum p(i) = 1$$





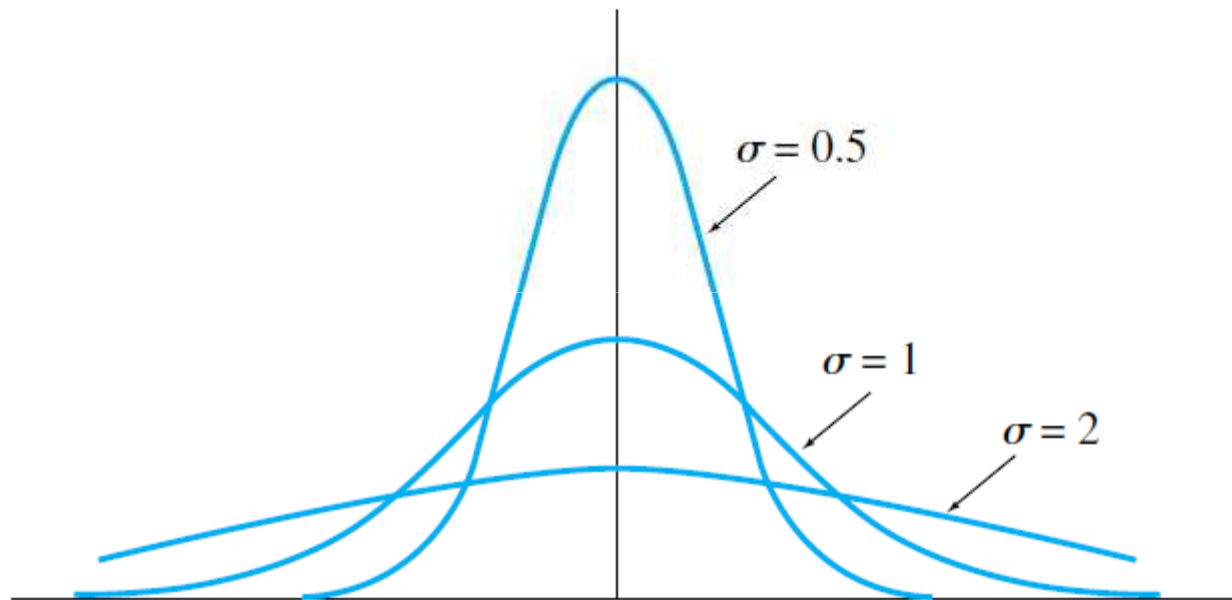
# Gaussian distribution – mean and standard deviation



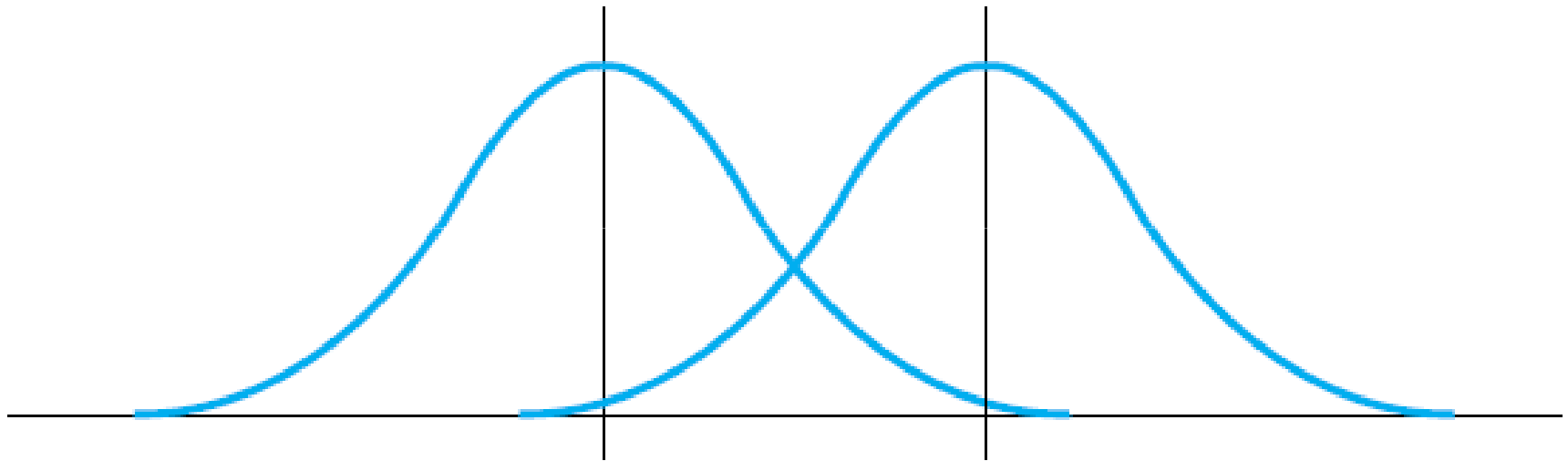
$$\frac{1}{\sqrt{2 \pi \sigma^2}} e^{-\frac{(x - \mu)^2}{2 \sigma^2}}$$

Defined from  
 $-\infty$  to  $+\infty$

# Same mean – different standard deviations



Same standard deviation – different means

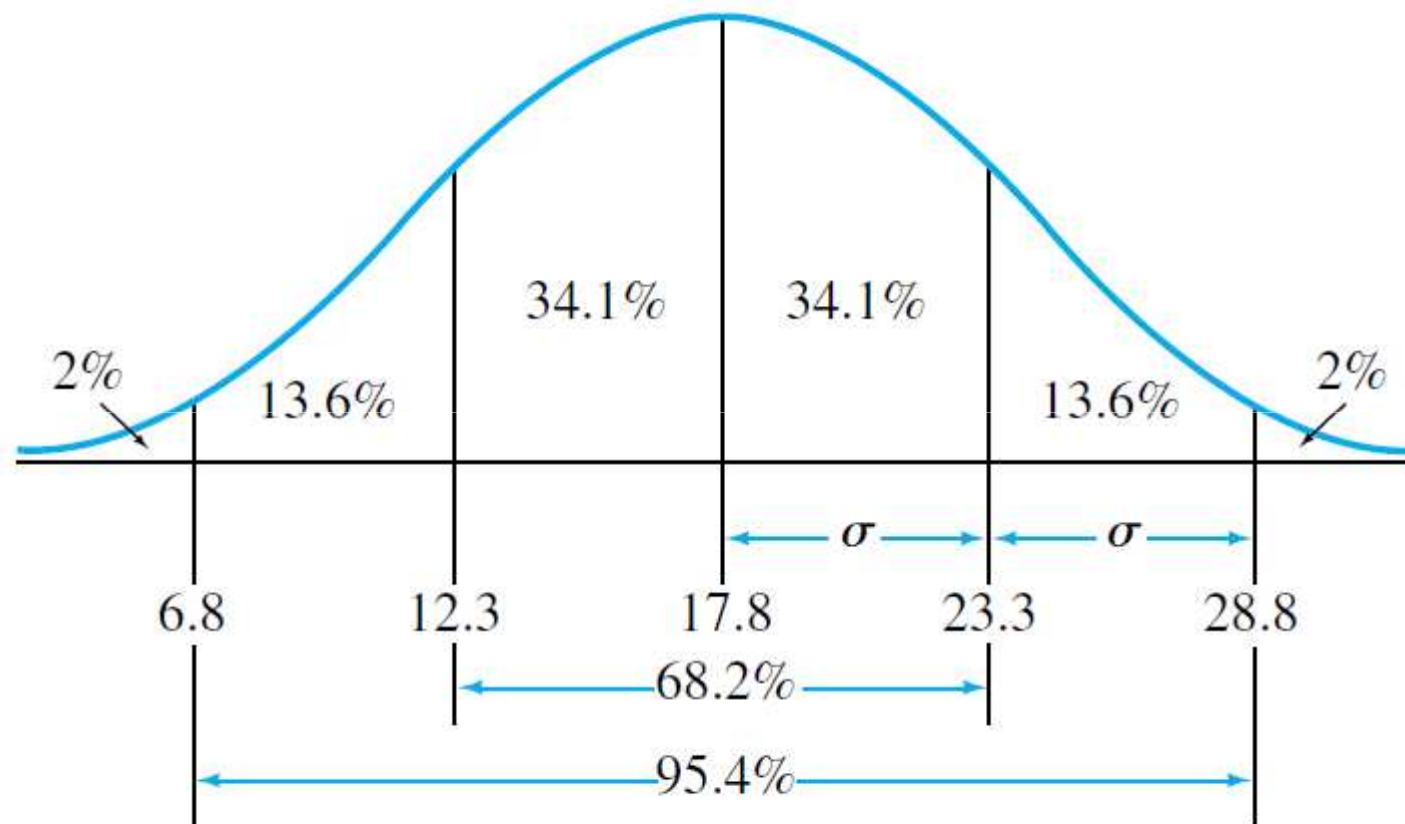


# Standard normal deviation

- Make,  $\sigma = 1$
- Area under the curve becomes 1
- can be used as probability measure



# Area under the normal curve



Normal curve with  $\mu = 17.8$  and  $\sigma = 5.5$

$$\overline{X} = \sum_{k=1}^n p_k x_k$$

*More generic form,*

$$\overline{\psi(X)} = \sum_{k=1}^n p_k \cdot \psi(x_k)$$



# Moments

$$E(X) = \bar{X} = \text{first-order moment of } X = \sum_{k=1}^n p_k x_k$$

$$E(X^2) = \overline{X^2} = \text{second-order moment of } X = \sum_{k=1}^n p_k x_k^2$$

$$E(X^3) = \overline{X^3} = \text{third-order moment of } X = \sum_{k=1}^n p_k x_k^3$$