



## Measures of dispersion or variation

Measures of centering are not sufficient to describe a set of data completely.

Some samples may have the same arithmetic mean despite the difference in the distribution of their data around their center.

The degree of data homogeneity, such as the following two sets of data.

(150, 125, 100, 50, 75), (95, 97, 100, 103, 105)

It is noted that they have the same arithmetic mean and median, which are 100, while the two groups differ in terms of their dispersion around the center or middle. The degree to which we tend to spread digital data around a middle value is called the dispersion or distribution of data. The most important measures



of dispersion are (range, variance, standard deviation, and mean deviation).

**Mean deviation:** It is the average of the absolute deviations from the arithmetic mean and is symbolized by the symbol (M.D) and is calculated according to

**for ungrouped data**

$$M.D = \frac{\sum |yi - \bar{y}|}{n}$$

**Example:**

**Find the mean deviation of the following values**

**$yi = 9, 8, 6, 5, 7$**

**Sol/**

$yi$	9	8	6	5	7	$\sum yi = 35; \bar{y} = 7$
$yi - \bar{y}$	2	1	-1	-2	0	$\sum (yi - \bar{y}) = 0$
$ yi - \bar{y} $	2	1	1	2	0	$\sum  yi - \bar{y}  = 6$

$$M.D = \frac{\sum |yi - \bar{y}|}{n} = \frac{6}{5} = 1.2$$



**for grouped data**

$$M.D = \frac{\sum f_i |y_i - \bar{y}|}{f_i}$$

**Example:**

**Find the mean deviation of the following frequency distribution table**

class	60-62	63-65	66-68	69-71	72-74
fi	5	18	42	27	8

**Sol/**

Class	Fi	yi	fi*yi
60-62	5	61	305
63-65	18	64	1152
66-68	42	67	2814
69-71	27	70	1880
72-74	8	73	584
Σ	100		6745



$$\bar{y} = \frac{\sum f_i y_i}{\sum f_i} = \frac{6745}{100} = 67.45$$

$ y_i - \bar{y} $	$f_i  y_i - \bar{y} $
6.45	32.25
3.45	62.10
0.45	18.90
2.55	68.85
5.55	44.40
	226.50

$$M.D = \frac{\sum f_i |y_i - \bar{y}|}{\sum f_i} = \frac{226.50}{100} = 2.265$$



## H.W/ Find the mean deviation of the following frequency distribution table

class	Frequency
51-57	7
58-64	6
65-71	7
72-78	8
79-85	16
86-92	10
93-99	5

## Variance and standard deviation

We notice that the sum of the deviations of the sample elements (items) from their arithmetic mean ( $\sum |y_i - \bar{y}| = 0$ ) because some of the deviations are positive and others are negative. To overcome this problem, it was treated by taking the absolute values of the deviations in (the average deviation), and it can be treated in another way, which is



squaring the values of the deviations to obtain the sum of the squares of the deviations (sum squares) and it is symbolized by (SS).

$SS = \sum (y_i - \bar{y})^2$  In order to take into account the sample size so that we can compare samples of different sizes, SS is divided by the degrees of freedom (n-1), which leads to obtaining the variance (Variance  $S^2$  ).

**for grouped data**

$$s^2 = \frac{\sum (y_i - \bar{y})^2}{n - 1} = \frac{\sum y_i^2 - \frac{(\sum y_i)^2}{n}}{n - 1}$$

In the case of society, it is calculated according to

$$\sigma^2 = \frac{\sum (y_i - \mu)^2}{N}$$



$\mu$  = In the middle of the computing community

$N$  = Number of community vocabulary

The reason for taking  $(n-1)$  in the case of a sample is that when a sample is drawn,  $(n-1)$  of the observations are free values, while the remaining observation (the last) must complete its deviation from the arithmetic mean to zero, so the number of free values in any sample is  $(n-1)$ , which is called degrees of freedom. In order to return the units of measurement for the sample items to their origin, the square root of the variance must be taken to obtain  $(S = \sqrt{S^2})$ , which is called the standard deviation.



## Derivation

$$\begin{aligned}SS &= \Sigma(yi - \bar{y})^2 = \Sigma(yi^2 - 2yi\bar{y} + \bar{y}^2) \\&= \Sigma yi^2 - 2\Sigma yi * \left(\frac{\Sigma yi}{n}\right) + n\left(\frac{\Sigma yi}{n}\right)^2 \\&= \Sigma yi^2 - 2\frac{(\Sigma yi)^2}{n} + n\left(\frac{\Sigma yi}{n}\right)^2 \\SS &= \Sigma yi^2 - \frac{(\Sigma yi)^2}{n} \\S^2 &= \frac{ss}{n-1} = \frac{\Sigma(yi - \bar{y})^2}{n-1} = \frac{\Sigma yi^2 - \frac{(\Sigma yi)^2}{n}}{n-1} \\\rightarrow S &= \sqrt{S^2} = \sqrt{\frac{\Sigma(yi - \bar{y})^2}{n-1}} = \sqrt{\frac{\Sigma yi^2 - \frac{(\Sigma yi)^2}{n}}{n-1}}\end{aligned}$$

for grouped data

$$S = \sqrt{S^2} = \sqrt{\frac{\Sigma fi(yi - \bar{y})^2}{\Sigma fi - 1}} = \sqrt{\frac{\Sigma fiyi^2 - \frac{(\Sigma fiyi)^2}{\Sigma fi}}{\Sigma fi - 1}}$$





### Example:

**Find the standard deviation of the following data using the shorthand method.**

**$y_i = 9, 8, 6, 5, 7$**

**Sol/**

$y_i$	9	8	6	5	7	$\Sigma y_i = 35$
$y_i^2$	81	64	36	25	49	$\Sigma y_i^2 = 255$

$$S = \sqrt{\frac{\Sigma y_i^2 - \frac{(\Sigma y_i)^2}{n}}{n-1}} = \sqrt{\frac{255 - \frac{(35)^2}{5}}{5-1}} = \sqrt{\frac{255 - 245}{4}} = \sqrt{\frac{10}{4}} = 1.58$$

### Example:

**Find the standard deviation of the following data:**

Class	$F_i$
60–62	5
63–65	18
66–68	42
69–71	27
72–74	8
$\Sigma$	100



**Sol/**

Class	$f_i$	$Y_i$	$y_i - \bar{y}$	$(y_i - \bar{y})^2$	$f_i * (y_i - \bar{y})^2$
60-62	5	61	-6.45	41.6025	208.0125
63-65	18	64	-3.45	11.9025	214.2450
66-68	42	67	-0.45	0.2025	8.5050
69-71	27	70	2.55	6.5025	175.5675
72-74	8	73	5.55	30.8025	246.4200
$\Sigma$	100	$\bar{y} = 67.45$			$\Sigma = 852.7500$

**H.W/**

**Find the standard deviation of the following data:**

class	Frequency
51-57	7
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93-99	5



## The most important properties of variance and standard deviation

1. When adding or subtracting a fixed number K from the values of observations, it does not affect the value of variance and standard deviation of those observations.

If  $X_i = y_i + k$  or  $X_i = y_i - k$  then  $S_x^2 = S_y^2$  or  $S_x = S_y$

Example/

$X_i = y_i + 3 = 11, 6, 5, 15, 13$

$y_i = 8, 3, 2, 12, 10$        $\Sigma y_i = 35$

$y_i^2 = 64, 9, 4, 144, 100$        $\Sigma y_i^2 = 321$

$$S = \sqrt{\frac{\Sigma y_i^2 - \frac{(\Sigma y_i)^2}{n}}{n-1}} = \sqrt{\frac{321 - \frac{(35)^2}{5}}{5-1}} = \sqrt{\frac{321 - 245}{4}} = \sqrt{19} = 4.36$$

$X_i = 11, 6, 5, 15, 13$        $\Sigma X_i = 50$

$X_i^2 = 121, 36, 25, 225, 169$        $\Sigma X_i^2 = 576$

$$S_X = \sqrt{\frac{\Sigma y_i^2 - \frac{(\Sigma y_i)^2}{n}}{n-1}} = \sqrt{\frac{576 - \frac{(50)^2}{5}}{5-1}} = \sqrt{\frac{576 - 500}{4}} = \sqrt{19} = 4.36$$



2. If each of the observation values is multiplied by a constant number (k), then

- The variance of the resulting values = the variance of the original values \* the square of k.
- The standard deviation of the resulting values = the standard deviation of the original values \* k.