

QMS 101 Introductory Statistics

Topic IV: Measures of Dispersion

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Learning Outcomes

By the end of this topic, students will be able to:

- ▶ Explain the meaning and purpose of **measures of dispersion**
- ▶ **Derive** the formulas for range, mean deviation, variance, standard deviation, and coefficient of variation
- ▶ Calculate all five measures for ungrouped and grouped data
- ▶ Use the **coefficient of variation** to compare variability across different datasets
- ▶ Interpret every result in plain language and link dispersion to the Topic III mean

Our Running Dataset

The Study Context

A farm manager in a rural village wants to estimate the typical monthly expenditure on farm inputs (seeds, fertilisers, pesticides) among small-scale farmers.

She collected the following data (**TZS '000**) from **30 farmers**:

45 32 55 38 47 60 33 50 42 38
55 47 30 65 42 50 38 55 47 44
52 36 48 58 41 53 37 61 43 49

Frequency table (from Topic III):

Class (TZS '000)	f	Midpoint x	F
30 – 38	7	34	7
39 – 47	10	43	17
48 – 56	7	52	24
57 – 65	6	61	30
Total	30		

4.1 What is Dispersion — and Why Does It Matter?

The Limitation of Central Tendency Alone

Two Villages — Same Average, Very Different Reality

A ministry official reports that the **average** monthly farm input expenditure in two villages is the same: **TZS 46,600**.

Village A (TZS '000)	Village B (TZS '000)
44, 45, 47, 46, 48, 47, 46	10, 20, 65, 80, 15, 72, 64
Mean = 46.7	Mean = 46.6

Are the two villages really the same?

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Mean = 46.7	Mean = 46.6

Are the two villages really the same?

No. Village A farmers all spend similarly — the mean is a reliable guide.

Village B farmers have wildly different spending — the mean misleads completely.

Definition

Measures of dispersion (also called measures of variability or spread) quantify how much the values in a dataset **differ from each other** and from the central value.

Visualising the Problem

Same Mean – Very Different Spread

Both villages have mean . TZS 46,600. But Village B values are far more scattered.

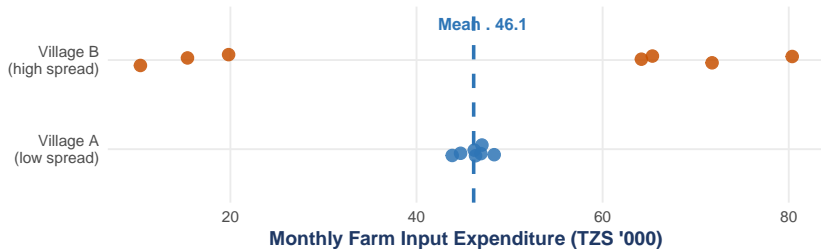


Figure 1

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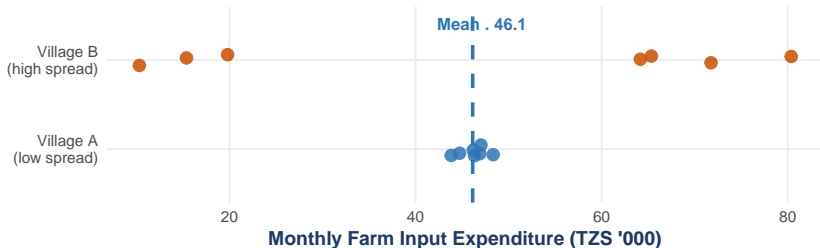


Figure 1

Central tendency answers: *Where is the centre?*

Dispersion answers: *How reliable is that centre as a summary?*

Five Measures of Dispersion

Measure (Symbol)	What It Measures
Range (R)	Distance from smallest to largest value
Mean Deviation (MD)	Average of absolute distances from the mean
Variance (σ^2 or s^2)	Average of squared distances from the mean
Standard Deviation (σ or s)	Typical distance from the mean (same units as data)
Coefficient of Variation (CV)	Relative spread as a percentage (unitless)

Tip

Each measure builds on the previous one.

By the end of this topic you will see that **standard deviation** is the most useful —

4.2 The Range

Deriving the Range

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$$\underbrace{x_{min}}_{\text{smallest}} \quad \dots \quad \underbrace{x_{max}}_{\text{largest}}$$

$$R = x_{max} - x_{min}$$

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$$R = x_{max} - x_{min}$$

That is all. No formula to derive — just the difference between the two extremes.

Range — Our Dataset

From raw data:

Min = 30, Max = 65

$$R = 65 - 30 = \mathbf{TZS\ 35,000}$$

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Interpretation

The farm input expenditure among these 30 farmers spans a range of **TZS 35,000**.

The most frugal farmer spends TZS 30,000; the highest spender pays TZS 65,000 per month.

Range — Advantages and Limitations

Same Range = 35 – But Very Different Distributions

Range uses only 2 values and ignores everything in between.

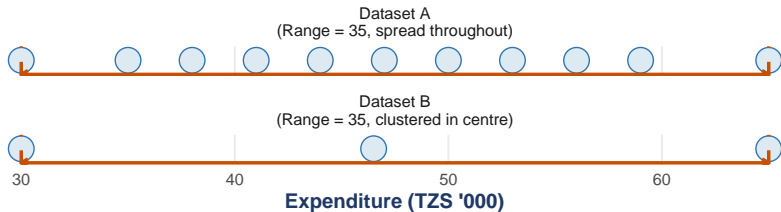


Figure 2

Range — Advantages and Limitations

Same Range = 35 – But Very Different Distributions

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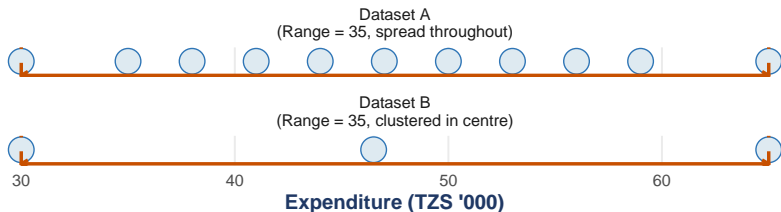


Figure 2

Advantage: Instantly simple to compute and understand.

Limitation: Uses only 2 values — one extreme value completely changes it.

Practice 4.1 — Range

Your Turn

An agricultural extension officer recorded the **number of bags of maize harvested** by 8 smallholder farmers:

12, 28, 15, 34, 9, 22, 31, 18

- (a) Calculate the range of the harvest data.
- (b) A ninth farmer joins the record with a harvest of **52 bags** — an unusually good season.

Recalculate the range. What does this show about the range's weakness?

- (c) The following year, all 9 farmers increase their harvest by 5 bags each.

Without recalculating, what is the new range? Explain your reasoning.

4.3 Mean Deviation

The Core Idea — Measuring from the Centre

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$$\frac{\sum(x_i - \bar{x})}{n} = \frac{0}{n} = 0 \quad \text{always!}$$

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Positive and negative deviations cancel out — giving zero every time.

Solution: Ignore the sign. Take the **absolute value** $|d_i| = |x_i - \bar{x}|$.

Deriving the Mean Deviation Formula

Step 1: Compute each deviation from the mean: $x_i - \bar{x}$

Step 2: Remove the sign (take absolute value): $|x_i - \bar{x}|$

Step 3: Average these absolute deviations:

$$MD = \frac{\sum |x_i - \bar{x}|}{n}$$

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For grouped data — weight each class deviation by its frequency:

$$MD = \frac{\sum f|x - \bar{x}|}{\sum f}$$

where x is the class midpoint and f is the class frequency.

Mean Deviation — Our Dataset (Ungrouped)

$\bar{x} = 46.37$ (ungrouped mean)

x	$x - \bar{x}$	$ x - \bar{x} $
30	-16.37	16.37
32	-14.37	14.37
33	-13.37	13.37
61	+14.63	14.63
65	+18.63	18.63
Total	0.00	$\sum x - \bar{x} = 302.93$

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Total	0.00	$\sum x - \bar{x} = 302.93$

$$MD = \frac{302.93}{30} = \mathbf{TZS\ 10,098}$$

Mean Deviation — Grouped Data

Using $\bar{x} = 46.6$ (grouped mean):

Class	f	x	$x - \bar{x}$	$ x - \bar{x} $	$f x - \bar{x} $
30–38	7	34	-12.6	12.6	88.2
39–47	10	43	-3.6	3.6	36.0
48–56	7	52	+5.4	5.4	37.8
57–65	6	61	+14.4	14.4	86.4
Total	30				248.4

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Interpretation

On average, a farmer's monthly expenditure deviates from the group mean of TZS 46,600 by about **TZS 8,280**. A smaller MD would mean farmers spend more uniformly; a larger MD

Practice 4.2 — Mean Deviation

Your Turn

A veterinary officer recorded the **weight of 6 goats (kg)** at a livestock market:

18, 24, 31, 22, 27, 16

- (a) Calculate the arithmetic mean weight.
(b) Complete the deviation table:

x	$x - \bar{x}$	$ x - \bar{x} $
18		
24		
31		
22		
27		
16		

4.4 Variance

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$$|x - \bar{x}| \longrightarrow (x - \bar{x})^2$$

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A better approach — square the deviations instead:

$$|x - \bar{x}| \longrightarrow (x - \bar{x})^2$$

Squaring achieves two things:

1. Removes the sign (negative squared = positive):
 $(-5)^2 = 25 = (+5)^2$
2. **Penalises large deviations more heavily** — a deviation of 10 gives $10^2 = 100$, while a deviation of 2 gives only $2^2 = 4$.

Deriving the Variance Formula

Start with the same idea as MD — average the “distances” from the mean.

Replace absolute values with squares:

Population variance:
$$\sigma^2 = \frac{\sum (x_i - \mu)^2}{N}$$

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For a sample (which is our case — 30 farmers, not all farmers):

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

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Why $n - 1$ and not n ?

When we estimate the population variance from a **sample**, dividing by n systematically **underestimates** the true population variance. Dividing by $n - 1$ corrects for this.

This correction is called “Bessel’s correction”.

Visualising Variance — Squaring the Deviations

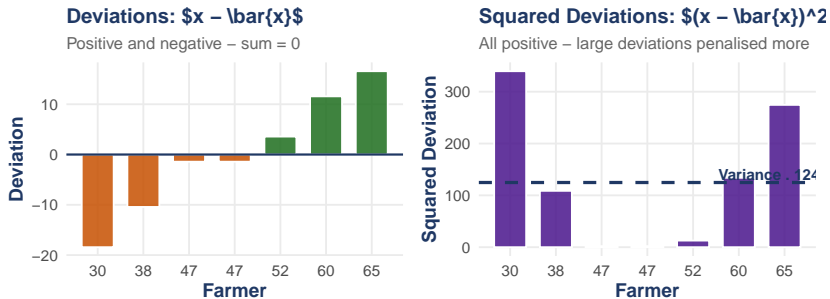


Figure 3

Deriving the Computational Shortcut

Working out $(x - \bar{x})^2$ for every observation is tedious for large datasets.

We can derive a faster equivalent formula algebraically in four steps.

Step 1 — Expand $(x - \bar{x})^2$

$$\sum (x - \bar{x})^2 = \sum (x^2 - 2x\bar{x} + \bar{x}^2)$$

...

Step 2 — Separate** the three sums (since \bar{x} is a constant):

$$\sum (x - \bar{x})^2 = \sum x^2 - 2\bar{x} \underbrace{\sum x}_{=n\bar{x}} + n\bar{x}^2$$

Step 3 — Substitute $\sum x = n\bar{x}$:

$$= \sum x^2 - 2n\bar{x}^2 + n\bar{x}^2 = \sum x^2 - n\bar{x}^2$$

Step 4 — Divide by n :

$$\sigma^2 = \frac{\sum x^2}{n} - \bar{x}^2 \quad (\text{ungrouped})$$

$$\sigma^2 = \frac{\sum fx^2}{\sum f} - \bar{x}^2 \quad (\text{grouped})$$

Shortcut vs. Definition — Quick Numerical Check

Verify with three values: 2, 4, 6 — $\bar{x} = 4$ (treat these data as the entire population)

Definition formula:

x	$x - \bar{x}$	$(x - \bar{x})^2$
2	-2	4
4	0	0
6	+2	4
Total	0	8

$$\sigma^2 = \frac{8}{3} = 2.67$$

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Shortcut formula:

$$\begin{aligned}\sum x^2 &= 4 + 16 + 36 = 56 \\ \bar{x}^2 &= 4^2 = 16\end{aligned}$$

$$\sigma^2 = \frac{56}{3} - 16 = 18.67 - 16 = \mathbf{2.67} \checkmark$$

Both give exactly the same answer.

The shortcut eliminates computing every $(x - \bar{x})$ individually.

Variance — Our Dataset (Ungrouped)

Using the 30 raw values with $\bar{x} = 46.37$: (treat the data as the entire population)

Step	Calculation	Result
$\sum x$	$45 + 32 + 55 + \dots + 49$	1,391
\bar{x}	$1,391 \div 30$	46.37
$\sum x^2$	$45^2 + 32^2 + 55^2 + \dots + 49^2$	66,731
$\frac{\sum x^2}{n}$	$66,731 \div 30$	2,224.37
\bar{x}^2	$(46.37)^2$	2,150.17
σ^2	$2,224.37 - 2,150.17$	74.20

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$\sum x^2$	$45^2 + 32^2 + 55^2 + \dots + 49^2$	66,731
$\frac{\sum x^2}{n}$	$66,731 \div 30$	2,224.37
\bar{x}^2	$(46.37)^2$	2,150.17
σ^2	$2,224.37 - 2,150.17$	74.20

Ungrouped vs. Grouped

Ungrouped Variance = **TZS 74,200** (exact, from raw data)

Variance — Our Dataset (Grouped)

Using $\bar{x} = 46.6$:

Class	f	x	x^2	fx^2
30–38	7	34	1,156	8,092
39–47	10	43	1,849	18,490
48–56	7	52	2,704	18,928
57–65	6	61	3,721	22,326
Total	30			$\sum fx^2 = 67,836$

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$$\sigma^2 = \frac{\sum fx^2}{\sum f} - \bar{x}^2 = \frac{67,836}{30} - (46.6)^2 = 2,261.2 - 2,171.56 = \mathbf{89.64} \text{ (TZ)}$$

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Warning

The variance has **squared units** — (TZS '000)² — which makes it hard to interpret directly.

Practice 4.3 — Variance

Your Turn

The following table shows the **daily water consumption (litres)** recorded at 5 households in a village:

20, 35, 28, 42, 15

- (a) Calculate the arithmetic mean.
(b) Complete the squared-deviation table:

x	$x - \bar{x}$	$(x - \bar{x})^2$
20		
35		
28		
42		
15		

- (c) Calculate the variance (divide by n here, since this is the

4.5 Standard Deviation

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The fix: Take the square root to return to the **original units** of the data.

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Grouped:
$$\sigma = \sqrt{\frac{\sum fx^2}{\sum f} - \bar{x}^2}$$

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The standard deviation answers: *“By how much does a typical observation differ from the mean?”*

It is the **most widely used** measure of dispersion in statistics, science, business, and medicine.

Standard Deviation — Our Dataset

From the variance already computed: $\sigma^2 = 89.64$

$$\sigma = \sqrt{89.64} = \mathbf{9.47 \text{ TZS '000}}$$

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Interpretation

A typical farmer's monthly expenditure on farm inputs deviates from the group mean of **TZS 46,600** by about **TZS 9,470**.

This is the “typical spread” around the average. Most farmers (roughly 68%, if the data is normally distributed) spend within the range:

$$\bar{x} \pm \sigma = 46.6 \pm 9.47 \implies \text{between TZS } \mathbf{37,130} \text{ and } \mathbf{56,070}$$

Visualising Standard Deviation

Standard Deviation – Typical Distance from the Mean

Green band = Mean \pm 1 SD. Most observations fall within this band.

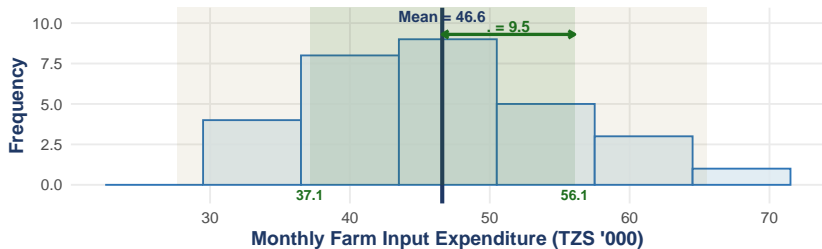


Figure 4

Properties of Standard Deviation

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1. **Always non-negative:** $\sigma \geq 0$. SD equals zero only when all values are identical.

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5. **Multiplying by a constant scales SD:** If all values double, SD also doubles. $\sigma(cx) = |c| \cdot \sigma(x)$.
6. **Basis for all further statistics:** The normal distribution, Z-scores, confidence intervals, and hypothesis tests are all built on SD.

Mean Deviation (MD) vs Standard Deviation (SD)

Feature	Mean Deviation (MD)	Standard Deviation (SD)
Formula	$\sum \ x - \bar{x}\ /n$	$\sqrt{\sum (x - \bar{x})^2/n}$
Uses absolute value?	Yes	No (uses squaring)
Penalises large deviations?	Equally	More (squares amplify large gaps)
Easy to interpret?	Yes — same units	Yes — same units
Used in advanced statistics?	Rarely	Extensively (basis for normal distribution, t-tests, regression)

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Tip

Use SD for almost all purposes. It is the universal standard in statistics and is required for Topic III's empirical rule, probability distributions, and all inferential work in later courses.

Practice 4.4 — Standard Deviation

Your Turn

A seed company tests the **germination time (days)** of a new maize variety on 5 plots:

8, 12, 10, 15, 9

- (a) Calculate the mean germination time.
- (b) Use the shortcut formula $\sigma^2 = \frac{\sum x^2}{n} - \bar{x}^2$ to compute the variance.
- (c) Calculate the standard deviation.
- (d) Apply the empirical rule: between which two values would you expect **68%** of germination times to fall (assuming approximate normality)?
- (e) A plot germinates in 18 days. How many standard deviations above the mean is this?

