

1. Introduction to Measures of Spread

Why Do We Measure Spread?

When we look at a set of data, it's not just about the **average** — we also want to know how **spread out** the data is.

Are most of the values close together? 🤝




Or are they far apart? 🌟

This is where **measures of spread** come in — they help us understand the **variability** in data. 📊

What Are Measures of Spread?




Measures of spread give us important information about **how data values differ from each other**.

They help answer questions like:

-  How wide is the range of scores?
 -  Are there outliers (unusual values)?
 -  Is the data tightly clustered or scattered?
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Common Measures of Spread

Here are three key tools you'll use to measure spread:


1.  **The Range** – The simplest measure, showing the difference between the highest and lowest values.
2.  **The Interquartile Range (IQR)** – Measures the spread of the middle 50% of data, less affected by outliers.
3.  **The Standard Deviation** – Tells us how much the data varies around the **mean**.

2. The Range

What is the Range?

The **range** is the **simplest measure of spread**.


It tells us how far apart the **smallest** and **largest** values in a dataset are.

 It gives a quick sense of how **wide** the data is spread.

Formula for the Range

$$R = \text{largest value} - \text{smallest value}$$

Just subtract the **smallest** number from the **largest**, and you've got it!

 **Large range** = More spread

 **Small range** = Less spread

Worked Example (Example 14): Finding the Range

Let's look at marks from **two different tasks** given to a group of students:

Task A Marks:

2, 6, 9, 10, ..., 94

 Smallest value = 2

 Largest value = 94

 Range for Task A:

$$R = 94 - 2 = \boxed{92}$$

Task B Marks:

11, 16, 19, ..., 91

 Smallest value = 11

 Largest value = 91

 Range for Task B:

$$R = 91 - 11 = \boxed{80}$$

Interpreting the Results


- Task A has a **larger range** (92) compared to Task B (80).
- At first glance, this suggests Task A's scores are **more spread out**.
- BUT... is that the whole story? 🤔
(Stay tuned for Section 3 to find out why **range** might not always tell the full truth! 🔍)

3. Limitations of the Range

The Problem with the Range

While the **range** is super easy to calculate, it's not always **reliable**.

Why? Because it can be heavily affected by **outliers** — values that are **much higher or lower** than the rest.

 Just one unusual number can make the range look **way bigger** than it should be!

Visual Check: Stem-and-Leaf Plots

Let's compare Task A and Task B using **stem-and-leaf plots**:

Task A

Stem	Leaf
0	2 6 9
1	0 1 2 3
2	3 4 6 6 7
3	4 5 8 8 9
4	2 6 7
5	2 2 6 6 9
6	0
7	
8	
9	1 4

Task B

Stem	Leaf
0	
1	1 6 9
2	1 3 8
3	1 1 3 8
4	1 9
5	2 3 4 6 9
6	3 5 8
7	1 2 3 5 8 8

8 | 6 8

9 | 1

What Do We See?

- Task A has **two very high scores** (91 and 94) that stretch the range. 🎯
 - Most of Task A's scores are **clustered in the middle**, with just a couple of outliers.
 - Task B's scores are **more evenly spread** — the range reflects the distribution **a bit more accurately**.
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Conclusion: We Need a Better Tool!

The **range** doesn't always give us the full picture 📊.

We need a measure that:

- Focuses on the **middle values**
- Ignores extreme outliers 🚫

👑 That's where the **Interquartile Range (IQR)** comes in — get ready for it in the next section! 📦
💡

4. The Interquartile Range (IQR)

Definition

The **interquartile range (IQR)** tells us how spread out the **middle 50%** of the data is. Unlike the range, it ignores the **lowest 25%** and **highest 25%** of values.

 It's a **more reliable** measure of spread, especially when outliers are present!






Formula for IQR

$$\text{IQR} = Q_3 - Q_1$$

Where:

- Q_1 = First Quartile (25th percentile)
 - Q_3 = Third Quartile (75th percentile)
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Steps to Find the IQR

1.  Order the data from smallest to largest
2.  Split the data into two equal halves
 - Lower half (first 50%)
 - Upper half (last 50%)
3.  Find Q_1 = median of the lower half
4.  Find Q_3 = median of the upper half
5.  Subtract:

$$\text{IQR} = Q_3 - Q_1$$

Worked Example (Example 15): Finding the IQR

Let's find the IQR for Task A and Task B:

Task A Data

- 30 total values
- Split into:
 - Lower half: 2, 6, 9, 10, 11, 12, 13, **22**, 23, 24, 26, 26, 27, 33, 34
 - $Q_1 = 22$
 - Upper half: 35, 38, 38, 39, 42, 46, **47**, 47, 52, 52, 56, 56, 59, 91, 94
 - $Q_3 = 47$

 IQR for Task A:

$$IQR = 47 - 22 = \boxed{25}$$



Task B Data

- 30 total values
- Split into:
 - Lower half: 11, 16, 19, 21, 23, 28, **31**, 31, 33, 38, 38, 41, 49, 52, 53
 - $Q_1 = 31$
 - Upper half: 54, 56, 59, 63, 65, 68, **71**, 72, 73, 75, 78, 78, 86, 88, 91
 - $Q_3 = 73$

 IQR for Task B:


$$IQR = 73 - 31 = \boxed{42}$$

Comparison & Conclusion

-  Task A IQR = 25 → Scores are more **concentrated**
-  Task B IQR = 42 → Scores are more **spread out**

 What it means:

The IQR shows that Task A's data is less variable than Task B's data.

It's a clearer picture of the real spread—without being tricked by outliers! 



5. The Standard Deviation



What Is the Standard Deviation?

The standard deviation (s) is a powerful measure of spread that tells us how much data values deviate from the mean (average).

If values are close to the mean →  Small standard deviation

If values vary a lot from the mean →  Large standard deviation



Formula

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

Where:

- x = individual data value
 - \bar{x} = mean (average) of the data
 - n = number of values
 - \sum = sum of all the values
 - s = sample standard deviation
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Breaking It Down

◆ 1. \bar{x} – The Mean

Find the average of all your values:

$$\bar{x} = \frac{\sum x}{n}$$

◆ 2. $x - \bar{x}$ – Deviation from the Mean

For each value, find how far it is from the mean.

◆ 3. $(x - \bar{x})^2$ – Squared Deviations

Square each deviation so that positive and negative differences don't cancel out. 

◆ 4. $\sum(x - \bar{x})^2$ – Total Variation

Add all the squared deviations together.

◆ **5. Divide by $n - 1$**

This is called **Bessel's correction** — it adjusts for the fact that we're working with a **sample** (not the whole population).

✅ It helps get a more accurate estimate of variability.

◆ **6. Take the Square Root**

This brings the units back to the original scale of the data.

! **Why Use $n - 1$ Instead of n ?**

Because we're often working with a **sample**, not the full population.

Using $n - 1$ prevents us from **underestimating** the spread.

📊 It corrects the bias and gives a **more accurate picture** of how variable the data really is.

✅ So, while the **range** and **IQR** give you a sense of spread...

📈 **Standard deviation** gives you a **precise, mathematical** look at how data clusters around the mean!

6. Worked Example: Calculating the Standard Deviation

Let's calculate the sample standard deviation for this simple data set:

 **Data Set:**

$$\{2, 3, 4\}$$


 **Step 1: Find the Mean \bar{x}**

$$\bar{x} = \frac{2 + 3 + 4}{3} = \frac{9}{3} = 3$$

 The mean is 3

 **Step 2: Create a Table**

x	$x - \bar{x}$	$(x - \bar{x})^2$
2	$2 - 3 = -1$	$(-1)^2 = 1$
3	$3 - 3 = 0$	$(0)^2 = 0$
4	$4 - 3 = 1$	$(1)^2 = 1$

 Sum of squared deviations:

$$1 + 0 + 1 = 2$$

 **Step 3: Apply the Formula**

$$s = \sqrt{\frac{\sum(x - \bar{x})^2}{n - 1}} = \sqrt{\frac{2}{3 - 1}} = \sqrt{\frac{2}{2}} = \sqrt{1}$$

$$s = 1$$

 **Final Answer:**



The sample standard deviation of the data set $\{2, 3, 4\}$ is:

$$1$$



What Does Standard Deviation Tell Us?



The **standard deviation** is a number that tells us how spread out the values in a dataset are from the mean (average).

-  **Small SD (e.g., SD = 1):**
Most values are **close to the mean** → data is **tightly clustered**.
→ Example: Test scores where most students scored around 80.
 -  **Larger SD (e.g., SD = 2 or more):**
Values are **more spread out** → more **variation or inconsistency**.
→ Example: Some students scored 50, some 90, even though the average is still 70.
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How to Interpret SD Values

Imagine a class of students with a mean test score of 70:

-  **SD = 1:** Most students scored **between 69 and 71** → very consistent performance
-  **SD = 10:** Scores vary **widely**, from **60 to 80** and beyond → large performance gap

So, SD helps you understand:

- Are the results **similar** (low SD)?
 - Or are they **all over the place** (high SD)?
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Real-Life Examples of Standard Deviation



1. Education – Test Scores

Let's say two classes have the same **average** test score (75), but:

- **Class A** has **SD = 2** → Most students scored around 75
- **Class B** has **SD = 15** → Some students got 40, some got 100

 This helps a teacher identify which class had **more consistent learning outcomes**.



2. Business – Sales Performance

A company tracks weekly sales across branches:

- Branch X has $SD = \$200$ → Stable sales each week
- Branch Y has $SD = \$1,000$ → Some weeks are great, some terrible

✅ Managers can use this to decide where to **invest training or resources** to stabilize performance.



3. Health – Blood Pressure Readings

In a clinical trial, if one group has blood pressure with **low SD**, the treatment is **working consistently**.

But if the SD is high, it suggests the treatment has **varying effects** — it might not be reliable.



4. Manufacturing – Quality Control

If a machine produces metal rods with length = 10cm:

- $SD = 0.1\text{cm}$ → Most rods are between 9.9 and 10.1cm → Excellent consistency
- $SD = 0.5\text{cm}$ → Lots of variation → May cause issues with product quality

✅ Lower SD = more **precision and quality**



Key Takeaway

Standard deviation is a tool to understand consistency.

The smaller the SD, the more predictable and reliable the data.

The larger the SD, the more uncertainty or variation there is.