

# VARIANCE, COVARIANCE, AND CORRELATION

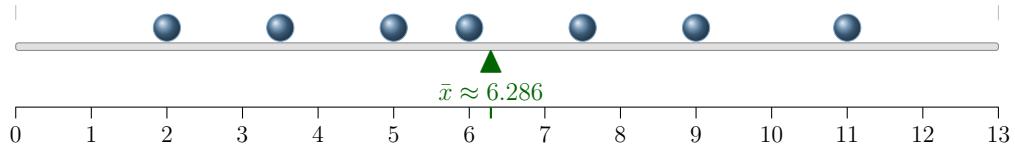
Mr. Merrick · September 29, 2025

## 1) Dataset and Means

Label	A	B	C	D	E	F	G	Totals
$x_i$	2.0	3.5	5.0	6.0	7.5	9.0	11.0	$\sum x_i = 44.0$
$y_i$	0.2	3.6	0.4	3.2	1.0	3.8	1.3	$\sum y_i = 13.5$

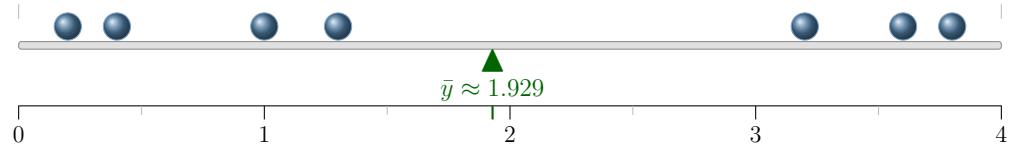
Think of each value as a small *weight* sitting on a beam. Without calculating, *eyeball* where the beam would balance and mark your guess on the ruler line below, and draw in a fulcrum.

Along the  $x$ -axis:



Mark the fulcrum at the balancing point ( $\bar{x}$ ) under the beam. Calculate  $\bar{x}$ .

Along the  $y$ -axis:



Mark the fulcrum at the balancing point ( $\bar{y}$ ) under the beam. Calculate  $\bar{y}$ .

### Quick practice (Means)

1. On the balance beam, do spheres closer to the balance point or farther from it have a greater effect on where it balances? Why?

Spheres farther from the balance point have a greater effect. Torque is weight  $\times$  lever arm. With equal weights, the contribution to shifting the balance is proportional to the distance  $|x_i - \bar{x}|$ .

2. If every  $y_i$  is increased by the same constant  $a$ , how does the balance point on the  $y$ -beam move?

It shifts up by  $a$ : if  $y'_i = y_i + a$ , then  $\bar{y}' = \bar{y} + a$ .

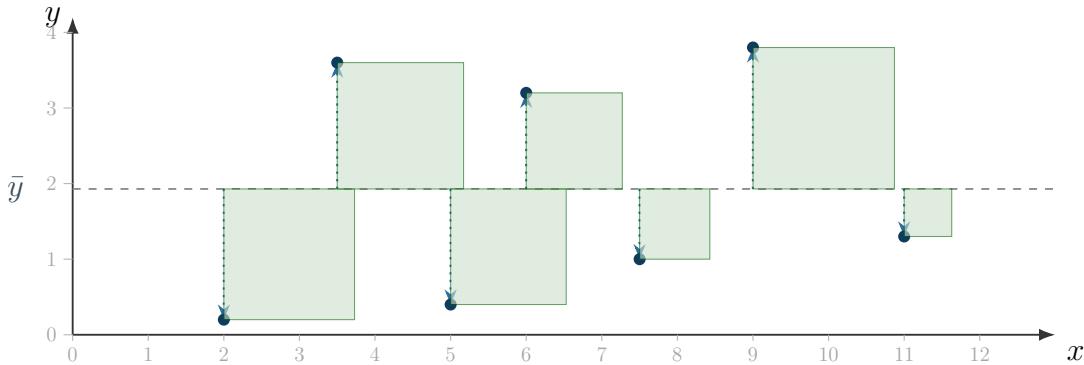
3. If all  $x$ -values are multiplied by a factor  $a$  (scaled), what happens to the balance point on the  $x$ -beam?

It scales by the same factor: if  $x'_i = ax_i$ , then  $\bar{x}' = a\bar{x}$ . For  $a > 0$  it stretches/compresses; for  $a < 0$  it also reflects across 0.

We will use these same seven points in every section.

## 2) Variance of $y$ (sample): average of squared deviations from mean

The horizontal dashed line is at  $\bar{y} = 1.929$ . Each dotted arrow has length  $|y_i - \bar{y}|$ . For every point, draw a **square** using that arrow as one side. Area =  $(y_i - \bar{y})^2$ . Your squares will overlap.



$$\text{Variance in } y \text{ (sample): } s_y^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$$

Point	$y_i$	$y_i - \bar{y}$	$(y_i - \bar{y})^2$
A	0.2	-1.729	2.988
B	3.6	1.671	2.794
C	0.4	-1.529	2.337
D	3.2	1.271	1.617
E	1.0	-0.929	0.862
F	3.8	1.871	3.502
G	1.3	-0.629	0.395
$\sum y_i = 13.5$		14.494	

### Practice (Variance in $y$ )

- Which point lies farthest from the mean line (largest vertical deviation)? Which is closest? Explain using the diagram.

Farthest: point F ( $y = 3.8$ ) with  $|y_i - \bar{y}| \approx 1.871$ . Closest: point G ( $y = 1.3$ ) with  $|y_i - \bar{y}| \approx 0.629$ . On the plot, F has the longest vertical dotted arrow from the mean line, and G has the shortest.

- If every  $y_i$  were shifted upward by +2, would the variance  $s_y^2$  change? Explain geometrically.

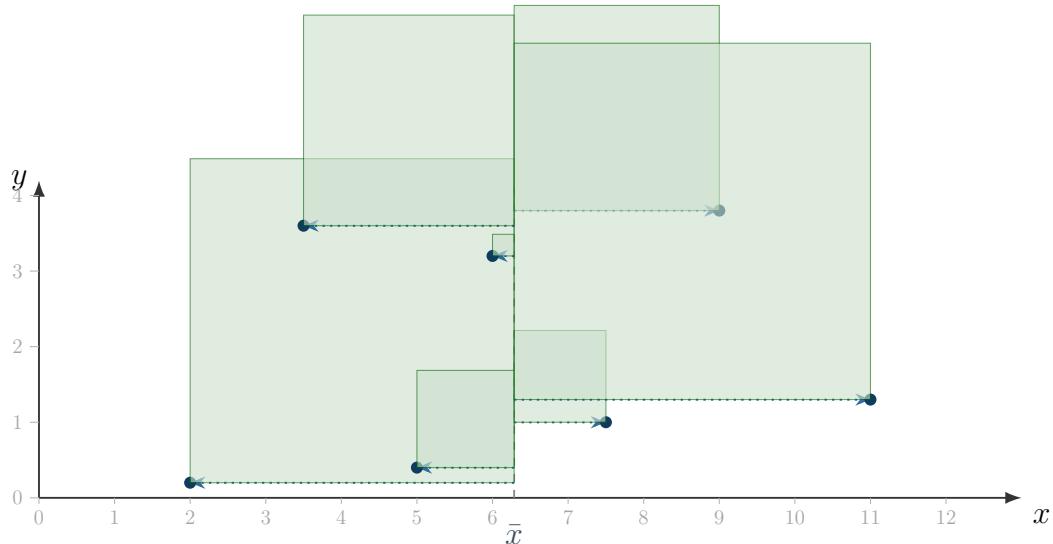
No change. A uniform shift replaces each deviation by  $y'_i - \bar{y}' = (y_i + 2) - (\bar{y} + 2) = y_i - \bar{y}$ , so the squared deviations and their average stay the same. Geometrically, all arrows translate without changing lengths.

- Compute the total sum of squares in  $y$ ,  $SST_y = \sum (y_i - \bar{y})^2$ . What proportion of this sum comes from points above the mean  $\bar{y}$ ?

$SST_y \approx 14.494$ . For the points above the mean (B, D, F):  $2.794 + 1.617 + 3.502 = 7.913$ . Proportion  $\approx 7.913/14.494 \approx 0.546$  (about 54.6%).

### 3) Variance of $x$ (sample): average of squared deviations from mean

The vertical dashed line is at  $\bar{x} = 6.286$ . Each dotted horizontal arrow has length  $|x_i - \bar{x}|$ . Draw squares using that arrow as one side. Area =  $(x_i - \bar{x})^2$ . Your squares will overlap.



$$\text{Variance in } x \text{ (sample): } s_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

Point	$x_i$	$x_i - \bar{x}$	$(x_i - \bar{x})^2$
A	2.0	-4.286	18.367
B	3.5	-2.786	7.760
C	5.0	-1.286	1.653
D	6.0	-0.286	0.082
E	7.5	1.214	1.474
F	9.0	2.714	7.367
G	11.0	4.714	22.224
$\sum x_i = 44.0$			58.929

#### Practice (Variance in $x$ )

- Which points contribute most strongly to  $s_x^2$ ? How can you tell just by looking at the diagram?

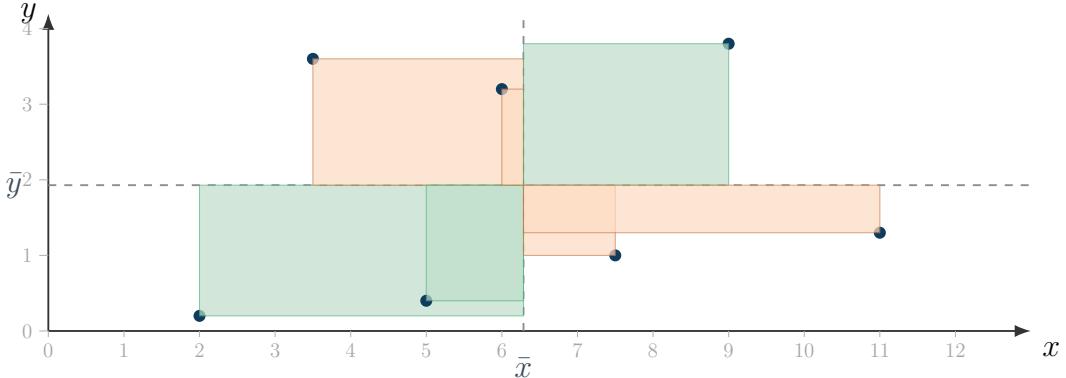
Points farthest from  $\bar{x}$  contribute most because each term is  $(x_i - \bar{x})^2$ . Here, G ( $x = 11.0$ ) with  $|x_i - \bar{x}| \approx 4.714$  and A ( $x = 2.0$ ) with  $|x_i - \bar{x}| \approx 4.286$  contribute the most; they have the longest horizontal dotted arrows.

- If every  $x$ -value were rescaled by a factor  $k$  ( $x'_i = kx_i$ ), how would the variance  $s_x^2$  change?

It scales by  $k^2$ :  $s_{x'}^2 = \frac{1}{n-1} \sum (kx_i - k\bar{x})^2 = k^2 \frac{1}{n-1} \sum (x_i - \bar{x})^2 = k^2 s_x^2$ . (For negative  $k$ , the sign flips but the square makes the factor  $k^2$ .)

#### 4) Covariance (sample): average of signed rectangle areas

Draw a rectangle for each point with side lengths  $|x_i - \bar{x}|$  and  $|y_i - \bar{y}|$ . Quadrants I & III are positive; Quadrants II & IV are negative. Your rectangles will overlap.



$$\text{Covariance (sample): } \text{Cov}(X, Y) = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

Point	$x_i$	$y_i$	$x_i - \bar{x}$	$y_i - \bar{y}$	$(x_i - \bar{x})(y_i - \bar{y})$
A	2.0	0.2	-4.286	-1.729	7.408
B	3.5	3.6	-2.786	1.671	-4.656
C	5.0	0.4	-1.286	-1.529	1.965
D	6.0	3.2	-0.286	1.271	-0.363
E	7.5	1.0	1.214	-0.929	-1.128
F	9.0	3.8	2.714	1.871	5.080
G	11.0	1.3	4.714	-0.629	-2.963
$\sum x_i = 44.0$		$\sum y_i = 13.5$		5.343	

#### Practice (Covariance)

- If you swapped the roles of  $x$  and  $y$ , would the covariance change? Why or why not?

No change. Covariance is symmetric:  $\text{Cov}(X, Y) = \text{Cov}(Y, X)$  because  $(x_i - \bar{x})(y_i - \bar{y})$  is the same product either way.

- For a scatterplot with a strong positive linear trend, what do you expect the sign and size of the covariance to be? What about a strong negative trend?

Positive trend  $\Rightarrow$  covariance  $> 0$  (points with  $x_i > \bar{x}$  tend to have  $y_i > \bar{y}$  and vice versa). Negative trend  $\Rightarrow$  covariance  $< 0$ . The tighter and more spread-out the cloud along the line, the larger the magnitude  $|\text{Cov}(X, Y)|$ .

- If all  $y$  values were doubled, how would the covariance change? Explain your reasoning.

It doubles:  $\text{Cov}(X, 2Y) = \frac{1}{n-1} \sum (x_i - \bar{x})(2(y_i - \bar{y})) = 2 \text{Cov}(X, Y)$ . In general, scaling one variable by  $c$  scales covariance by  $c$ .

## 5) Correlation

After computing the sample variances and the sample covariance above, compute the (sample) correlation:

$$r = \frac{\text{Cov}(X, Y)}{s_x s_y} = \frac{1}{n-1} \sum_{i=1}^n \frac{(x_i - \bar{x})(y_i - \bar{y})}{s_x s_y} \quad \text{where } s_x = \sqrt{s_x^2}, \quad s_y = \sqrt{s_y^2}.$$

**Summary table (from your work above):**

	$s_x^2$	$s_y^2$	$\text{Cov}(X, Y)$	$r = \frac{\text{Cov}(X, Y)}{s_x s_y}$
Values	9.821	2.416	0.890	0.183

### Practice (Correlation)

- If  $x_i$  is measured in centimeters and  $y_i$  in grams, why might correlation ( $r$ ) be easier to interpret than covariance?

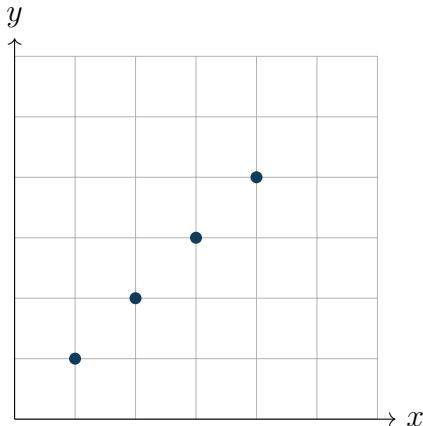
Because correlation is unitless and always between  $-1$  and  $1$ . Covariance depends on the units (cm·g here) and can be hard to interpret in absolute terms. Correlation standardizes by  $s_x$  and  $s_y$ , making it scale-free.

- Two datasets can have the same correlation  $r$  but look very different when graphed.

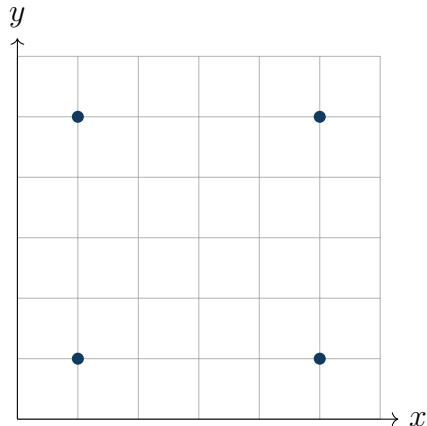
For example, one dataset could be tightly clustered around a line, while another could be more spread out but still linear. Or one could have two distinct subgroups aligned along the same slope. Both give the same  $r$ , but the shape and distribution differ.

- Draw two scatterplots with 4 points each: one with correlation  $r = 1$  (perfect positive linear relationship), and one with correlation  $r = 0$  (no linear relationship).

For  $r = 1$ , all four points lie exactly on an increasing straight line. For  $r = 0$ , the points are arranged so there is no linear trend (e.g. a square or cross shape).



$$r = 1$$



$$r = 0$$

4. If  $x$  is rescaled from centimeters to meters, how does the correlation  $r$  change (if at all)? Explain.

It does not change. Correlation is invariant under positive rescaling of either variable: multiplying all  $x_i$  by a constant rescales both numerator and denominator equally, leaving  $r$  unchanged.