

LEARN

Deborah owns a smoothie cart, and she tracked daily high temperatures alongside how many smoothies she sold. When she plotted the data, the points formed a clear upward trend: Hotter days meant more sales. To turn that trend into something she could actually use for planning, she needed a single equation that best fit all those data points. That equation is a **least squares regression line**, and it lets her plug in a temperature and get a predicted sales number.

The chart below shows the daily high temperatures and smoothie sales Deborah recorded over several weeks; this is the actual data her regression equation was built from.



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The smoothie cart's least squares regression model is:

$$y = 5.2x - 140$$

As a reminder: x represents the input value (the daily high temperature, in °F), and y represents the predicted output value (the number of smoothies sold).

This equation has two key parts: the **constant term** and the **rate of change**. Expand each section below to examine the algebra.

The Constant Term (y -intercept = -140)



The constant term is the value of y when $x = 0$.

Substitute $x = 0$:

$$y = 5.2(0) - 140$$

$$y = 0 - 140$$

$$y = -140$$

Selling **-140 smoothies** is physically impossible. This does not represent a real prediction at 0°F . The constant term anchors the regression line at the correct position on the graph; it is a mathematical artifact, not a meaningful forecast. Always check whether the y -intercept makes sense in context before interpreting it literally.

The Rate of Change (slope = 5.2)



The slope tells you how much y changes for every 1-unit increase in x .

Verify at two consecutive temperatures:

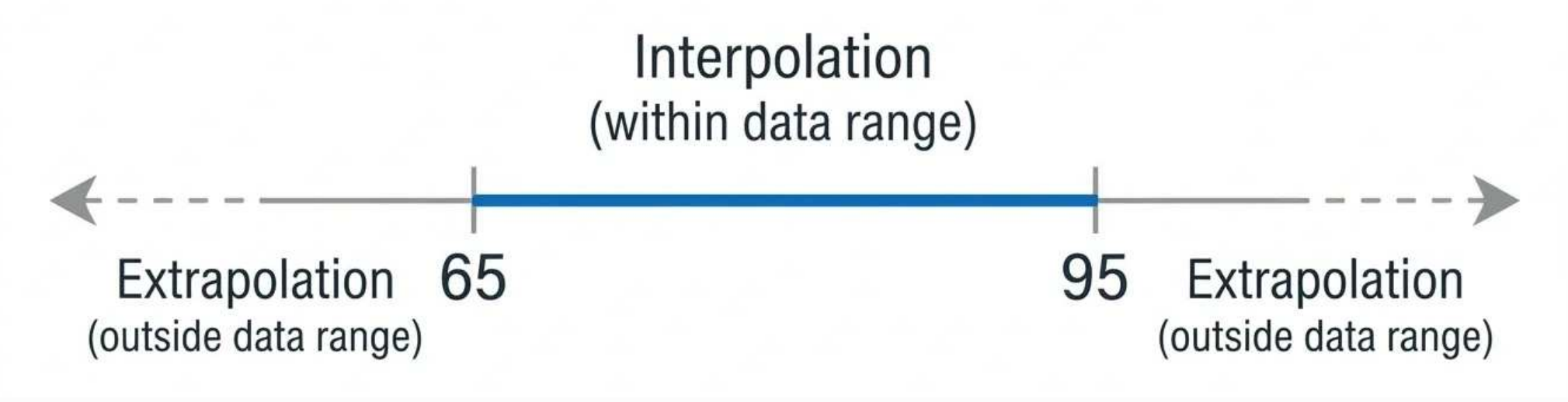
$$y(80) = 5.2(80) - 140 = 416 - 140 = 276$$

$$y(81) = 5.2(81) - 140 = 421.2 - 140 = 281.2$$

$$\text{Difference} = 281.2 - 276 = 5.2 \checkmark$$

Interpretation: For every additional 1°F in temperature, the model predicts **5.2 more smoothies** sold. Always state slope with units (in this case, *smoothies per degree Fahrenheit*).

Not every prediction from a regression model is equally trustworthy. Whether an x -value falls inside or outside the range of the original data changes how much you can trust the result. Let's look at both cases: interpolation and extrapolation.



Interpolation is a prediction made using an x -value that falls *within* the range of the original data (the smoothie cart's data spans $65^{\circ}F$ – $95^{\circ}F$). Because the model was built using data in this range, interpolated predictions are generally reliable.

Worked example: $x = 85^{\circ}F$

$$y = 5.2(85) - 140$$

$$y = 442 - 140$$

$$y = 302 \text{ smoothies}$$

$85^{\circ}F$ falls inside the data range, so this is an interpolation.

Try it yourself. Use the model $y = 5.2x - 140$ to predict smoothie sales when $x = 90^{\circ}F$. Explain what y is, and show your substitution steps.

Type answer

Extrapolation is making a prediction using an x -value that falls *outside* the range of the original data. The algebra still works, but the real-world meaning may not.

Worked example: $x = 112^{\circ}F$

$$y = 5.2(112) - 140$$

$$y = 582.4 - 140$$

$$y = 442.4 \text{ smoothies}$$

The original data was collected over a normal range: roughly $65^{\circ}F$ to $95^{\circ}F$. At $112^{\circ}F$, parks issue heat advisories, people stay indoors, and a linear trend no longer holds. **The model keeps going. Reality does not.**

Try it yourself. A forecaster predicts a $105^{\circ}F$ day. Use the model $y = 5.2x - 140$ to predict smoothie sales at $x = 105^{\circ}F$, and then state whether this is an interpolation or an extrapolation and why.

Type answer

Interpolation vs. Extrapolation

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Interpolation: prediction within the data range → Generally reliable

Extrapolation: prediction outside the data range → Use with caution

The further outside the data you predict, the less you can trust the model.

THINK

Use the regression model $y = 5.2x - 140$ to predict smoothie sales when $x = 72^\circ F$. Show the substitution steps and final answer.

Type answer

A heat wave is forecast to reach 112°F. The regression model gives $y = 5.2(112) - 140 = 442.4$ smoothies. Which statement **best** explains why this prediction is unreliable?

- Select one
- 1

The equation contains a calculation error: 5.2×112 does not equal 582.4.

The arithmetic is actually correct: $5.2 \times 112 = 582.4$, and $582.4 - 140 = 442.4$. The problem isn't the calculation; it's what the number means.
- 2

112°F is outside the original data range, making this an extrapolation. The model has no data to support a prediction at that temperature.

Extrapolation extends a model beyond the data it was built on. At 112°F, real-world behavior (heat advisories, park closures, changes in foot traffic) may make the linear trend completely invalid.
- 3

The slope of 5.2 is too steep for high temperatures, so it automatically produces inflated predictions above 100°F.

The slope doesn't change based on temperature range; it's a fixed value in the model. The real issue is that the model was never tested at 112°F, not that the slope behaves differently there.
- 4

Regression models can only be trusted within 10°F of the average temperature in the data set.

There is no fixed-degree rule like this. The concern is whether the prediction falls outside the data range entirely, which is what extrapolation describes.

Tracking Plant Growth

A researcher measures a sunflower seedling's height once a week. She collects data from **week 0 to week 25** and fits a regression model:

$y = 3.5x + 4$,

where x is the number of weeks since planting, and y is the predicted height in centimeters.



Which statement correctly interprets **both** the slope and the y-intercept of $y = 3.5x + 4$ in this context?

- Select one
- 1

For every additional week, the plant's predicted height increases by 3.5 cm; at week 0, the model predicts a starting height of 4 cm.

The slope (3.5) is the weekly growth rate, and the y-intercept (4) is the predicted height at week 0.
- 2

For every additional week, the plant's predicted height increases by 4 cm; at week 0, the model predicts a starting height of 3.5 cm.

The values are swapped: 3.5 is the rate of change (per week), and 4 is the starting height at week 0, not the other way around.
- 3

The plant grows to a maximum height of 3.5 cm; 4 is the number of weeks it takes to reach that height.

This describes neither value correctly. 3.5 is a weekly growth rate, not a maximum height, and 4 is a starting height, not the number of weeks.
- 4

The plant's total height after 25 weeks is 3.5 cm, and 4 is an unrelated rounding constant.

Neither value works this way. 3.5 is the slope (rate of change per week), and 4 is the y-intercept (starting height at week 0).

Predict a future value. Use the model to predict the seedling's height at **week 20**. Show the substitution and calculate the result.

Type answer

Discuss extrapolation limits. The researcher only collected data from week 0 to week 25. If someone used this model to predict the plant's height at **week 200**, the equation gives $y = 3.5(200) + 4 = 704$ cm (about 23 feet tall). Is this a reasonable prediction? Explain why or why not, using the idea of extrapolation.

Type answer