

# Simple Linear Regression Concepts

Supervised Learning

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# Contents of This Video

In this video, we will cover:

- Definition and purpose of simple linear regression
- The linear equation and its components
- Interpretation of slope and intercept coefficients
- Visual representation of the regression line
- The meaning of errors/residuals
- Using regression for prediction and inference
- Real-world applications and limitations

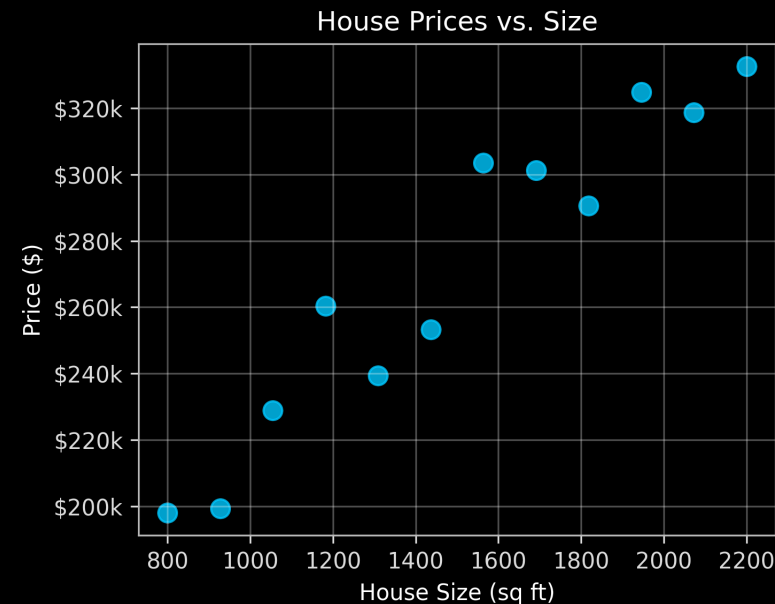
# What is Linear Regression?

- A supervised learning method for predicting continuous outcomes
- Models the relationship between:
  - Input variable(s) (features/predictors)
  - Output variable (response/target)
- **Simple** linear regression: 1 input  $\rightarrow$  1 output
- **Multiple** linear regression: multiple inputs  $\rightarrow$  1 output

# A Motivating Example: House Prices

**Can we predict house price from size?**

- Input (X): House size (square feet)
- Output (Y): House price (dollars)
- Each data point: One house sale
- Relationship: Generally, bigger houses → higher prices



# The Simple Linear Regression Model

The mathematical equation:

$$Y = \beta_0 + \beta_1 X + \varepsilon$$

Where:

- $Y$  is the target variable (e.g., house price)
- $X$  is the input feature (e.g., house size)
- $\beta_0$  is the intercept ( $Y$ -value when  $X = 0$ )
- $\beta_1$  is the slope (change in  $Y$  for 1-unit increase in  $X$ )
- $\varepsilon$  (epsilon) is the error term

# Understanding the Line: Visualizing $\beta_0$ and $\beta_1$

# Interpreting the Coefficients

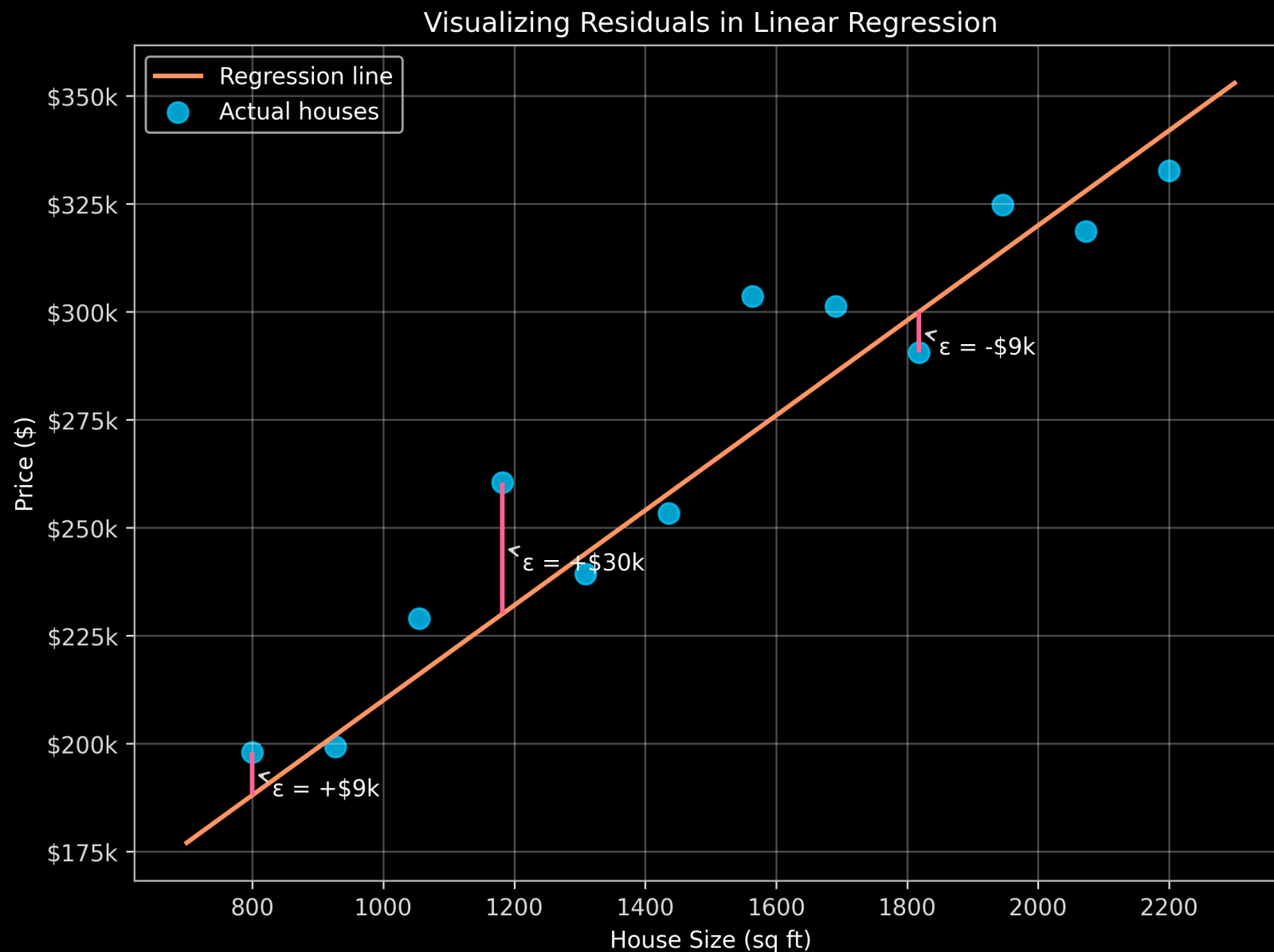
## Intercept ( $\beta_0$ )

- Value of Y when  $X = 0$
- In our example:
  - $\beta_0 = \$100,000$
  - Baseline house price when size = 0
  - Often not literally meaningful
  - Sets the vertical position of the line

## Slope ( $\beta_1$ )

- Change in Y for a 1-unit increase in X
- In our example:
  - $\beta_1 = \$110$  per square foot
  - Each extra sq ft adds \$110 to price
  - Positive slope:  $X \uparrow \rightarrow Y \uparrow$
  - Negative slope:  $X \uparrow \rightarrow Y \downarrow$
  - Zero slope: X has no effect on Y

# Understanding the Error Term ( $\epsilon$ )





# Using the Model for Prediction

If our fitted model is:

$$\text{Price} = \$100,000 + \$110 \times \text{Size}$$

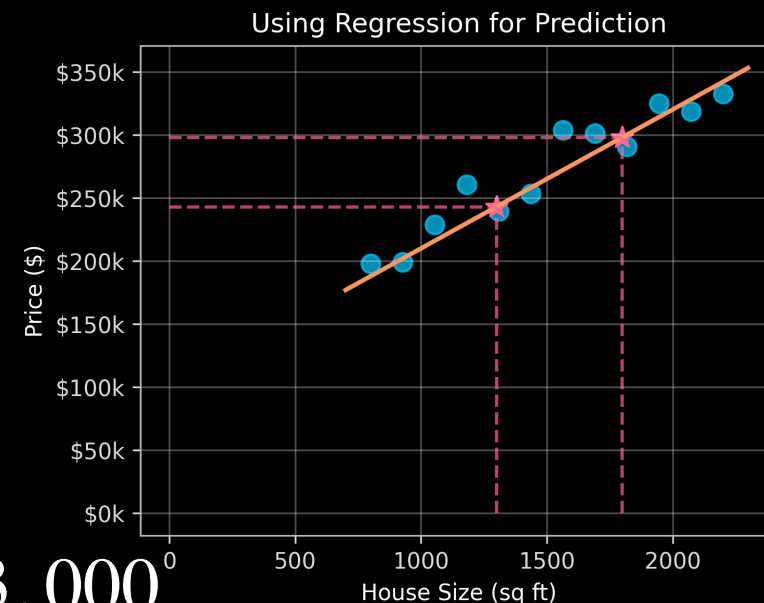
We can predict the price of a new house:

For a 1,300 sq ft house:

$$\text{Price} = \$100,000 + \$110 \times 1,300 = \$243,000$$

For a 1,800 sq ft house:

$$\text{Price} = \$100,000 + \$110 \times 1,800 = \$298,000$$



# Using the Model for Inference

Linear regression provides insights about relationships:

- **Direction:** Is the relationship positive or negative?
  - $\beta_1 > 0$ : X and Y move in same direction
  - $\beta_1 < 0$ : X and Y move in opposite directions
- **Magnitude:** How strong is the effect?
  - $|\beta_1|$  large: X has a strong effect on Y
  - $|\beta_1|$  small: X has a weak effect on Y
- **Practical meaning:** What does  $\beta_1$  tell us about our domain?
  - In our example: Each square foot adds about \$110 to home value

# Applications Beyond House Prices

**Linear regression is used for many prediction tasks:**

- **Finance:** Predicting stock returns based on economic indicators
- **Healthcare:** Estimating patient recovery time based on treatment dosage
- **Education:** Predicting student test scores from study hours
- **Marketing:** Forecasting sales based on advertising spend
- **Environmental science:** Modeling temperature changes over time

All follow the same principle: find the best-fitting line to describe the relationship between  $X$  and  $Y$ .

# Limitations of Simple Linear Regression

## Assumes a linear relationship

- Real data may have nonlinear patterns
- Can miss complex relationships
- May oversimplify reality

## Only uses one predictor

- Most real-world outcomes depend on multiple factors
- House prices depend on more than just size

## Limited predictive power

- Single feature → more unexplained variation

## Won't capture interactions

- When the effect of one variable depends on another

## Assumes constant variance

- Error may vary across the range of  $X$

# What We've Covered

In this video, we've learned:

- The basic concept and equation of simple linear regression
- How to interpret the intercept ( $\beta_0$ ) and slope ( $\beta_1$ ) coefficients
- The meaning of the error term ( $\epsilon$ )
- How to visualize a regression line and its components
- Using regression for both prediction and inference
- Limitations of the simple linear model