

# CORRELATION AND LINEAR REGRESSION

- Scatterplots
- Covariance and Correlation
- Linear Regression
- Residuals
- R-Squared
- Types of Observations



## SCATTERPLOTS

- **Scatterplot:** graphical display of the relationship between two quantitative variables
  - **Response variable:** variable plotted along y-axis that we are trying to explain or predict
  - **Predictor variable:** variable plotted along x-axis that we are using to explain changes about the response variable
  - Observations plotted as ordered pairs

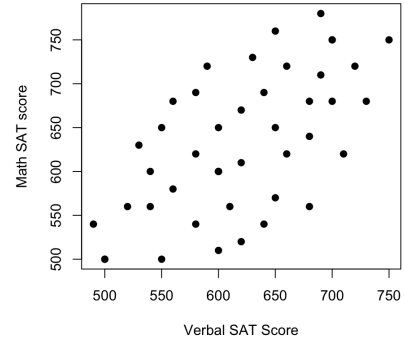
## DESCRIBING A SCATTERPLOT

- **Direction:** As predictor variable increases...
  - **Positive:** response tends to increase
  - **Negative:** response tends to decrease
  - **Neither:** no obvious change in response
- **Form:** What is the general trend of the points?
  - **Linear:** response tends to increase at about the same rate across all values of predictor
  - **Curved:** rate at which response changes depends on value of predictor
  - **No pattern**
- **Strength:** How tightly clustered together are the points?
  - Usually described as **strong**, **moderate**, or **weak**

## EXAMPLE: DESCRIBING A SCATTERPLOT

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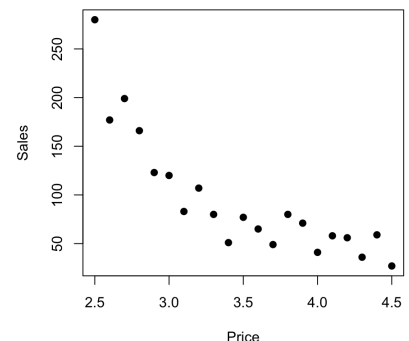
- **Scenario:** Want to know if verbal SAT scores yields any information about math SAT scores using a sample of 44 high school seniors
- **Question:** What are the predictor and response variables.
- **Answer:**
  - Response: \_\_\_\_\_ SAT score
  - Predictor: \_\_\_\_\_ SAT score
- **Task:** Describe the relationship.
- **Answer:** \_\_\_\_\_, \_\_\_\_\_, and \_\_\_\_\_
  - Math scores tend to \_\_\_\_\_ as verbal scores increase
  - Math scores change at \_\_\_\_\_ regardless of verbal score



## EXAMPLE: DESCRIBING A SCATTERPLOT

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- **Scenario:** Supermarket increases the price for a gallon a milk by 10 cents every day for 3 weeks and records number of units sold.
- **Task:** Describe the relationship.
- **Answer:** \_\_\_\_\_, \_\_\_\_\_, and \_\_\_\_\_
  - Sales tend to \_\_\_\_\_ as price increases
  - Not \_\_\_\_\_
    - Sales decline \_\_\_\_\_ when \_\_\_\_\_ prices are increased compared to when \_\_\_\_\_ prices are increased



## COVARIANCE

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- **Covariance:** measure of the joint variability between two quantitative variables

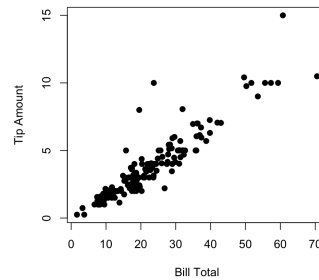
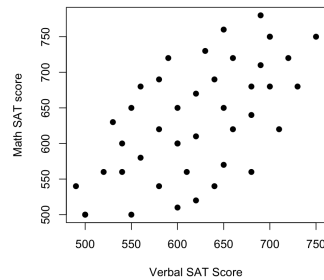
$$s_{XY} = \frac{(x_1 - \bar{x})(y_1 - \bar{y}) + \cdots + (x_n - \bar{x})(y_n - \bar{y})}{n - 1}$$

- Sign of covariance dictates direction of relationship
- Value is unbounded: Ranges from  $-\infty$  to  $\infty$
- **Problem:** Does not help us interpret the \_\_\_\_\_ of relationship

## EXAMPLE: COMPARING RELATIONSHIPS

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- **Scenario:** Two scatterplots shown below:
  - **Left:** Use verbal SAT score to explain math SAT score
  - **Right:** Use restaurant bill to explain amount left as tip
- **Question:** Which scatterplot has the stronger linear relationship?



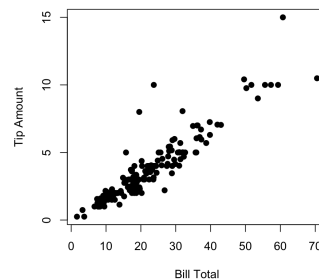
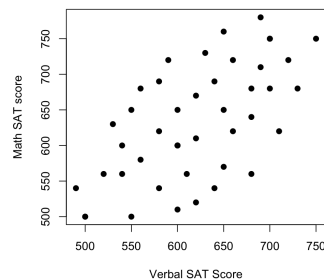
- **Answer:** \_\_\_\_\_ vs. \_\_\_\_\_
  - Points are more \_\_\_\_\_ around \_\_\_\_\_

## EXAMPLE: UNDERSTANDING COVARIANCE

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- **Scenario:** Two scatterplots shown below:
  - **Left:** Use verbal SAT score to explain math SAT score
  - **Right:** Use restaurant bill to explain amount left as tip
- **Question:** What does the covariance reveal?

$$s_{XY} = 3331$$



$$s_{XY} = 26.93$$

- **Answer:** \_\_\_\_\_ by itself
  - Covariance will be \_\_\_\_\_ if the \_\_\_\_\_ of the observations are large regardless of how \_\_\_\_\_ the linear relationship is

## CORRELATION

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- **Correlation:** measure of the strength and direction of the linear relationship between two quantitative variables
  - **Population correlation:** Denoted by  $\rho$  (Greek letter "rho")
    - Parameter  $\rightarrow$  Generally unknown
  - **Sample correlation:** Denoted by  $r$ 
    - Statistic  $\rightarrow$  Good approximation of  $\rho$

$$r = \frac{s_{XY}}{s_X s_Y}$$

Standard deviation of predictor values

Standard deviation of response values

# CORRELATION FACTS

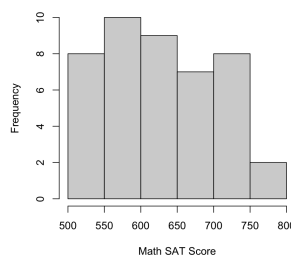
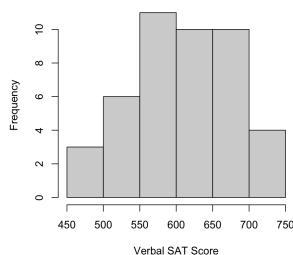
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- Bounded between -1 and +1
  - Sign dictates direction of relationship (**positive** or **negative**)
  - Magnitude dictates strength of relationship
    - **Rule of Thumb:** If the magnitude is...
      - Between 0.70 and 1.00, the strength of the relationship is **strong**.
      - Between 0.40 and 0.70, the strength of the relationship is **moderate**.
      - Between 0.10 and 0.40, the strength of the relationship is **weak**.
      - Between 0.00 and 0.10, there is **little to no** relationship.
- Scatterplot must have a **linear** form for the correlation to make sense.
  - As the scatterplot becomes more curved, the correlation becomes less accurate as a means of describing the relationship.

## EXAMPLE: CALCULATING CORRELATION

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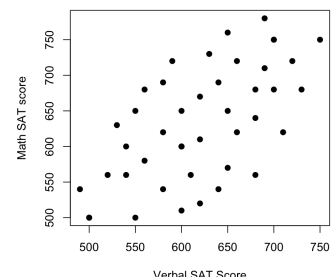
- **Scenario:** Use verbal SAT score to describe math SAT score for a random sample of 44 students



Variable	Mean	Std. Dev.	Covariance
Verbal	619.32	67.35	3331
Math	631.59	79.94	

- **Question:** What is the correlation?

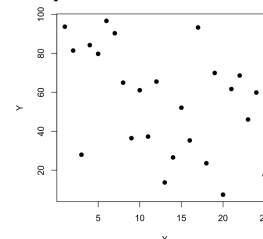
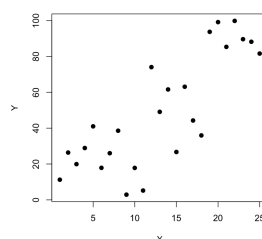
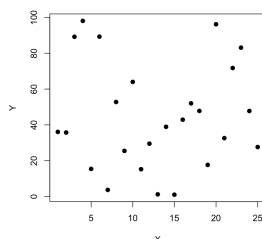
- **Answer:**  $r = \underline{\hspace{2cm}} = \underline{\hspace{2cm}}$ 
  - Indicates                                  linear relationship



## EXAMPLE: APPROXIMATING CORRELATIONS

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- **Scenario:** Scatterplots show observations with predictor values from 1 to 25 and responses from 0 to 100.
- **Task:** Sort the scatterplots from weakest correlation to strongest.
- **Answer:**     ,     ,
- **Task:** Approximate the correlation in each scatterplot.



- **Answers:**  $r = \underline{\hspace{2cm}}$        $r = \underline{\hspace{2cm}}$        $r = \underline{\hspace{2cm}}$

## SIMPLE LINEAR REGRESSION LINE

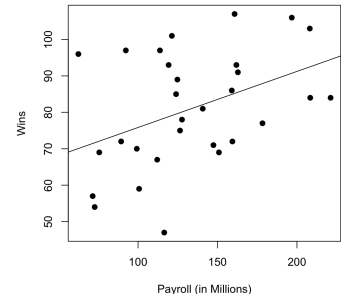
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- **Simple linear regression line:** the best fitting line between a single quantitative predictor ( $X$ ) and a quantitative response ( $Y$ )

$$\hat{Y} = \hat{\beta}_0 + \hat{\beta}_1 X$$

Predicted Value of Response: "Y-hat"      Intercept      Slope      Value of predictor

- Describes how  $Y$  tends to change as  $X$  increases
- Allows us to make predictions about the response given some value of the predictor
- Predictions are not perfect, but provide good estimates



## SIMPLE LINEAR REGRESSION LINE

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- **Simple linear regression line:** the best fitting line between a single quantitative predictor ( $X$ ) and a quantitative response ( $Y$ )

$$\hat{Y} = \hat{\beta}_0 + \hat{\beta}_1 X$$

Predicted Value of Response: "Y-hat"      Intercept      Slope      Value of predictor

- **Slope:**  $\hat{\beta}_1 = r \frac{s_Y}{s_X}$ 
  - Expected increase in prediction given a one unit increase in the predictor
- **Intercept:**  $\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$ 
  - Predicted value of response variable when predictor equals 0

## EXAMPLE: CALCULATING REGRESSION LINE EQUATION

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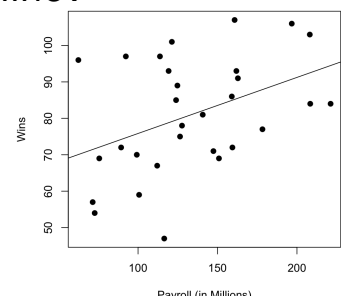
- **Scenario:** Use payroll (in millions) to predict wins in Major League Baseball in 2019

Variable	Mean	Std. Dev.	Correlation
Payroll	133.47	42.66	0.4135
Wins	81	15.87	

- **Question:** What is the equation of the regression line?

- **Answer:**

- **Slope:**  $\hat{\beta}_1 = \underline{\hspace{2cm}} = \underline{\hspace{2cm}}$
- **Intercept:**  $\hat{\beta}_0 = \underline{\hspace{2cm}} = \underline{\hspace{2cm}}$
- **Regression Line:**  $\underline{\hspace{4cm}}$



## EXAMPLE: INTERPRETING REGRESSION EQUATION

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- **Scenario:** Use payroll (in millions) to predict wins in Major League Baseball in 2019

$$\hat{Y} = 60.47 + 0.1538X$$

- **Question:** What are the interpretations of the slope and intercept?

- **Answer:**

- **Intercept:** A team with a \_\_\_\_\_ would be \_\_\_\_\_ games.
- **Slope:** For every additional \_\_\_\_\_ a team spends on its payroll, their predicted win total \_\_\_\_\_.

## EXAMPLE: MAKING PREDICTIONS

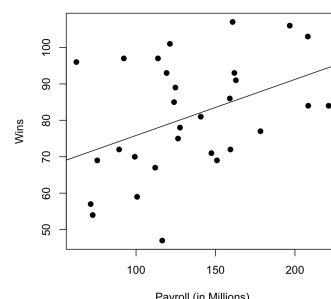
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- **Scenario:** Use payroll (in millions) to predict wins in Major League Baseball in 2019

$$\hat{Y} = 60.47 + 0.1538X$$

- **Question:** How many games would we expect a team with a payroll of \$100 million to win?

- **Answer:**  $\hat{Y} =$  \_\_\_\_\_  
= \_\_\_\_\_  
= \_\_\_\_\_



## RESIDUALS

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- **Residual:** the difference between the predicted value of a response and the observed value from the data

$$e = Y - \hat{Y}$$

Residual (Error)      Observed Value From Data      Predicted Value From Regression Line

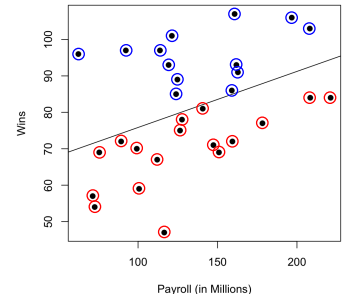
- **Observations that are:**

- **Close** to the regression line will have \_\_\_\_\_ residuals
- **Far away** from the regression line will have \_\_\_\_\_ residuals
- **Above** the regression line will have \_\_\_\_\_ residuals
- **Below** the regression line will have \_\_\_\_\_ residuals

## EXAMPLE: RESIDUALS

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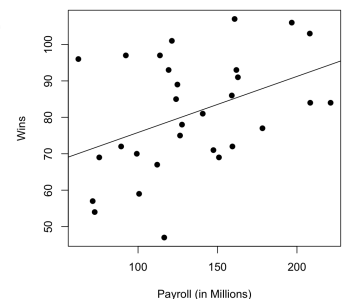
- **Scenario:** Use payroll (in millions) to predict wins in Major League Baseball in 2019
- **Question:** What can be said about the blue circled observations?
- **Answer:** All have \_\_\_\_\_ residuals
  - Actual win totals were \_\_\_\_\_ than their \_\_\_\_\_ win total
  - \_\_\_\_\_ for the season based on payroll alone
- **Question:** What can be said about the red circled observations?
- **Answer:** All have \_\_\_\_\_ residuals
  - Actual win totals were \_\_\_\_\_ than predicted win total



## EXAMPLE: RESIDUALS

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- **Scenario:** Kansas City Royals had a \$100 million payroll in 2019 and won 59 games. (Recall the predicted value was 75.85.)
- **Question:** What was their residual?
- **Answer:**  $e = \text{_____} = \text{_____}$ 
  - Observation lies \_\_\_\_\_ the regression line
- **Question:** What does this value mean in context?
- **Answer:** Kansas City \_\_\_\_\_ during the season, winning \_\_\_\_\_ than would have been expected given its \_\_\_\_\_ payroll.



## VARIATION IN THE LINEAR MODEL

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- **R-squared:** the percentage of the variability in the response ( $Y$ ) that is explained by the predictor ( $X$ )
  - Calculated by squaring the correlation:  $r^2$ 
    - Observations in straight line if  $r^2 = 1 \rightarrow$  Perfect model
    - No relationship between variables if  $r^2 = 0 \rightarrow$  Useless model
  - Higher R-squared indicates a stronger relationship between  $X$  and  $Y$
  - Remainder of variation ( $1 - r^2$ ) is due to natural fluctuations of the response in the sample and comes from the residuals.

## EXAMPLE: R-SQUARED

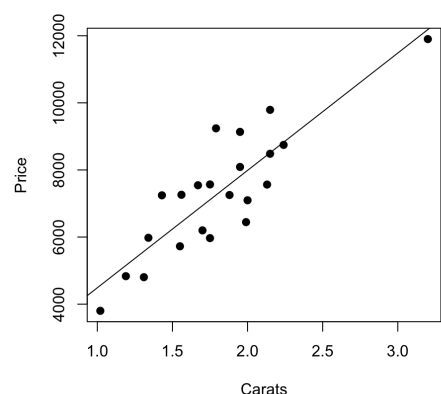
- **Scenario:** Use payroll (in millions) to predict wins in Major League Baseball in 2019. Correlation is 0.4135.
- **Question:** What is the R-squared value?
- **Answer:**  $r^2 = \underline{\hspace{2cm}} = \underline{\hspace{2cm}}$
- **Question:** What does the R-squared mean in context?
- **Answer:**  $\underline{\hspace{2cm}}$  of the  $\underline{\hspace{2cm}}$  is explained by their  $\underline{\hspace{2cm}}$ .
  - Remaining  $\underline{\hspace{2cm}}$  is due to  $\underline{\hspace{2cm}}$  in the sample, such as:
    - Talented younger,  $\underline{\hspace{2cm}}$  players
    - Aging,  $\underline{\hspace{2cm}}$  veterans who underperform
    - Randomness within  $\underline{\hspace{2cm}}$

## TYPES OF POINTS IN REGRESSION

- **Outlier:** an observation with an unusually large residual
  - Extreme values of the response, but can take on any predictor value
- **High leverage observation:** an observation where the value of the predictor is unusually far away from the rest of the data
  - Extreme values of the predictor, but can take on any response
  - Have the *potential* to drastically impact the regression line
- **Influential point:** an observation that strongly influences the slope and/or direction of the regression line
  - Extreme values of both the predictor and response

## TYPES OF POINTS IN REGRESSION

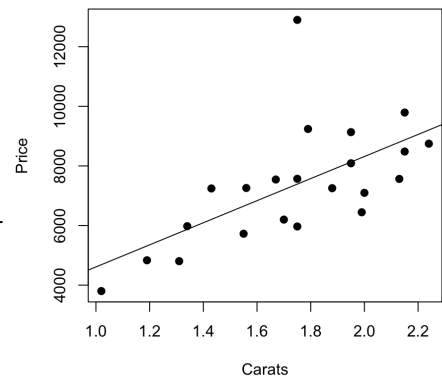
- **Scenario:** Random sample of 22 clear diamonds, one of which was 3.16 carats and sold for \$11,900.
- **Question:** How should we classify the diamond with the weight of 3.16 carats?
- **Answer:**  $\underline{\hspace{2cm}}$ 
  - **Outlier?:**  $\underline{\hspace{2cm}}$ 
    - Residual is  $\underline{\hspace{2cm}}$
  - **High Leverage?:**  $\underline{\hspace{2cm}}$ 
    - Predictor value is  $\underline{\hspace{2cm}}$  from the rest of data
  - **Influential Point?:**  $\underline{\hspace{2cm}}$ 
    - Observation follows  $\underline{\hspace{2cm}}$  of data; removal would  $\underline{\hspace{2cm}}$  regression line by much





## TYPES OF POINTS IN REGRESSION

- **Scenario:** Random sample of 21 clear diamonds and one blue diamond that was 1.75 carats but sold for \$12,900.
- **Question:** How should we classify the blue diamond?
- **Answer:** \_\_\_\_\_
  - **Outlier?:** \_\_\_\_\_
    - \_\_\_\_\_ residual
  - **High Leverage?:** \_\_\_\_\_
    - \_\_\_\_\_ value of the predictor
  - **Influential Point?:** \_\_\_\_\_
    - Slope of regression line not \_\_\_\_\_ impacted – got \_\_\_\_\_ slightly



## TYPES OF POINTS IN REGRESSION

- **Scenario:** Random sample of 21 clear diamonds and one black diamond that was 2.90 carats but sold for \$3,200.
- **Question:** How should we classify the black diamond?
- **Answer:** \_\_\_\_\_
  - **Outlier?:** \_\_\_\_\_
    - \_\_\_\_\_ residual
  - **High Leverage?:** \_\_\_\_\_
    - \_\_\_\_\_ predictor value
  - **Influential Point?:** \_\_\_\_\_
    - Drastically \_\_\_\_\_ regression line, creating a slope that is closer to \_\_\_\_\_

