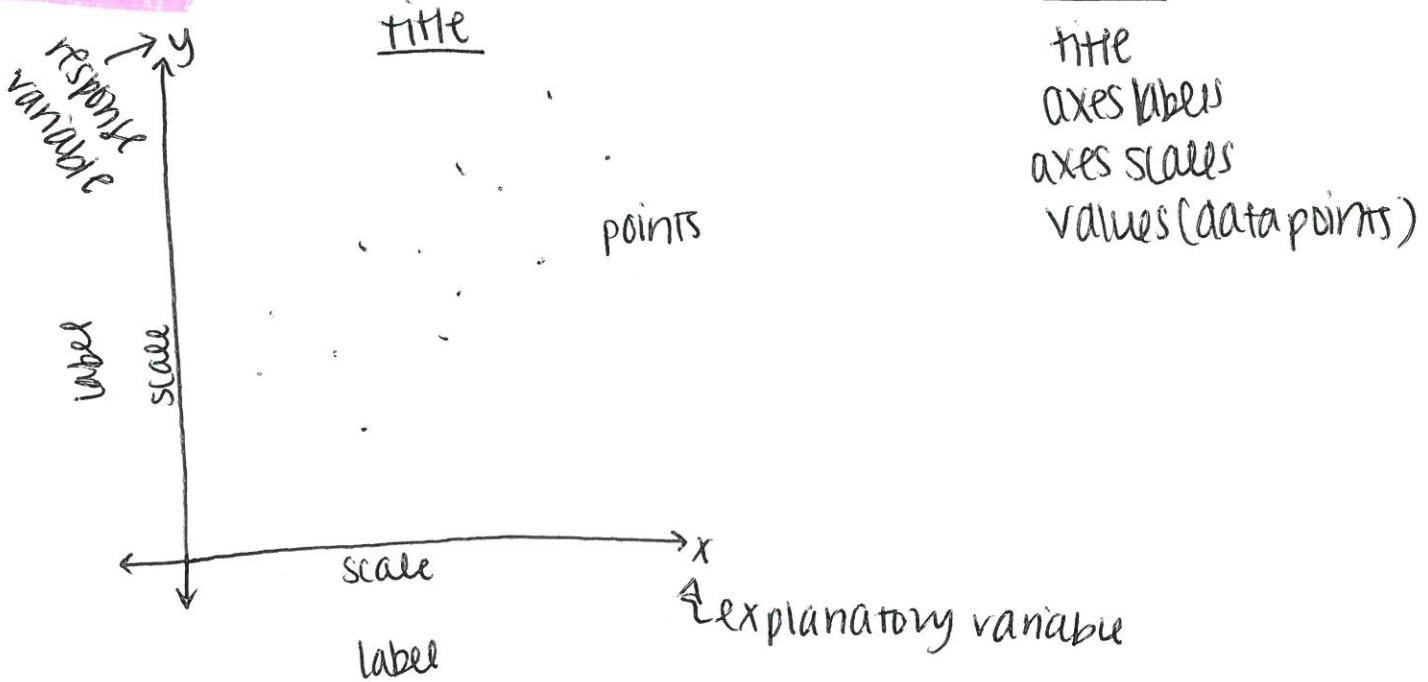


Scatterplots:



Describe using:

**Form:** clusters, gaps, outliers, linear or curved

**Direction:** positive slope vs. negative slope

**Strength:** strong, moderate, weak (how tight the points are in a line/curve)

**Correlation:** measures the strength of the linear relationship between  $x$  and  $y$ .

**Correlation Coefficient ( $r$ ):** same as correlation, but given as a numerical value.  $-1 \leq r \leq +1$ . As  $|r|$  approaches 1, the correlation increases.

**Coefficient of Determination ( $R^2$ ):** sentence frame! the % variation in  $y$  explained by  $x$  (or the LSRL of  $y$  on  $x$ ). becomes stronger).

LSRL: least squares regression line.

the line that gets closest to most data points.

it best fits the data.

$$\hat{y} = a + bx$$

*Y-hat* ← predicted response variable  
a ← y-intercept  
b ← slope  
x ← explanatory variable

Three ways to find the LSRL equation:

**METHOD #1:** Using a list of data points:

1. Put data in L1 and L2 of your calculator
2. STAT > CALC >  
8. LinReg(a+bx)
3. Place a and b values into your equation and be sure to write your equation with Y-HAT and IN CONTEXT.

**METHOD #2:** Using calculated values for mean, standard deviation, and r.

1. Use the equations on the equation sheet to calculate a and b.
2. Place a and b values into your equation and be sure to write your equation with Y-HAT and IN CONTEXT.

**METHOD #3:** Using a MiniTab Output:

1. Identify a and b. The value of a can be found in the Constant row, Coef column. The value of b can be found in the Variable row, Coef column.
2. Place a and b values into your equation and be sure to write your equation with Y-HAT and IN CONTEXT.

**METHOD #1 EXAMPLE:** Create the LSRL for predicting number of bags of dog food used per month based on the number of dogs a person owns. Use the data table below.

# of dogs	Bags of dog food used/month
1	3
2	6
3	8
2	7
3	10
4	11
5	15
5	15
4	12
3	9
6	17
9	25
8	24
7	20

$$\text{Dogs of dog food} = 0.636 + 2.792(\# \text{ of dogs})$$

OR

$$\hat{y} = 0.636 + 2.792(x)$$

$\hat{y}$  = predicted # of bags of dog food used per month

x = # of dogs owned

If you define y instead of  $\hat{y}$ , omit "predicted".

**METHOD #2 EXAMPLE:** Create the LSRL for predicting number of bags of dog food used per month based on the number of dogs a person owns. Use the values below.

$$\bar{x} = 4.429 \\ s_x = 2.377 \\ \bar{y} = 13 \\ s_y = 6.668 \\ r = 0.9951$$

$$b = r \frac{s_y}{s_x} = 0.9951 \cdot \frac{6.668}{2.377} = 2.7915$$

$$a = \bar{y} - b\bar{x}$$

$$a = 13 - 2.7915(4.429) = 0.6366$$

# of bags of dog food =  $0.6366 + 2.7915(\# \text{ of dogs})$

Or... define  $x$  and  $y$  or  $\hat{y}$  and write the equation of the LSRL w/ variables ( $x$  and  $\hat{y}$ ).

**METHOD #3 EXAMPLE:** Create the LSRL for predicting number of bags of dog food used per month based on the number of dogs a person owns. Use the MiniTab Output below.

Predictor	Coef	SE Coef	T	P
Constant	0.6362	whatevs	whatevs	whatevs
Bags	2.7918	whatevs	whatevs	whatevs
S = whatevs	R-Sq = 0.9902		R-Sq(adj) = whatevs	

this is the variable  $x$   
so slope goes next to it

to find  $r$ , take  $\sqrt{R^2}$  and use the same sign as slope ( $\pm$ )

we don't use this

# of bags of dog food =  $0.6362 + 2.7918(\# \text{ of dogs})$

OR...

Other important things:

### Calculating & Interpreting Residuals:

residual = observed value - predicted value

EXAMPLE: Calculate and interpret the residual value for a person who owns 7 dogs.

Find the actual value (from table):  $(7, 20)$

$$y = 20 \text{ bags of dog food}$$

# of dogs at dog food

Find the predicted value (from LSRL):  $(7, 20.179)$

$$\hat{y} = 0.1636 + 2.792(7) = 20.179$$

predicted # of bags of dog food

### Residual Plot:

usually  $x$  vs. residuals  
(for each point)

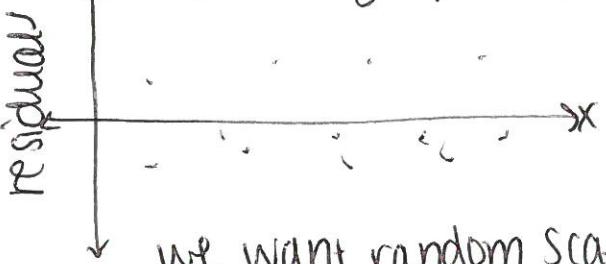
$$\text{residual} = y - \hat{y}$$

\* compare to graph of  $L_1$  vs.  $L_2$

\* do linreg on calc to plot residuals!

$$= 20 - 20.179 = -0.179$$

A person who owns 7 dogs is predicted to need 0.179 more bags of dog food per month than they actually need.



We want random scatter if the LSRL is a good fit.

### Influential Point:

if removed, the LSRL drastically changes. Could be an outlier ( $x$ -direction)

or not!

Extrapolation: When we use the model to predict  $y$  for values of  $x$  not used to create the model. BAD! in the domain

OR A person who owns 7 dogs needs 0.179 bags of dog food per month less than they are predicted to need.

EXAMPLE: Would it make sense to use our model to predict the number of bags of dog food needed for a person who has 49 dogs? Why or why not?

Nope! This is extrapolation. 49 is not in the domain of  $x$  values used to create this model.

## Interpreting slope and y-intercept:

Slope can be interpreted as follows: for every [one unit increase] in [x variable in context], we predict [y in context] will increase/decrease by [slope value in context w/ units].

y-int: If [x in context] is 0 (or there are none, in context), the value of [y in context] is, [yint value

### EXAMPLE:

Slope:

For every one additional dog owned by a person, the number of bags of dog food they need per month is predicted to increase by 2.792 bags.

y-intercept: A person who owns no dogs is expected/predicted to need 0.6366 bags of dog food per month.

**Example:** describe the scatterplot.

F there is a moderately strong, positive, linear relationship between # of dogs owned and # of bags of dog food needed per month.  
D  
S

