Lecture 2 Simple Linear Regression

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Learning Objectives

- ☐ How to load data from a text file
- ☐ How to visualize data via a scatter plot
- Describe a linear model for data
 - Identify the target variable and predictor
- □ Compute optimal parameters for the model using the regression formula
- ☐ Fit parameters for related models by minimizing the residual sum of squares
- \square Compute the R^2 measure of fit
- □ Visually determine goodness of fit and identify different causes for poor fit



Outline

Motivating Example: Predicting the mpg of a car

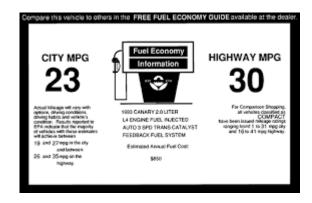
- ☐ Linear Model
- ☐ Least Squares Fit Problem
- ☐ Sample Mean and Variance; LS Fit Solution
- ☐ Assessing Goodness of Fit





Example: What Determines mpg in a Car?

- ☐ What engine characteristics determine fuel efficiency?
- ■Why would a data scientist be hired to answer this question?
- Not to help purchasing a specific car.
 - The mpg for a currently available car is already known.
 - (If the car company isn't lying?)
- ☐ To guide building new cars.
 - Understand what is reasonably achievable before full design
- ☐ To find cars that are outside the trend.
 - Example: What cars give great mpg for the cost or size?



Demo in Github

https://github.com/sdrangan/introml/blob/master/simp_lin_reg/auto_mpg.ipynb

Simple Linear Regression for Automobile mpg Data

In this demo, you will see how to:

- . Load data from a text file using the pandas package
- · Create a scatter plot of data
- · Handle missing data
- · Fit a simple linear model
- Plot the linear fit with the test data
- Use a nonlinear transformation for an improved fit

Loading the Data

The python pandas library is a powerful package for data analysis. In this course, we will use a small portion of its features -- just reading and writing data from files. After reading the data, we will convert it to numpy for all numerical processing including running machine learning algorithms.

We begin by loading the packages.

```
In [86]: import pandas as pd
         import numpy as np
```

The data for this demo comes from a survey of cars to determine the relation of mpg to engine characteristics. The data can be found in the UCI library: https://archive.ics.uci.edu/ml/machine-learning-databases/auto-mpg

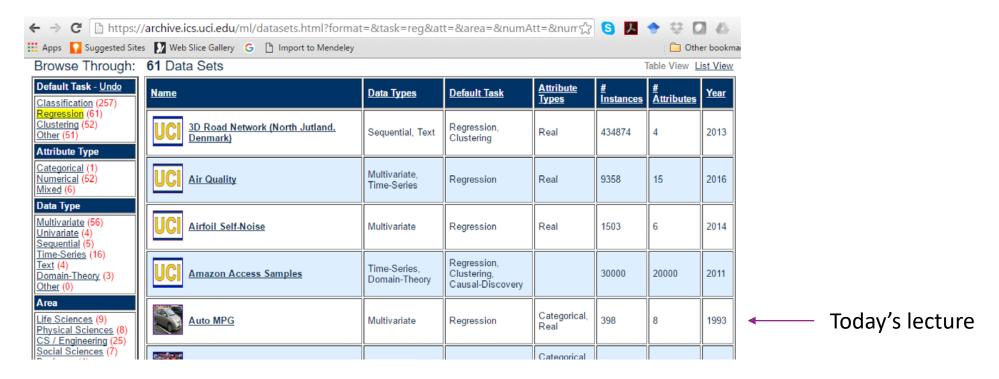
You can directly read the data in the file, https://archive.ics.uci.edu/ml/machine-learning-databases/auto-mpg/auto-mpg.data We will load the data into ipython notebook, using the pandas library. Unfortunately, the file header does not include the names of the fields,





Getting Data

□ Data from UCI dataset library: https://archive.ics.uci.edu/ml/datasets.html





Loading the Data in Jupyter Notebook Try 1: The Wrong Way!

```
☐ Python pandas library
        import pandas as pd

    Read csv command.

        import numpy as np

    Read URL or file location.

 In [67]: names = ['mpg', 'cylinders', 'displacement', 'horsepower',
                                                                                                 ☐ Creates a dataframe object
                 'weight', 'acceleration', 'model year', 'origin', 'car name']

    http://pandas.pydata.org/pandas-

In [122]: | df = pd.read_csv('https://archive.ics.uci.edu/ml/machine-learning-databases/auto-mpg/auto-mpg.data')
                                                                                                       docs/stable/dsintro.html#dataframe
In [123]: df.head(6)
Out[123]:
           18.0 8 307.0 130.0 3504, 12.0 70 1 "chevrolet chevelle malibu
                                                                                                 □ Problems
         0 15.0 8 350.0 165.0 3693. 11...
           18.0 8 318.0 150.0 3436. 11...
                                                                                                 □ Does not parse columns
         2 16.0 8 304.0 150.0 3433. 12...
         3 17.0 8 302.0 140.0 3449. 10...

    All data in a single column

         4 15.0 8 429.0 198.0 4341. 10..

    Read csv assumes columns are delimited

         5 14.0 8 454.0 220.0 4354. 9...
                                                                                                       by commas
```





☐ Mistakes first line as header

Loading the Data in Jupyter Try 2: Fixing the Errors

You can display a first few lines of the dataframe by using head command:

In [126]: df.head(6)

Out[126]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18	8	307	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15	8	350	165	3693	11.5	70	1	buick skylark 320
2	18	8	318	150	3436	11.0	70	1	plymouth satellite
3	16	8	304	150	3433	12.0	70	1	amc rebel sst
4	17	8	302	140	3449	10.5	70	1	ford torino
5	15	8	429	198	4341	10.0	70	1	ford galaxie 500

- ☐ Fix the arguments in read_csv
- ☐ Pandas routines have many options
- □When you get a problem:
 - Google is your friend!
 - You are not the first to have these problems.
- □ Dataframe has three components
 - df.columns, df.index, df.values
- More in recitation





Visualizing the Data

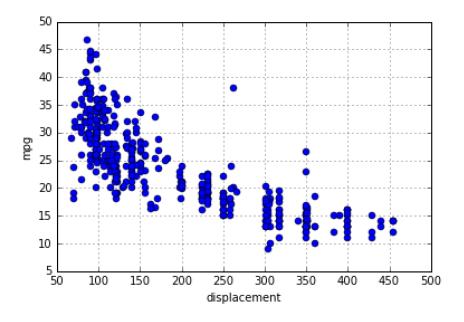
```
In [150]: xstr = 'displacement'
    x = np.array(df[xstr])
    y = np.array(df['mpg'])

In [146]: import matplotlib
    import matplotlib.pyplot as plt
    %matplotlib inline

In [151]: plt.plot(x,y,'o')
    plt.xlabel(xstr)|
    plt.ylabel('mpg')
    plt.grid(True)
```

EΘ

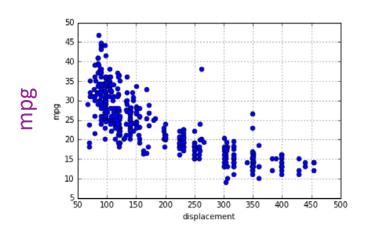
- ☐When possible, look at data before doing anything
- ☐ Python has MATLAB-like plotting
 - Matplotlib module

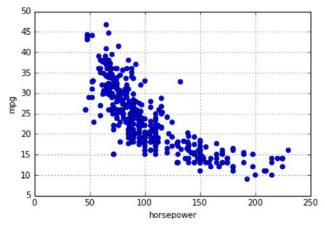


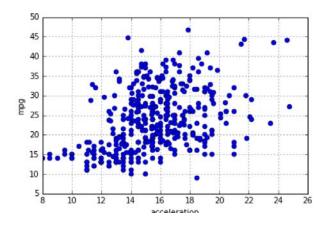


Postulating a Model

- ☐ What relationships do you see?
- □ Is there a mathematical model relating the variables?
- ☐ How well can you predict mpg from these variables?







Displacement

Horsepower

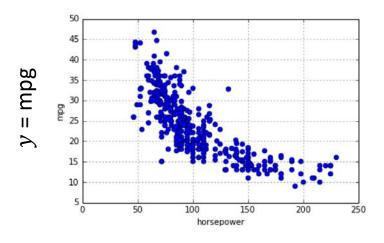
Acceleration

Outline

- ☐ Motivating Example: Predicting the mpg of a car
- Linear Model
- ☐ Least Squares Fit Problem
- ☐ Sample Mean and Variance; LS Fit Solution
- ☐ Assessing Goodness of Fit

Data

- $\Box y$ = variable you are trying to predict.
 - Called many names: Dependent variable, response variable, target, regressand, ...
- $\Box x$ = what you are using to predict:
 - Predictor, attribute, covariate, regressor, ...
- \square Data: Set of points, (x_i, y_i) , i = 1, ..., n
 - Each data point is called a sample.
- ☐Scatter plot



x = horsepower

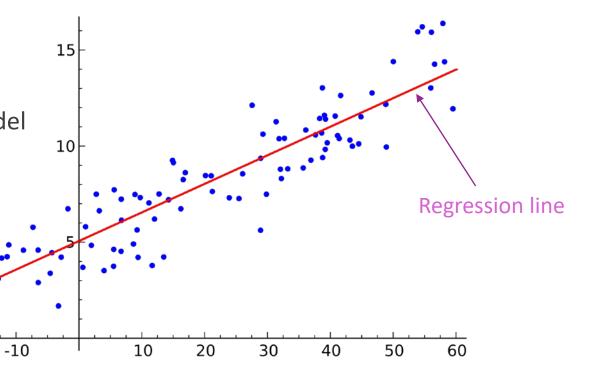
Linear Model

☐ Assume a linear relation

$$y \approx \beta_0 + \beta_1 x$$

-20

- $\beta_0 = \text{intercept}$
- \circ β_1 = slope
- $\Box \beta = (\beta_0, \beta_1)$ are the parameters of the model
- \square What are the units of β_0 , β_1 ?
- ☐When is this model good?



Why Use a Linear Model?

- ☐ Many natural phenomena have linear relationship
- ☐ Predictor has small variation
 - Suppose y = f(x)
 - If variation of x is small around some value x_0 , then

$$y \approx f(x_0) + f'(x_0)(x - x_0) = \beta_0 + \beta_1 x$$

$$\beta_0 = f(x_0) - f'(x_0)x_0, \qquad \beta_1 = f'(x_0)$$

- ☐ Gaussian random variables
- ☐ Simple to compute
- ☐ Easy to interpret relation

Outline

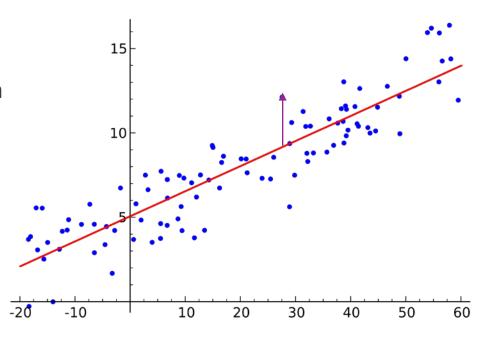
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Linear Model Residual

- \square Knowing x does not exactly predict y
- ☐ Add a residual term

$$y = \beta_0 + \beta_1 x + \epsilon$$

- ☐ Residual = component the model does not explain
 - Predicted value: $\hat{y}_i = \beta_1 x_i + \beta_0$
 - Residual: $\epsilon_i = y_i \hat{y}_i$
- □ Vertical deviation from the regression line



Least Squares Model Fitting

- □ How do we select parameters $\beta = (\beta_0, \beta_1)$?
- $\Box \text{ Define } \hat{y}_i = \beta_1 x_i + \beta_0$
 - Predicted value on sample *i* for parameters $\beta = (\beta_0, \beta_1)$
- ☐ Define average residual sum of squares:

RSS
$$(\beta_0, \beta_1) := \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$

- Note that \hat{y}_i is implicitly a function of $\beta = (\beta_0, \beta_1)$
- Also called the sum of squared residuals (SSR) and sum of squared errors (SSE)
- \square Least squares solution: Find (β_0, β_1) to minimize RSS.
 - Geometrically, minimizes squared distances of samples to regression line

Finding Parameters via Optimization A general ML recipe

General ML problem

- ☐ Find a model with parameters
- ☐Get data
- □ Pick a loss function
 - Measures goodness of fit model to data
 - Function of the parameters

Simple linear regression

Linear model: $\hat{y} = \beta_0 + \beta_1 x$

Data: $(x_i, y_i), i = 1, 2, ..., N$

Loss function:

$$RSS(\beta_0, \beta_1) := \sum (y_i - \beta_0 + \beta_1 x_i)^2$$

 \square Find parameters that minimizes loss \longrightarrow Select β_0, β_1 to minimize $RSS(\beta_0, \beta_1)$



Outline

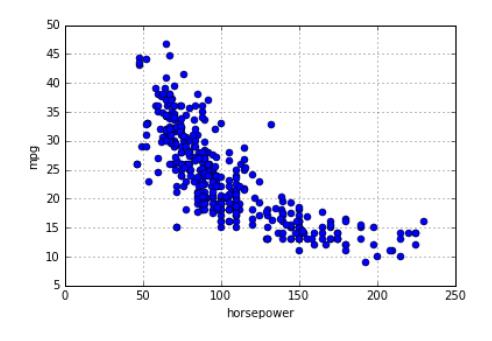
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Visualizing Mean and SD on Scatter Plot Question

Using the picture only (no calculators), estimate the following (roughly):

- \Box The sample mean mpg and horsepower: \bar{x} , \bar{y}
- \Box The sample std deviations: S_x , S_y



Sample Mean and Standard Deviations

■Sample mean

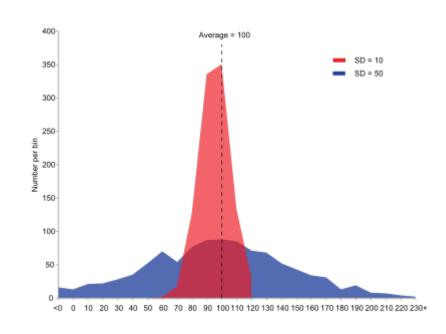
$$\bar{x} = \frac{1}{N} \sum_{i=1}^{N} x_i, \qquad \bar{y} = \frac{1}{N} \sum_{i=1}^{N} y_i$$

☐ Sample variances

$$s_x^2 = \frac{1}{N} \sum_{i=1}^{N} (x_i - \bar{x})^2$$
, $s_y^2 = \frac{1}{N} \sum_{i=1}^{N} (y_i - \bar{y})^2$

$$s_y^2 = \frac{1}{N} \sum_{i=1}^{N} (y_i - \bar{y})^2$$

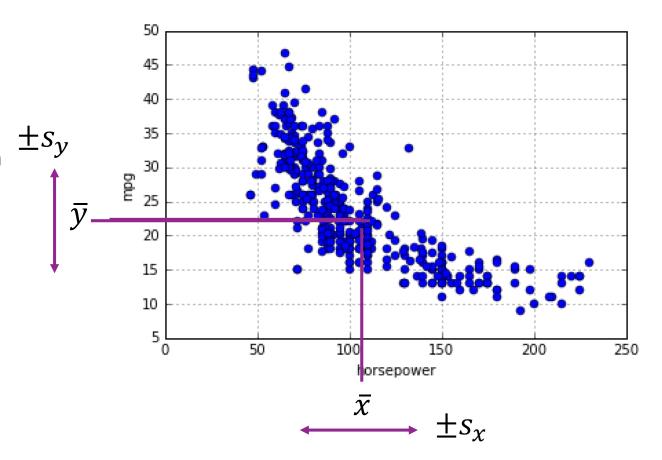
- Some formulae have a N-1 on denominator
- Creates an unbiased estimate. More on this later.
- ■Sample standard deviation
 - $^{\circ}$ S_{χ} , S_{V}
 - Square root of variances



Visualizing standard deviation https://en.wikipedia.org/wiki/Standard_deviation

Visualizing Mean and SD on Scatter Plot Approximate answer

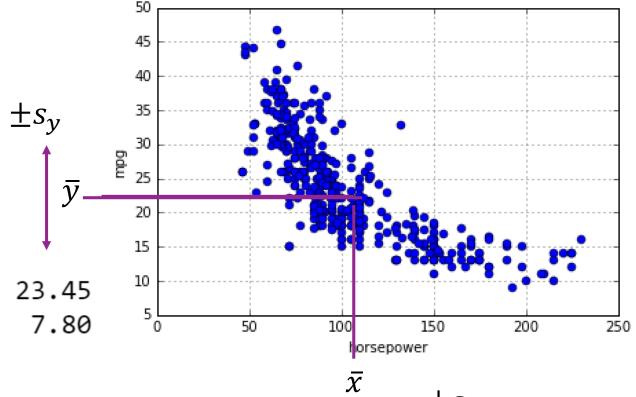
- lacktriangle Means: $ar{x}$ and $ar{y}$
 - Weighted center of the points in each axis
- \square Standard deviations: s_x and s_y
 - Represents "variation" in each axis from mean
 - With Gaussian distributions:1% of points are 3 SDs from mean

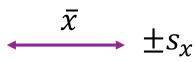


Computing Means and SD in Python

☐ Exact answer can be computed in python

```
xm = np.mean(x)
ym = np.mean(y)
syy = np.mean((y-ym)**2)
syx = np.mean((y-ym)*(x-xm))
sxx = np.mean((x-xm)**2)
beta1 = syx/sxx
beta0 = ym - beta1*xm
```









Sample Covariance

☐ Sample covariance:

$$s_{xy} = \frac{1}{N} \sum_{i=1}^{N} (x_i - \bar{x})(y_i - \bar{y})$$

- ☐ Will interpret this momentarily
- \square Cauchy-Schartz Law: $|s_{xy}| < s_x s_y$
- ☐ Sample correlation coefficient

$$r_{xy} = \frac{S_{xy}}{S_x S_y} \in [-1,1]$$

Alternate Equation for Variance

- □ Recall sample variance: $s_x^2 = \frac{1}{N} \sum (x_i \bar{x})^2$
 - Alternate formula:

$$s_x^2 = \frac{1}{N} \sum x_i^2 - \left(\frac{1}{N} \sum x_i\right)^2 = \frac{1}{N} \sum x_i^2 - \bar{x}^2$$

☐ Similarly, for covariance:

$$s_{xy} = \frac{1}{N} \sum x_i y_i - \overline{xy}$$



Notation

- ☐ This class will use the following notation
- ☐ We will try to be consistent
- Note: Other texts use different notations

Statistic	Notation	Formula	Python	
Sample mean	\bar{x}	$\frac{1}{n}\sum_{i=1}^{n}x_{i}$	xm	
Sample variance	$s_x^2 = s_{xx}$	$\frac{1}{n}\sum_{i=1}^n(x_i-\bar{x})^2$	sxx	
Sample standard deviation	$s_x = \sqrt{s_{xx}}$	$s_x = \sqrt{s_{xx}}$	sx	
Sample covariance	$S_{\chi y}$	$\frac{1}{n}\sum_{i=1}^n(x_i-\bar{x})(y_i-\bar{y})$	sxy	



Minimizing RSS

 \square To minimize RSS(β_0 , β_1) take partial derivatives:

$$\frac{\partial RSS}{\partial \beta_0} = 0, \qquad \frac{\partial RSS}{\partial \beta_1} = 0$$

☐ Taking derivatives we get two conditions (proof on board):

$$\sum_{i=1}^{N} \epsilon_i = 0, \qquad \sum_{i=1}^{N} x_i \epsilon_i = 0 \quad \text{where } \epsilon_i = y_i - \beta_0 - \beta_1 x_i$$

- ☐ Regression equation:
 - After some manipulation, (proof on board), solution to optimal slope and intercept:

$$\beta_1 = \frac{s_{xy}}{s_x^2} = \frac{r_{xy}s_y}{s_x}, \qquad \beta_0 = \bar{y} - \beta_1 \bar{x}$$

Simple Example

☐From:

http://stattrek.com/regression/regressio
n-example.aspx?Tutorial=AP

- Very nice simple problems
- ☐ Predict aptitude on one test from an earlier test
- ☐ Draw a scatter plot and regression line

How to Find the Regression Equation

In the table below, the x_i column shows scores on the aptitude test. Similarly, the y_i column shows statistics grades. The last two rows show sums and mean scores that we will use to conduct the regression analysis.

	Student	xį	y _i	(x _i - x)	(y _i - y)	$(x_i - \bar{x})^2$	$(y_i - \bar{y})^2$	$(x_i - \bar{x})(y_i - \bar{y})$
	1	95	85	17	8	289	64	136
	2	85	95	7	18	49	324	126
	3	80	70	2	-7	4	49	-14
	4	70	65	-8	-12	64	144	96
	5	60	70	-18	-7	324	49	126
Sum		390	385			730	630	470
Mean	ı	78	77					

The regression equation is a linear equation of the form: $\hat{y} = b_0 + b_1 x$. To conduct a regression analysis, we need to solve for b_0 and b_1 . Computations are shown below.

$$b_1 = \sum [(x_i - x)(y_i - y)] / \sum [(x_i - x)^2]$$

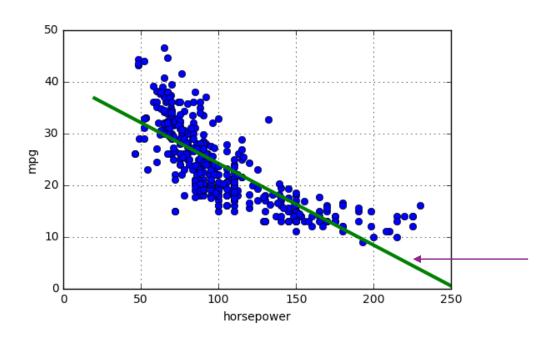
$$b_1 = 470/730 = 0.644$$

$$b_0 = y - b_1 * x$$

$$b_0 = 77 - (0.644)(78) = 26.768$$

Auto Example

☐ Python code



Regression line:

$$mpg = \beta_0 + \beta_1 horsepower$$

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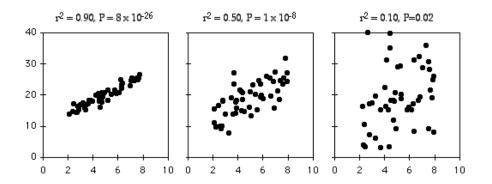


Mimimum RSS

☐ Minimum RSS (Proof on board)

$$\min_{\beta_0, \beta_1} RSS(\beta_0, \beta_1) = N(1 - r_{xy}^2) s_y^2$$

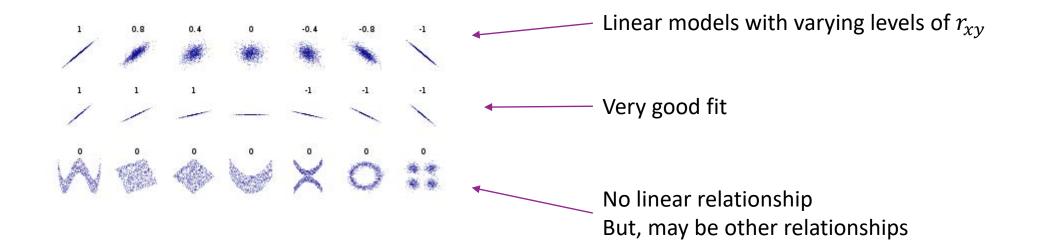
- \square Coefficient of Determination: $R^2 = r_{xy}^2$
 - \circ Explains portion of variance in y explained by x
 - s_{ν}^2 =variance in target y
 - $(1-R^2)s_y^2$ = residual sum of squares after accounting for x



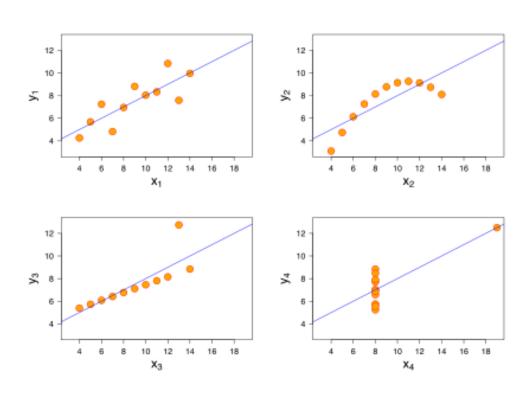
Visually seeing correlation

- $\square R^2 = r_{xy}^2 \approx 1$: Linear model is a very good fit
- $\square R^2 = r_{xy}^2 \approx 0$: Linear model is a poor fit.

$$\Box \beta_1 = \frac{r_{xy}s_y}{s_x} \Rightarrow \operatorname{Sign}(\beta_1) = \operatorname{Sign}(r_{xy})$$



When the Error is Large...



- ☐ Many sources of error for a linear model
- □Always good to visual inspect the scatter plot
 - Look for trends
- ☐ Example to the left
 - All four data sets have same regression line
 - But, errors and their reasons are different
- ☐ How would you describe these errors?



A Better Model for the Auto Example

- $\Box \text{Fit the inverse: } \frac{1}{\text{mpg}} = \beta_0 + \beta_1 \text{horsepower}$
- ☐ Uses a nonlinear transformation
- Will cover this idea later

