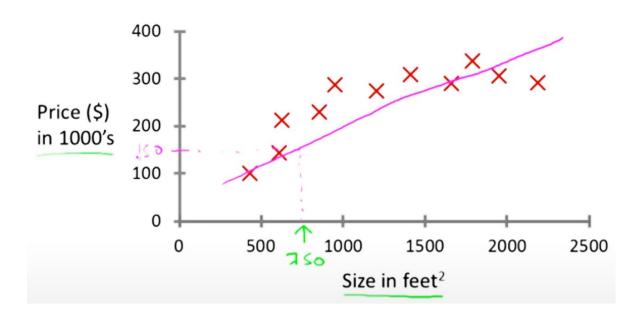
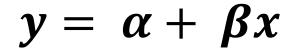
# Simple Linear Regression

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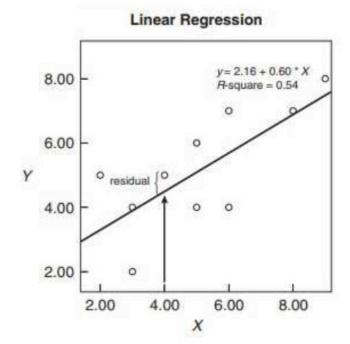


y =target (or dependent variable)

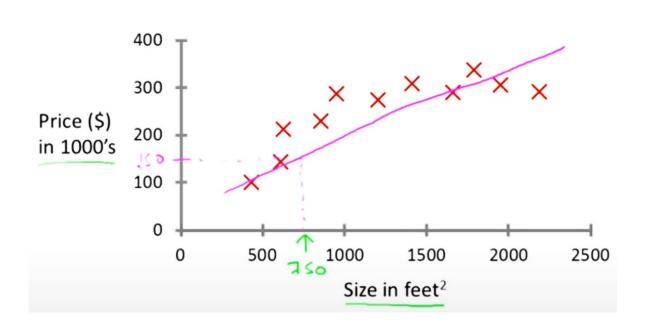
 $\alpha$  = y-intercept

 $\beta$  = slope

x = predictor (or independent variable)



## **Linear Regression**



The model function

$$y = \alpha + \beta x$$

y =target (or dependent variable)

 $\alpha$  = y-intercept

 $\beta$  = slope

x =predictor (or independent variable)

Fitted line minimizes the sum or mean of the squares of the errors Also known as ordinary least squares (OLS) regression – very popular!

# **Example**

Find values of  $\alpha$  and  $\beta$  that best fit y – try and get my predicted value of y as close as possible to the actual value of y

$$y = \alpha + \beta x$$

y = wage

 $\alpha$  = wage-intercept

 $\beta$  = slope

x = years of education

| Years of Education | Wage   |
|--------------------|--------|
| 16                 | 52,000 |
| 18                 | 65,000 |
| 16                 | 45,000 |
| 21                 | 80,000 |
| 14                 | 40,000 |
| 12                 | 50,000 |
|                    | •••    |

#### **Interpretation**

Find values of  $\alpha$  and  $\beta$  that best fit y – try and get my predicted value of y as close as possible to the actual value of y

$$y = \alpha + \beta x$$

y = wage

 $\alpha$  = wage-intercept indicates the value of the target (wage) when all predictors are zero  $\beta$  = slope indicates how much the target (wage) changes when the predictor (years of education) changes

x = years of education

 $R^2$  = percentage of variance in the target explained by the predictors, ranges from 0 to 1

#### **Disadvantages**



High sensitivity to the data: **Erroneous** or otherwise **outlying data points** can severely skew the resultant linear function



The data may require intensive manual manipulation and transformations

## Regression Analysis – Two Approaches

#### Explanation (Inference) – Traditional Use

Determine which predictors are the most useful for estimating the outcome variable.

Determine the amount of variance in the outcome variable that is explained by the predictor variable(s).

#### Prediction - Machine Learning

Predict values of the outcome variable from values of the predictor variable.

## **Traditional Use**

#### **Explanatory modeling**

• The goal is to explain the relationship between predictors (independent variables) and target (dependent variable)

#### **Model Evaluation**

Fit the data well and understand the amount of variance explained as well as the statistical significance of each predictor

 Evaluation (of goodness of fit) involves the use of (Adjusted) R-squared and p-values

# **Significance**

What does it mean for a predictor to be statistically significant?

- A measure of the probability that the observed effect was due to random chance.
- In practice, a p-value < 0.05.</li>
- A p < 0.05 means we are 95% confident the result is not a mistake (i.e., not driven by randomness)

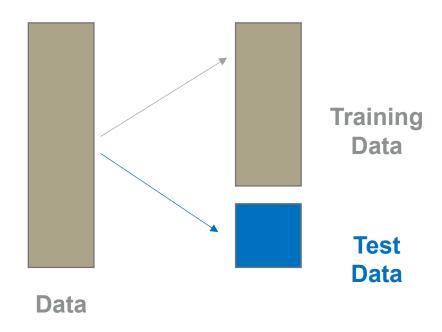
## **Significance**

What does it mean for a predictor to be statistically significant?

- Can also look at confidence intervals
- You can claim statistical significance (i.e., reject the null hypothesis) when the CI does not include zero.

## **Machine Learning Use**

- Predictive modeling
- Evaluate based on prediction error



# **Train and Test the Model**

