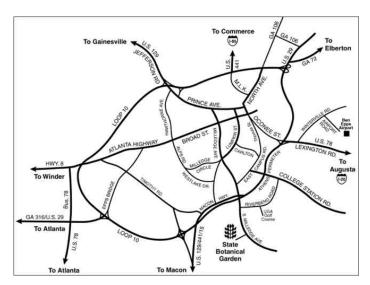
#### **Model Basics**

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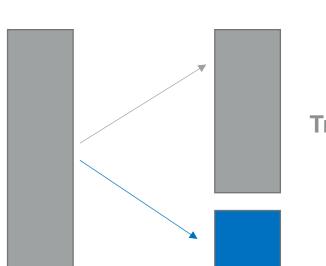
Business Intelligence Spring 2021





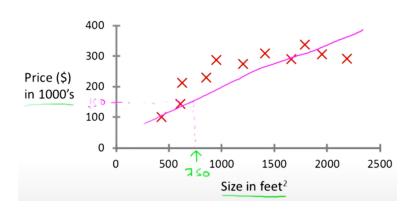






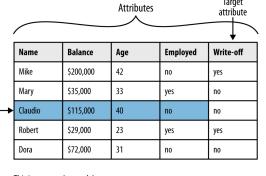
Data





**Training Data** 

**Test** 



This is one row (example). Feature vector is: <Claudio,115000,40,no> Class label (value of Target attribute) is no

#### **Model Defined**

A <u>simplified\* representation</u> of reality created for a <u>specific purpose</u>

• \*based on some assumptions about what is and is not important, or sometimes based on constraints on information or tractability

### **Model Goal**

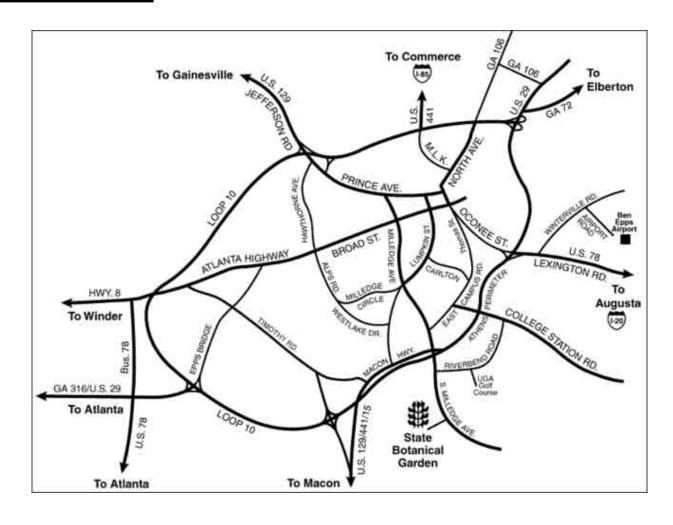
Not to uncover truth, but to discover a simple approximation that is still useful

• i.e., capture true "signals" (or patterns generated by the phenomenon of interest) and ignore "noise" (or random variation that you're not interested in)

#### "All models are wrong, but some are useful"

George Box

## **Model Example**



### Predictive (or Supervised Learning) Model

A formula for estimating the unknown value of interest:

# The target!

 The formula could be mathematical or a logical statement, such as a rule. Often, it is a hybrid

#### **More Terminology**

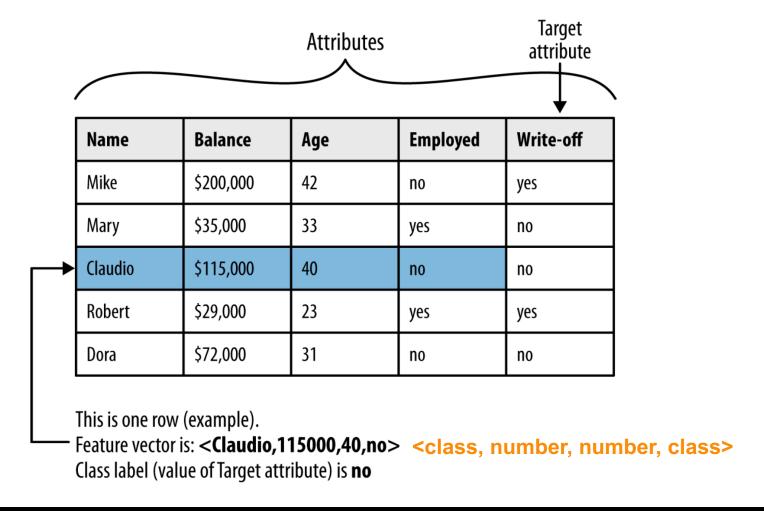
Instance / example = a fact or data point described by a set of attributes (also known as variables, columns, or features)

Model induction = the creation of models from data

Training data = the input data used for model induction

Test data = the input data used for model testing

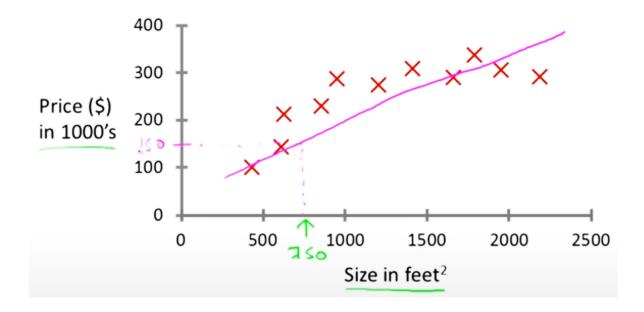
#### **More Terminology**



#### **Predictive Model: Regression**

Suppose you want to predict house prices, and you have some data about the price of a house (in thousands of \$) over size (sqft)

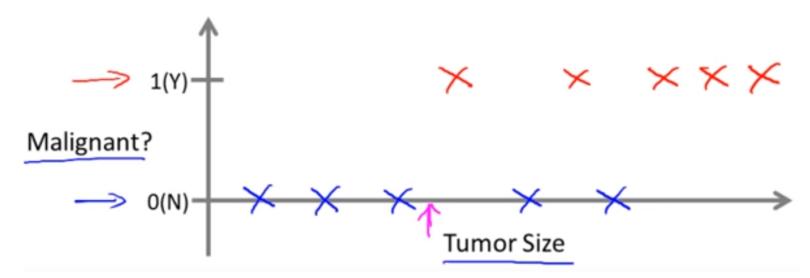
#### Estimate numeric value (e.g., with a linear regression)



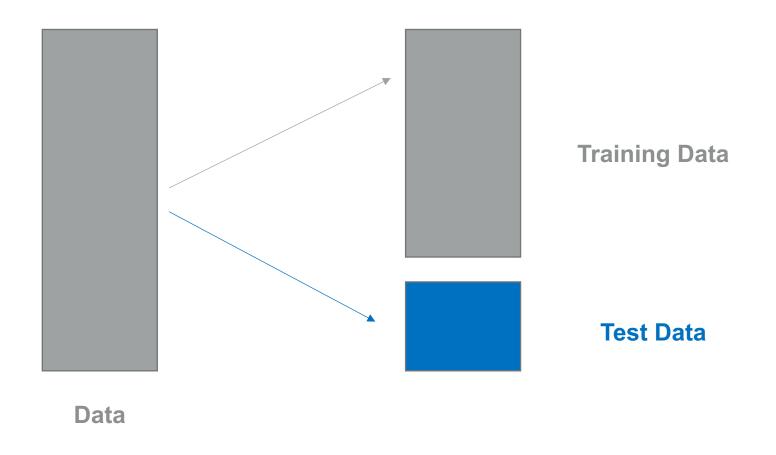
#### **Predictive Model: Classification**

Suppose you want to predict whether someone's breast cancer is malignant

#### Estimate class probability (e.g., with a logistic regression)



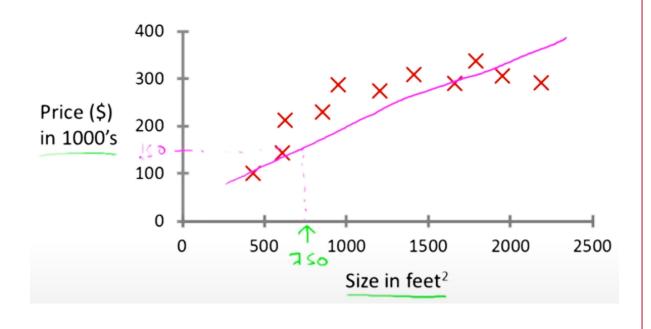
# **Predictive Model: Data Splitting**



#### Predictive Model: Regression Performance

Error = Predicted – Actual

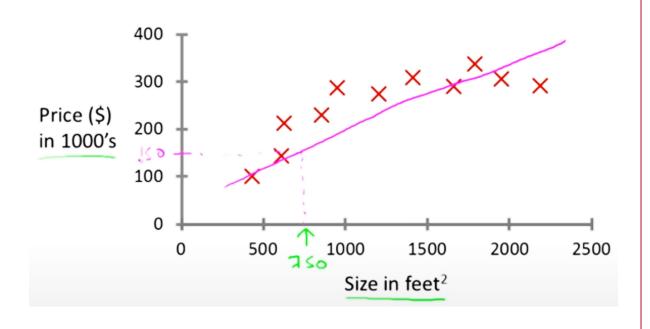
• To quantify the total amount of error, we take the square of each error and then sum all the squares, creating a measure called Total Sum of Squared Error (Total SSE)



#### Predictive Model: Regression Performance

Error = Predicted – Actual

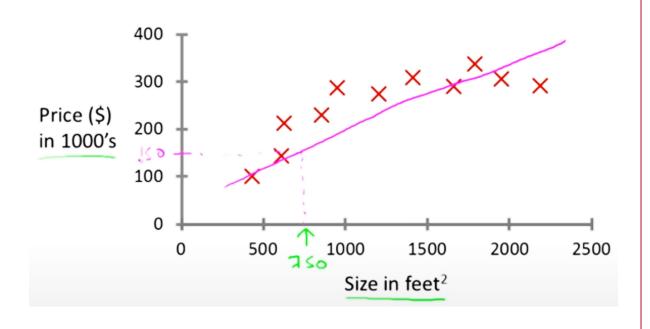
 Other popular measures that focus on error are RMSE (root mean squared error); MAE (mean absolute error); and MAPE (mean absolute percentage error)



### Predictive Model: Regression Performance

Error = Predicted – Actual

 RMSE is useful because intuitively, it captures how much the predicted values diverge from the actual values on average



Accuracy = Number of correct classifications made /
Total number of classifications

#### The confusion matrix

 Separates out the decisions made by the model, making explicit how one class is being confused for another

Predicted Positive True Positives
Negative False Negatives

Positive Negative

True Positives False Positives

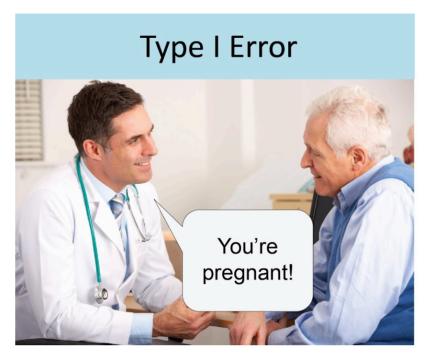
False Negatives

True Negatives

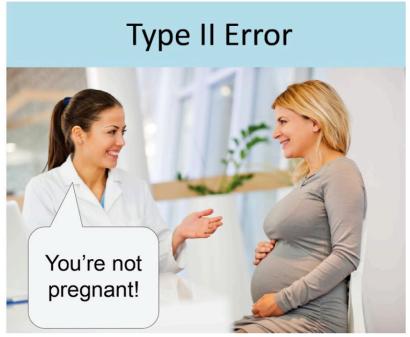
Type II Error

**Actual** 

**False Positive** 



#### **False Negative**



#### The F-Measure

- Special-purpose combination of precision and recall
  - Precision = true positives / true positives + false positives
  - Recall = true positives / true positives + false negatives
- Always between 0 (worst performance) and 1 (best performance)

### Things to Consider

- Is there a specific, quantifiable target that you are interested in predicting?
  - If yes, is it a class or a number?
    - Think about the decision
- Do you have data on the target?
  - Do you have enough data?
    - If the target is a class, a min of ~500 for each class type is needed

### **Another Thing to Consider**

- Do you have relevant data prior to the decision?
  - Think about the timing of decision and action leading up to it



### **Leakage**

The use of data in the model training process that would not be available in practice at the time you would want to use the model to make a prediction, leading to unrealistically good predictions

- Leaking from the future into the past
- Leaking test data into the training data

#### Leakage (Example)

Suppose you want to predict who will get pneumonia. Below is a preview of your data:

got_penumonia	age	weight	male	took_antibiotic	
false	65	100	false	false	
false	72	130	true	false	
true	58	100	false	true	

People take antibiotics after getting pneumonia in order to recover. Using **took\_antibiotic** to predict **got\_pneumonia** will lead to data leakage

#### **Avoiding Leakage**

- Do you have relevant data prior to the decision?
  - Think about the timing of decision and action leading up to it



#### **Summary**

- A model is not reality, but rather a simplified version of it that serves a specific purpose
- Good models are useful, all models are wrong
- Predictive models estimate an unknown value, which can be a class (classification) or a number (regression)

#### **Summary**

- When building a predictive model, split data into training and testing sets (70-30 or 80-20)
- We evaluate classification models differently than regression models
- Many things to consider when building predictive models, including the data type of target variable, how much data are available, and issues of leakage

# Thank You!