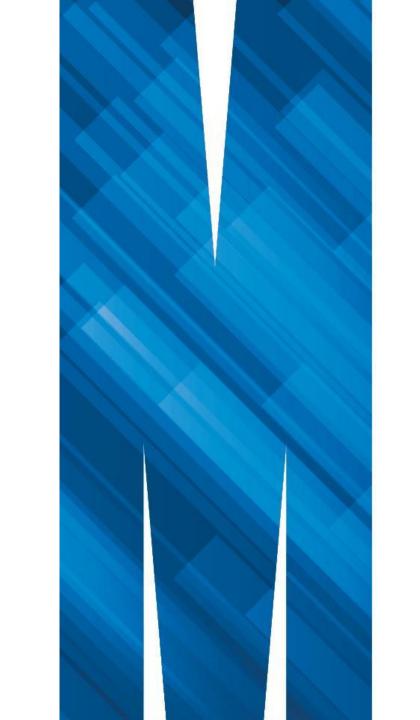


#### FIT1043 Introduction to Data Science

Week 5: Models

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School of Information Technology
Monash University Malaysia

With materials from Wray Buntine, Mahsa Salehi



# Week 4 Coverage Data Sources Data Wrangling





### Week 4 Coverage

#### **Data Sources and Wrangling**

Open Data

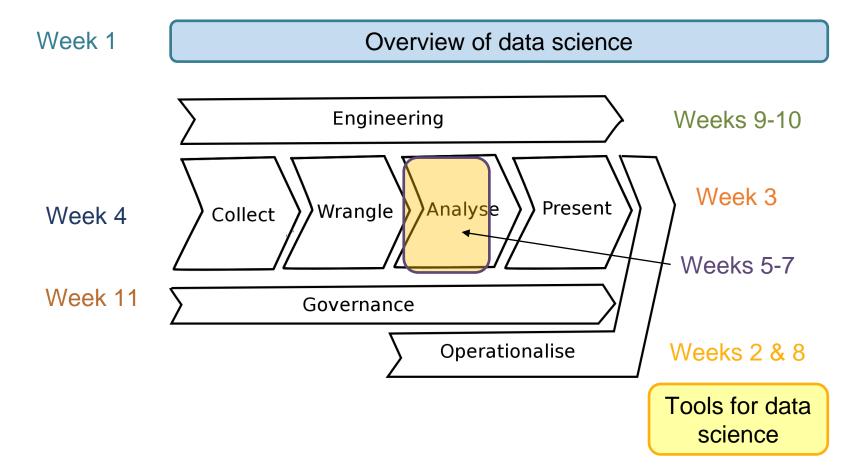
API

**Data Quality** 

**Data Auditing** 

Techniques to handle data quality issues







Week	Activities	Assignments
1	Overview of data science	
2	Introduction to Python for data science	
3	Data visualisation and descriptive statistics	
4	Data sources and data wrangling	
5	Data analysis theory	Assignment 1
6	Regression analysis	
7	Classification and clustering	
8	Introduction to R for data science Assignme	
9	Characterising data and "big" data	
10	Big data processing	
11	Issues in data management	Assignment 3
12	Industry guest lecture (tentative)	



#### Week 5 Outline

#### **Introduction to Data Analysis**

- What is model?
- What are predictive models?
- How to evaluate predictive models?

#### **Overview of Machine Learning**

- Machine learning styles
- What is learning theory
- Linear Regression
- Polynomial regression



#### **Learning Outcomes**

Week 5

#### By the end of this week you should be able to:

- Explain what are models and predictive models
- Analyse predictive models in different examples
- Understand how to evaluate predictive models
- Analyse how to estimate linear regression model
- Apply linear regression and polynomial regression on different data sets using Python



# Data Model









Can you draw this ...



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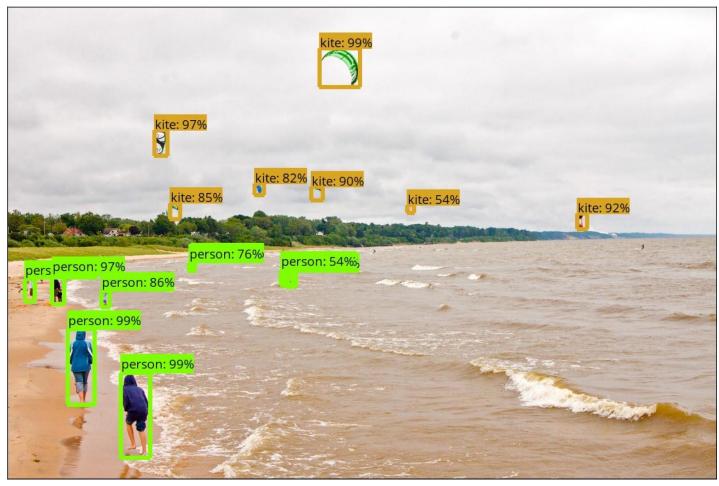
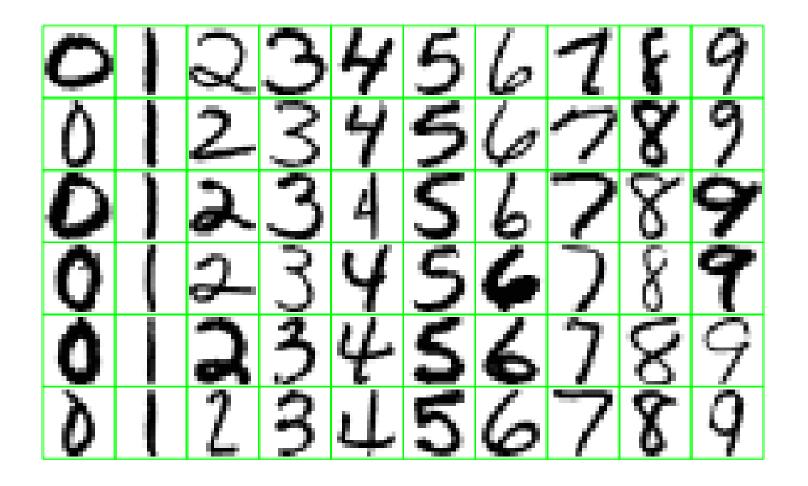


Image Source: https://medium.com/@WuStangDan/step-by-step-tensorflow-object-detection-api-tutorial-part-1-selecting-a-model-a02b6aabe39e



- (1) Help us understand how something works, and
- (2) Help us to predict the unknown.

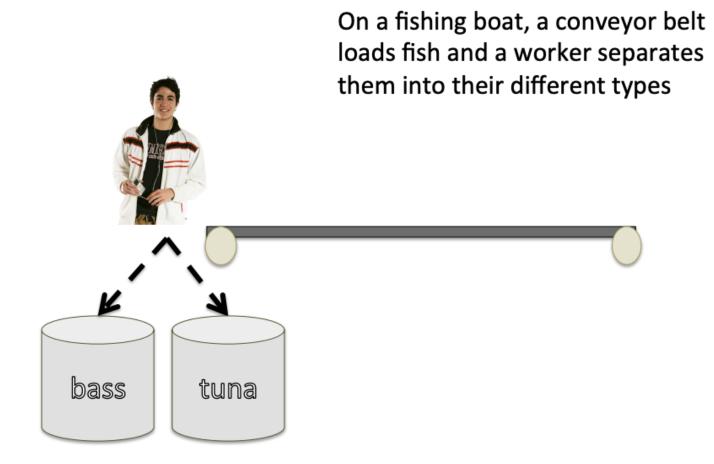




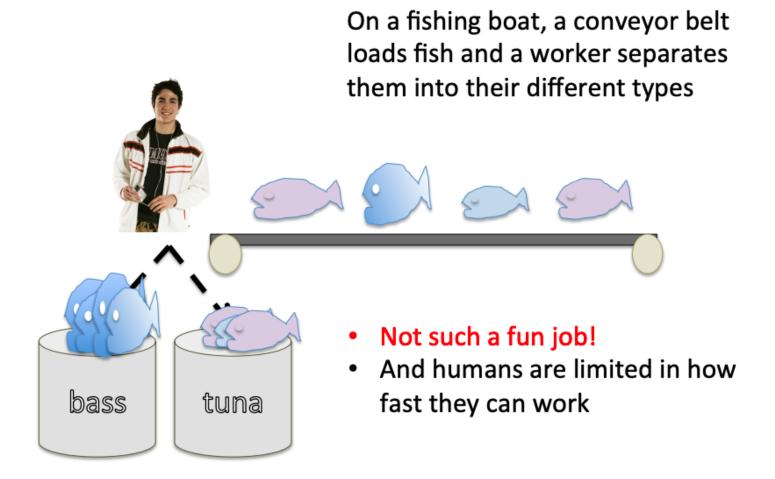
# A brief Introduction to Predictive Models For Data Science









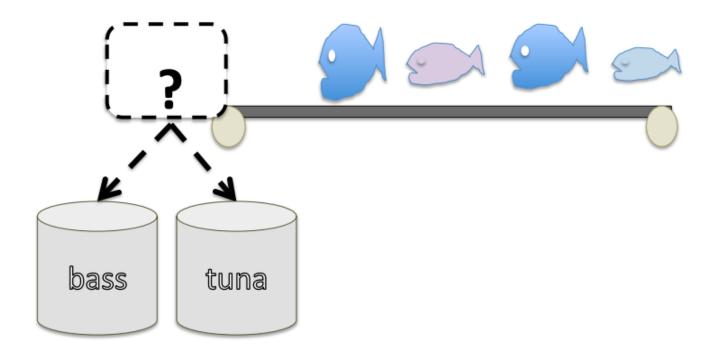




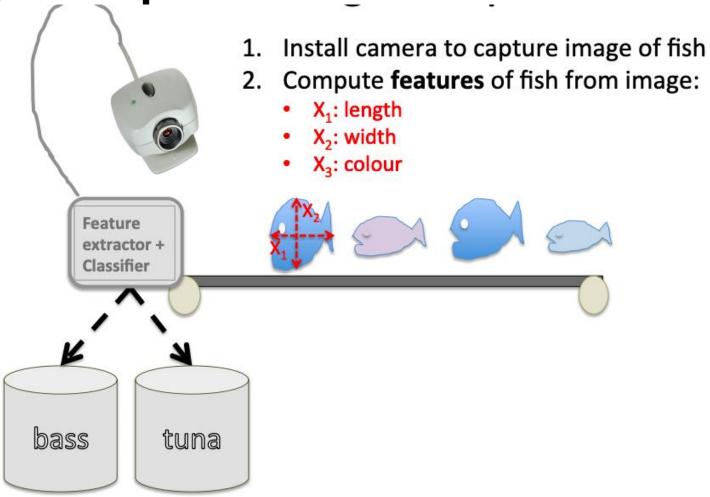




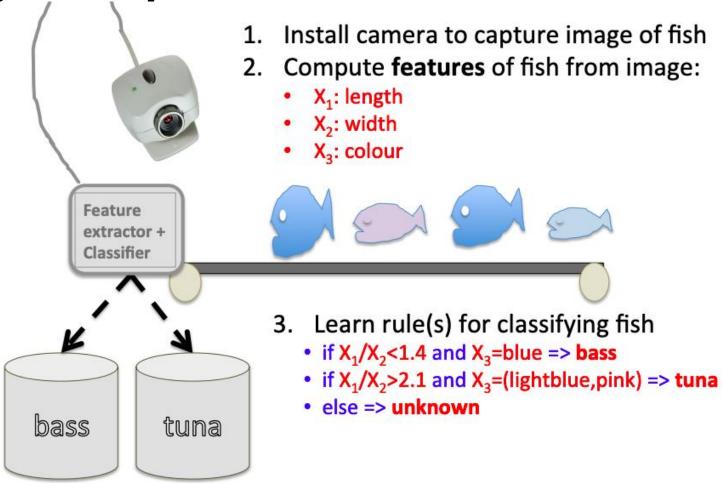
Question: Can we build a system to do the task automatically?











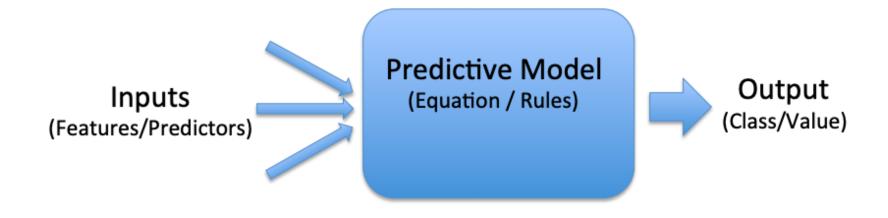




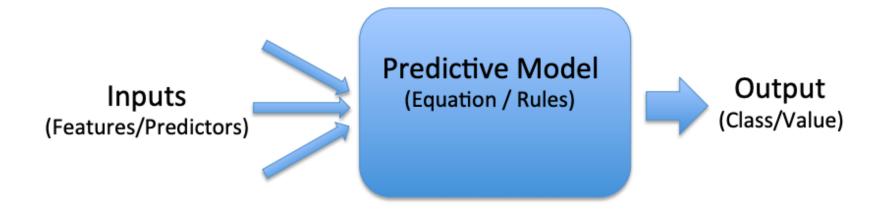


A predictive model is any model that makes a prediction

- Usually based on a set of features describing an object.
- The prediction could be:
  - A binary outcome (spam, not-spam)
  - Categorical (bass, tuna, other)
  - A real value (the age of the fish)
  - A vector of real values (probability of bass, tuna)
  - Etc.

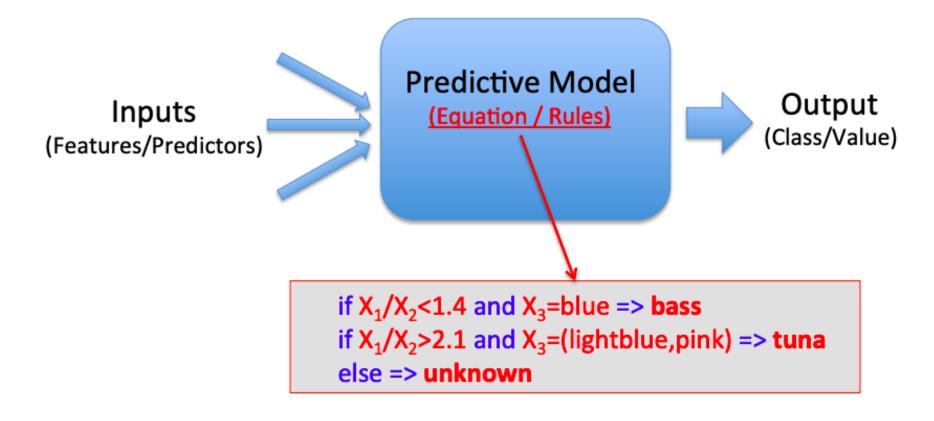






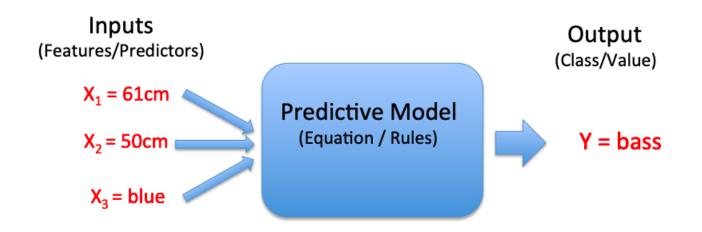
- If the predicted value is binary/categorical we usually refer to the model as a classifier
- If it predicts real values we refer to it as regression
- Although there are many other types of models, such as rankings, translation (your predictive words) and so on.

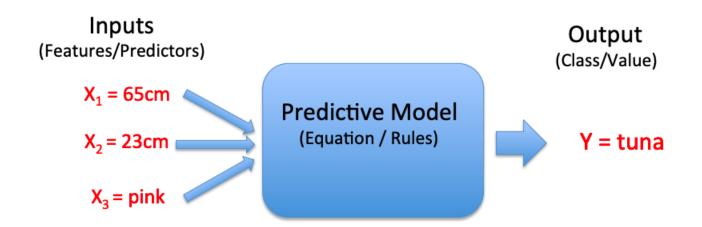




 The predictive model uses equations/rules to map the input features to output values









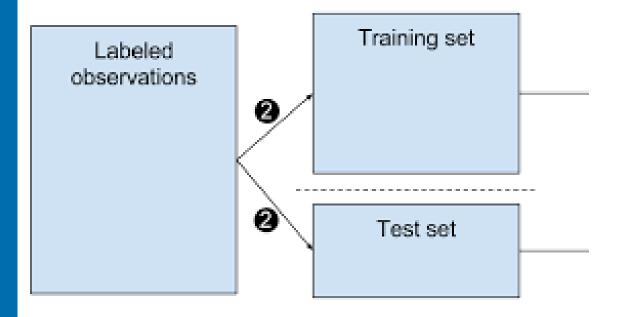
# **Models are Built from Examples**

Most models are developed through learning from examples

Instance	X1 = length	X2 = width	X3 = colour	Y = class
	55	51	blue	bass
	65	23	pink	tuna
	67	54	blue	bass
	54	20	light-blue	tuna
	62	26	pink	tuna
	44	62	blue	bass
	47	55	light-blue	bass
	73	31	pink	tuna
	54	48	light-blue	bass
	57	23	light-blue	tuna



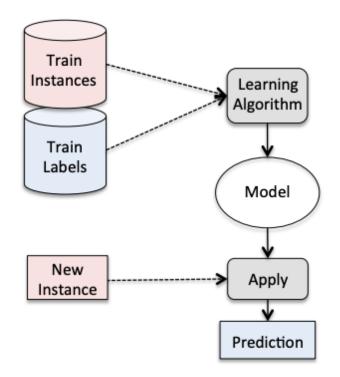
# Training and Testing Models





# **Training a Model**

Predictive models are learnt from training data and then applied to make predictions on new instances





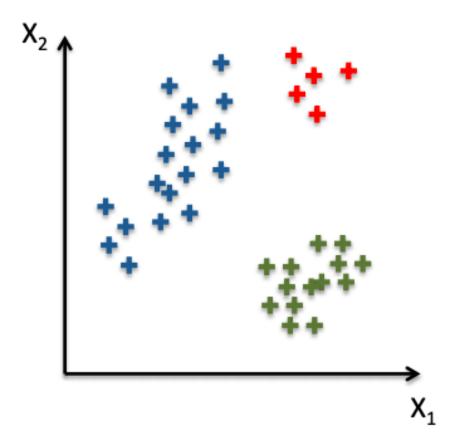
# Training a Model

How are models derived (learnt)?

Each training instance (fish in our case) is just a point in some feature space

Here the colour denotes the class

- blue = bass,
- green = tuna,
- red = unknown

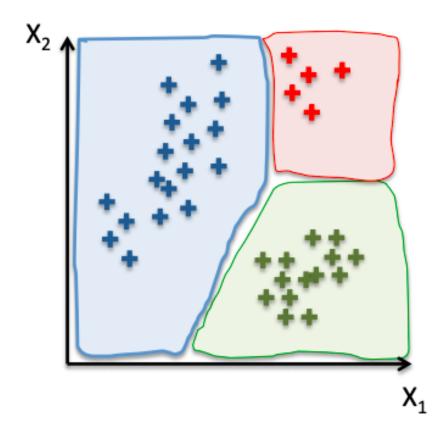




## **Training a Model**

How are models derived (learnt)?

Many (classification) learning algorithms work by dividing the feature space into regions of the same type

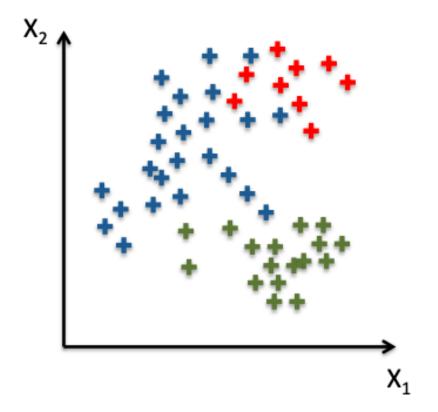




#### In Practise

In practice, the data is usually **overlapping** 

Making it hard to separate the classes

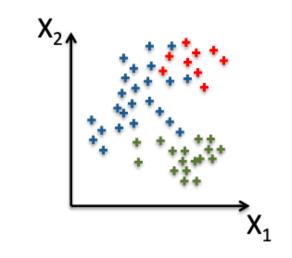


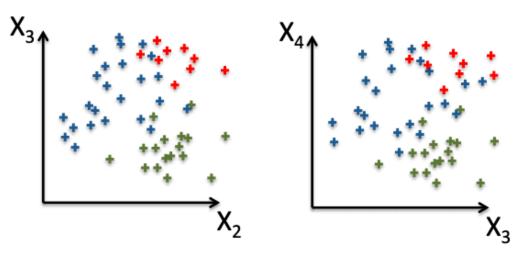


#### In Practise

And we have many feature dimensions

With some features more useful than others

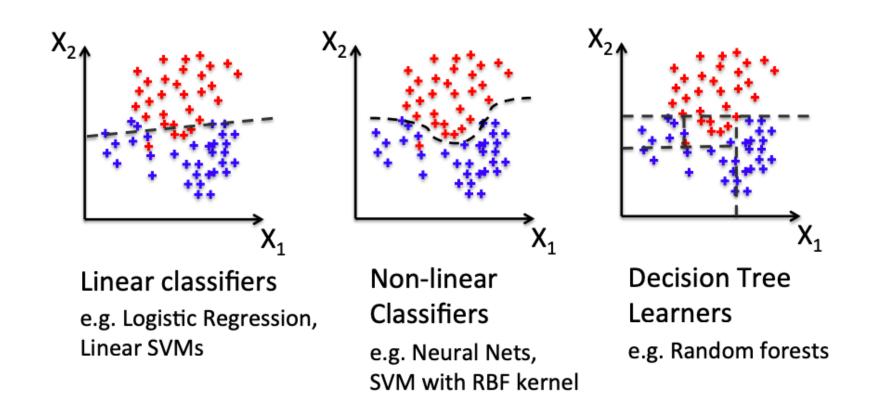






#### **Different Models**

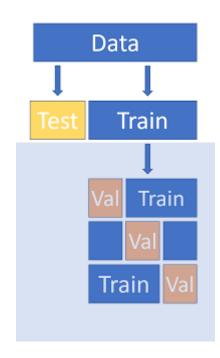
There are many different types of models that we can train to classify objects





#### **Different Models**

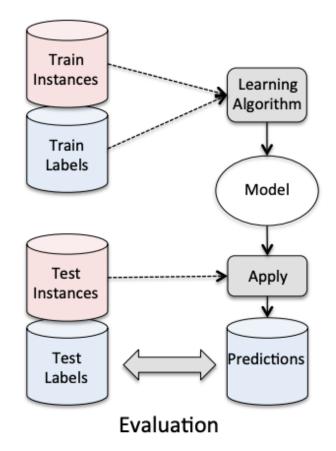
How can we decide which model is better?





# **Testing Models**

We evaluate predictive models based on how well they predict the labels for test instances (not used in training)





#### **Performance of Predictive Models**

#### Generally:

- The more training data the better the test performance
- And (providing there is sufficient training data) the more features the better performance will be
  - Caveat: There is a limit to how many features





#### **Home Activities**

Suggested Activities for the week

#### **Videos**

Watch <u>Jeremy Howard</u>, "The wonderful and terrifying implications of computers that can learn", TEDxBrussels

Watch Fei-Fei Li, "How we are teaching computers to understand pictures", TED2015

#### **Articles**

Read <u>Tarang Shah</u>, "About Train, Validation and Test Sets in <u>Machine Learning</u>", towards data science







#### **Recap: Learning Outcomes**

Week 5

#### By the end of this week you should be able to:

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