



# FIT5145 – A very Brief Introduction to Predictive Models

Mahsa Salehi\*

Faculty of Information Technology, Monash University

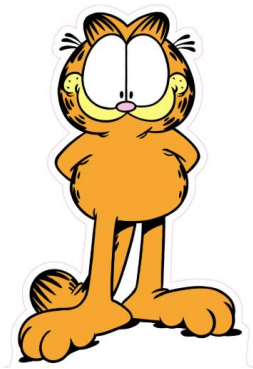
Semester 2, 2019

# What is Model?



# What is Model?

**Can you draw a CAT..**



# FLUX Question

Do you think you drew a perfect model?

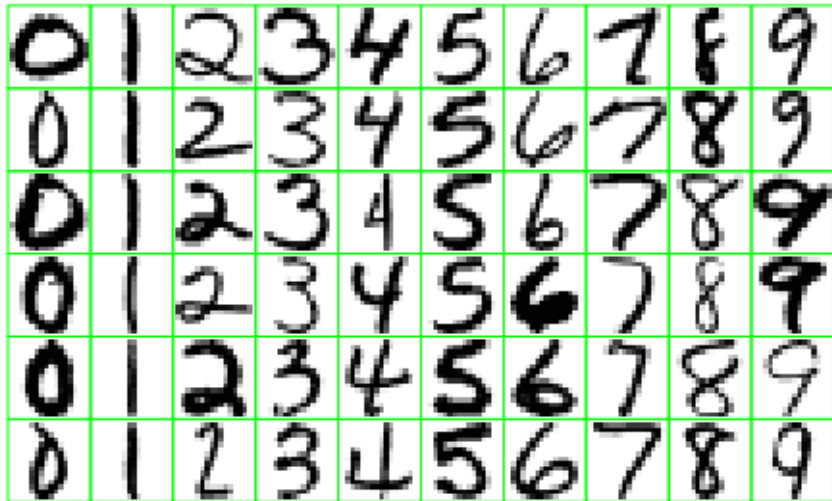
- A. Yes
- B. No
- C. Not sure



# What is Model?



# What is Model?



# FLUX Question

Which group does this horse belong to?



**Group A**



**Group B**

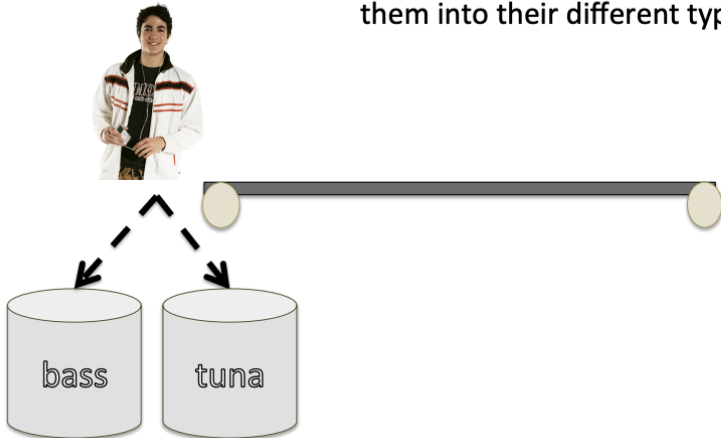
# A brief Introduction to Predictive Models For Data Science

(Example from Duda & Hart, PaCern Classification & Scene Analysis, 1973)



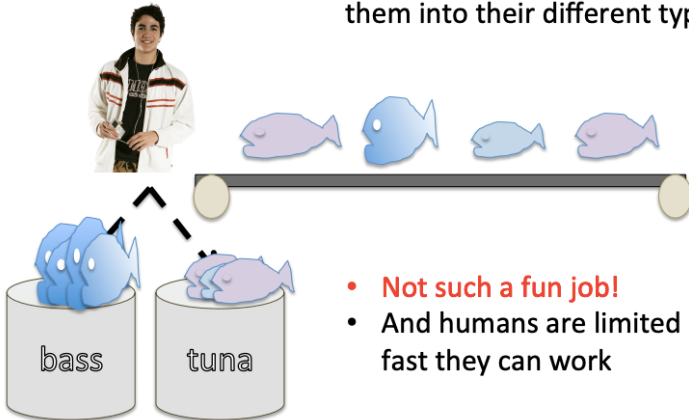
# Motivating Example

On a fishing boat, a conveyor belt loads fish and a worker separates them into their different types



# Motivating Example

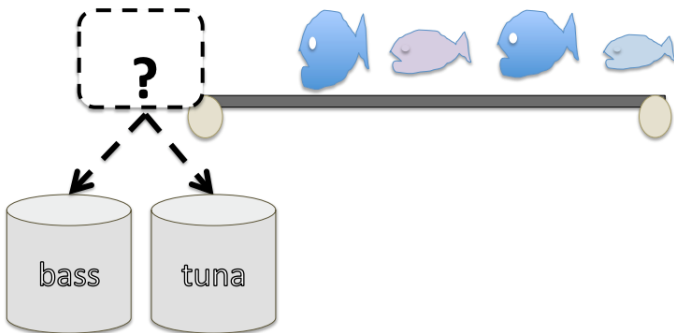
On a fishing boat, a conveyor belt loads fish and a worker separates them into their different types



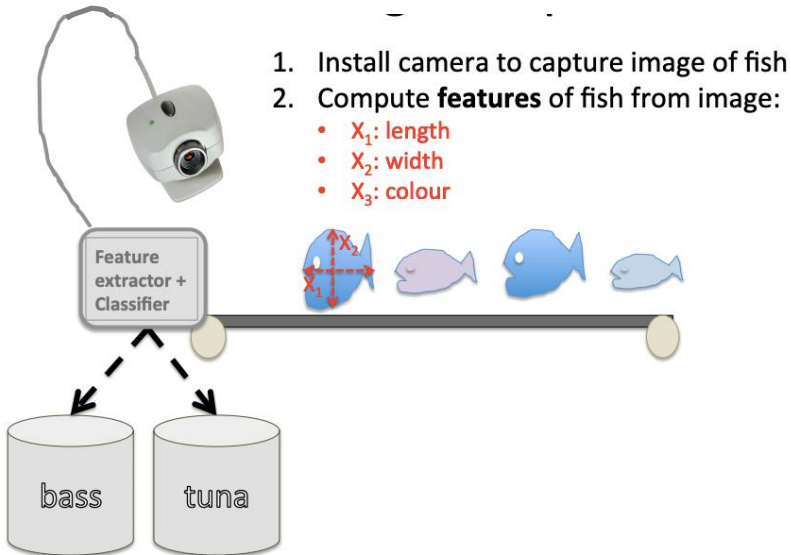
- **Not such a fun job!**
- And humans are limited in how fast they can work

# Motivating Example

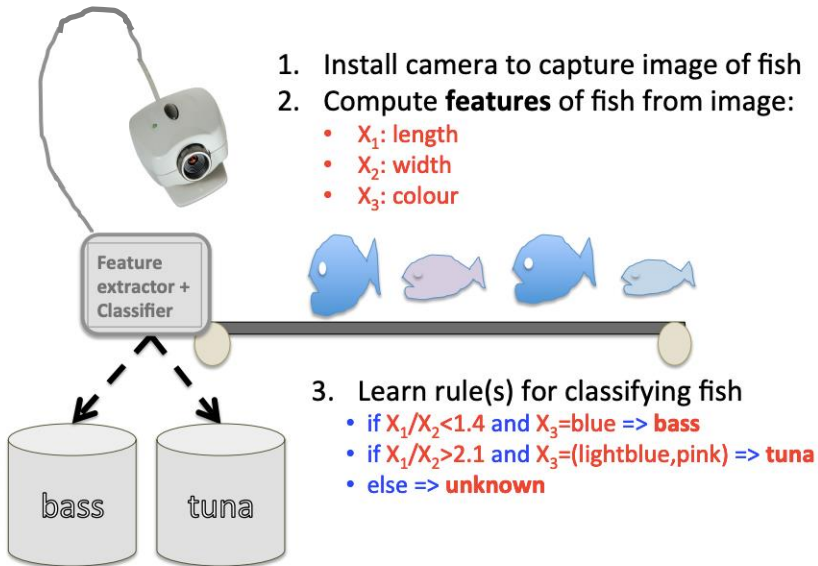
**Question:** Can we build a system to do the task automatically?



# Motivating Example

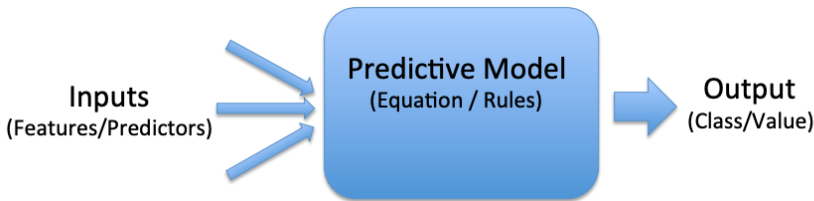


# Motivating Example



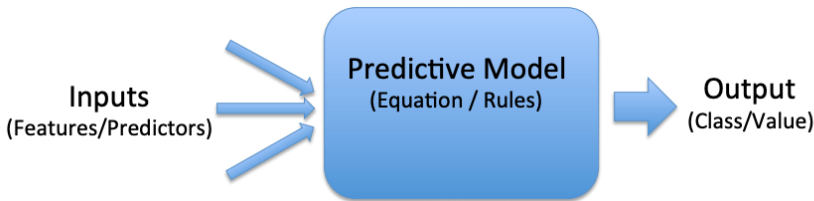
# Predictive Models

- 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.



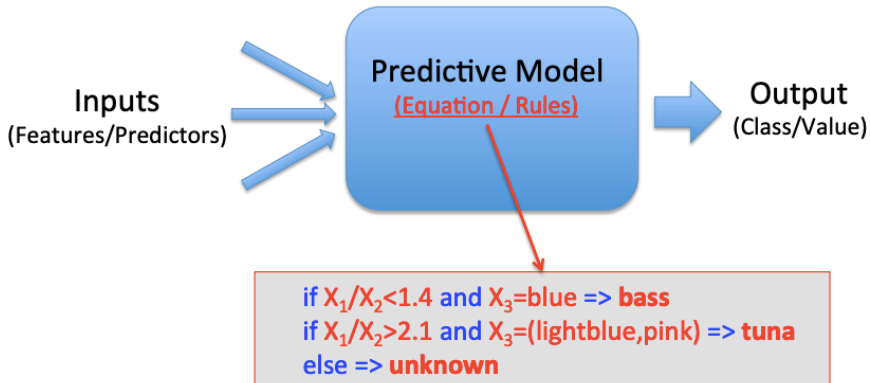
# Predictive Models

- ▶ 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 (e.g. ranking, translation, etc.)



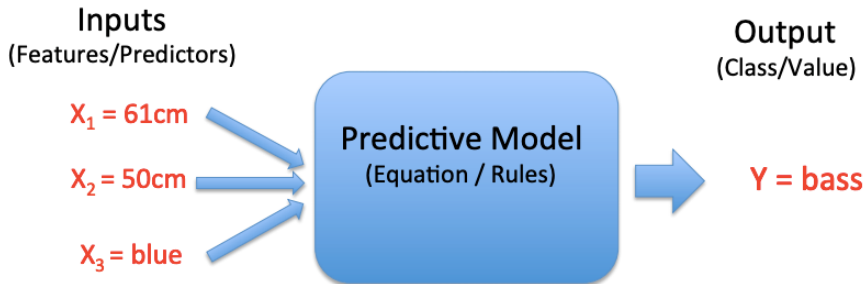
# Predictive Models

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

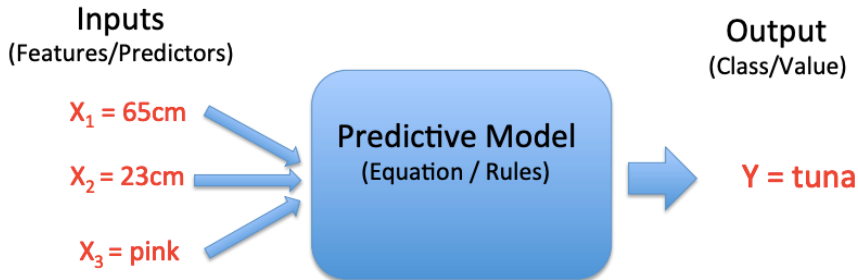




# Predictive Models



# Predictive Models

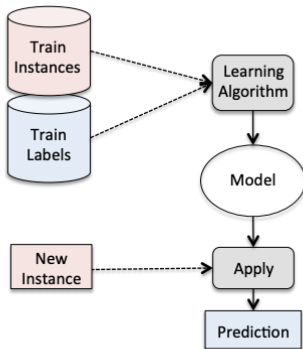


# Models are learnt from Examples

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

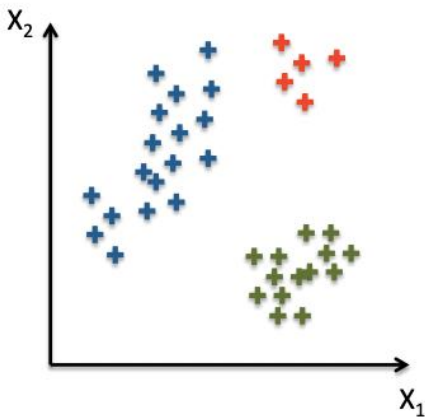
# Training a Model

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



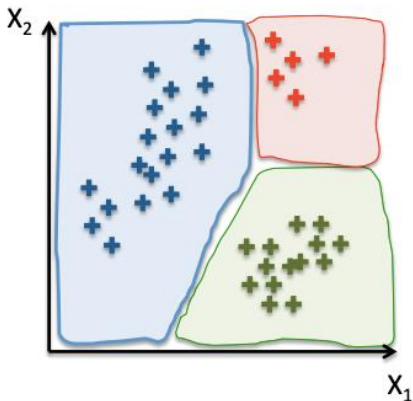
# How are models learnt?

- 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)



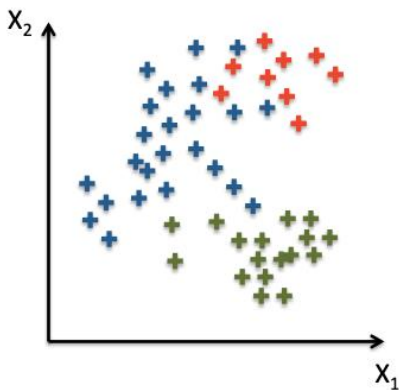
# How are models learnt?

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



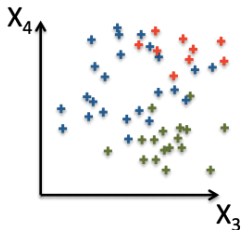
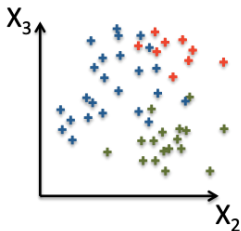
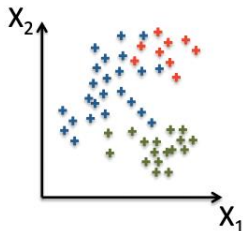
# In Practice

- In practice, the data is usually overlapping
- Making it **hard to separate the classes**



# In Practice

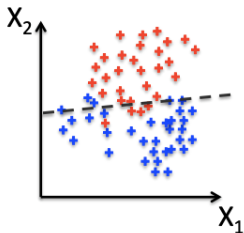
- And we have many feature dimensions
- With **some features more useful than others**



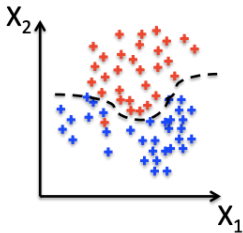


# Different Models

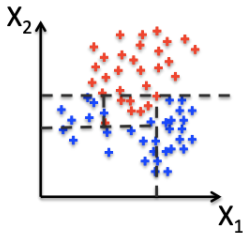
- There are many different types of models that we can train to classify objects



**Linear classifiers**  
e.g. Logistic Regression,  
Linear SVMs



**Non-linear Classifiers**  
e.g. Neural Nets,  
SVM with RBF kernel



**Decision Tree Learners**  
e.g. Random forests

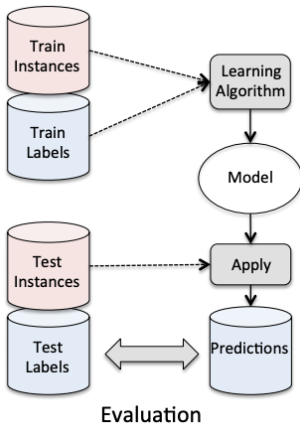
# FLUX Question

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



# End of Introduction

We'll talk more about predictive models in the coming weeks, especially in module 5.