# My First Classifier

by Dane Brown

The simplest solution is sometimes the best solution

### cvML Methodology

- Initialization: Call the cv or scikit model by name to create an empty instance of the model
- Set parameters: can be default, e.g. k -NN:
  specify k for more than one neighbour
- Train the model: train or fit is used to fit the model to some data.
- **Predict new labels:** use *predict*, to guess the labels of new (**unseen**) data.
- Score the model: refer to slide 10 & 13: works for both cv and scikit

## Unlike the Previous Program

### 2\_k-NN.py

- will take in (generated) training data points
- predict the label of the test (new) data points using k-NN
- The previous program simply showed how labels work (by simulating a classifier)

#### k -NN in a Nutshell

The class of a new (unseen) data point is **predicted** by finding the closest data point to it in the training dataset, i.e. its nearest neighbour

**Assumption:** the new data point probably belongs to the same class as its neighbour.

Multiple neighbours: In this case k-nearest neighbours are considered instead of a single one.

### How to k -NN (in OpenCV)

- Acquire some training data (generate mock data)
- Acquire some test data (generate mock data)
- Create a k-NN object
- Specify an odd number for k (optional)
- Find the nearest neighbors of a test data point, to be classified
- Assign the class label of the test data point by majority vote
- Plot the result

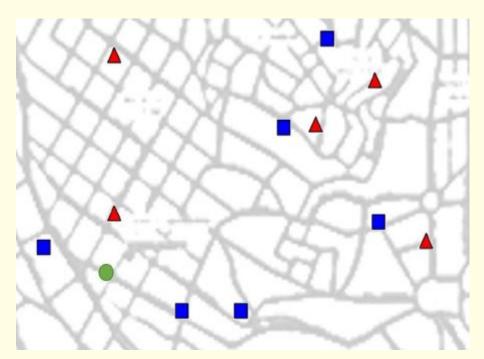
### Why an odd number for *k*?

It is advisable to only assign odd values for **binary** classification to avoid ties

Non-binary classification is a different story

## Predict a Rugby Fan

- Fans of two teams, Reds and Blues live here
- Fans always live closer to their fellow fans
- Add a new point somewhere
  - Predict whether the new point is a Reds or Blues fan



Check it out -> CV\_ML