Data Analytics

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## Chapter 1

### Lecture 6

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#### 1.1 Introduction

- Using basketball dataset (win loss & margins) found on OnQ under week 6.
- What's the point of these datasets?
  - The use of a model made form this data depends from case to case. There's no single rule for the usefulness of a model.
  - Hindsight?

#### 1.2 Design = Good Choices

- Generic workflow: file read  $\rightarrow$  binning & normalization & missing value inputation  $\rightarrow$  model  $\rightarrow$  scorer
  - Diagram (Based James?)
  - Random forests and Bagging/boosting diverge from the more typical workflows.
- When would one use binning?
  - Mapping numerical values to categorical values which better reflect meaning (e.g., numeric ages to categorical age ranges)
- When would one use normalization?
  - Depends on what will happen later in the workflow.
  - e.g., using neural networks, k-nearest neighbour
- Should try to design from "Right to left" of workflow since we're usually focused on the end result.
- Next stage:
  - Partitioning

- cross-validation (e.g., x-partition, x-aggregator)
- Bootstrap sample or Out Of Bag (OOB)
- Then we build the model
- Finally, score the results
- Clustering is thrown into all of that somehow
- Types of Models:
  - One R (Never use)
  - Bayes
    - \* Bad for redundant attributes
    - \* Good when there are many attributes that may be useful
  - k-NN
    - \* Bad for datasets with a large number of datapoints (big n), since scales based off of the size of the dataset
    - \* Great for datasets where the classes aren't clumped together (geometrically). Hard to determine with testing it.
  - Decision trees (Never use)
  - SVMs
    - \* Bad for problems with a large number of classes
    - \* Pretty good go to option otherwise (always)
  - Rules (Never use)
  - Random forests
    - \* No immediate major downsides
    - \* Therefore, practically always a good option
  - Neural networks
    - \* Bad if training cost is an important factor (big n)
    - \* Great when there's a small amount of info in each attribute. e.g., images, audio
- From the left side, use knowledge of dataset to make decisions on the workflow
- From the right side, use knowledge of the problem to make decisions on the workflow

# Chapter 2

# Lecture 7

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