# Monday May 19

# Session 1. Get Data

## **1.1 Learning Outcomes**

A successful student shall be able to:

- Explain different sources of data.
- *Explain different types of data formats.*
- Load data from open data sources into R.

# **1.2 Lecture/Demonstration**

- Introduction of the Data Science Life Cycle.
- Introduction to RStudio in Posit Connect
- Sources of data to include Open data.
- Different types of data formats: .csv files, excel files, urls, compressed data, Arrow-Parquet
- Rectangular data, vectors and data frames in R
- Methods for getting rectangular data from other sources or files into R data frames.
- Viewing data and getting summary statistics about data in R data frames.

# 1.3 Exercises

• Exercise on loading data from various sources into R and getting summary statistics.

# Session 2. Clean and Reshape Data

# **2.1 Learning Outcomes**

A successful student shall be able to:

- Explain the need to clean data.
- Clean and shape rectangular data: filter rows, select columns, rename columns, ...
- *Explain basic methods for reshaping data: pivots and joins.*

# **2.2 Lecture/Demonstration**

- Cleaning data
- Filtering rows using logical comparisons and %in%.
- Selecting columns using names and tidyselect.
- *Reshaping data using pivots.*

# 2.3 Exercises

- Exercise on cleaning data in R filtering, selecting. renaming
- Example on reshaping data

# **Tuesday May 20**

# Session 3 Visualize Univariate Data

# 3.1 Learning Outcomes

A successful student shall be able to:

- Create histograms, density plots, and bar plots.
- Customize plot titles, axis labels, ...

## **3.2 Lecture/Demonstration**

- *Variable types: continuous vs categorical*
- *Plots for continuous variables*
- Plots for categorical variables
- Customizing plots
- Example of dynamic titles

## **3.3 Exercises**

• Exercise on making and customizing univariate plots.

## Session 4. Visualize Multi-variate Data

## 4.1 Learning Outcomes

A successful student shall be able to:

- Create bivariate point plots and box plots.
- Use plot aesthetics to code by additional variables.
- Add and interpret linear and non-linear smoothers.

#### 4.2 Lecture/Demonstration

- Create bivariate point plots and box plots.
- Use plot aesthetics to code by additional variables.
- Add and interpret linear and non-linear smoothers.

#### 4.3 Exercises

• *Exercise on plotting bivariate and multi-variate data in R.* 

# Wednesday May 21

# Session 5. Statistical Tests and Models

#### **5.1 Learning Outcomes**

A successful student shall be able to:

- Explain the purpose for statistical tests and a Null hypothesis.
- Use and interpret the t.test function, aov/anova and lm functions in R.
- Generate residual plots in R.

## 5.2 Lecture/Demonstration

- Why statistical tests and understanding a Null hypothesis.
- Interpreting a p-value.
- Using and interpreting the t.test function, aov/anova, and lm functions in R.
- Generating and interpreting residual plots in R.

#### 5.3 Exercises

• *Exercises on t.test, linear regression from a scatter plot, and analysis of variance from Box Plot.* 

# Session 6. Classification Models

#### **6.1 Learning Outcomes**

A successful student shall be able to:

- Explain the difference between regression and binary classification.
- Use the glm function to generate and interpret a logistic regression.
- *Explain False Positives and False Negatives from a confusion matrix.*

#### 6.2 Lecture/Demonstration

- Binary Classification
- Logistic Regression
- Confusion Matrices and metrics

#### **6.3 Exercises**

• *Exercise on generating a logistic regression model and interpreting results.* 

# **Thursday May 22**

# Session 7. Overfitting and Bias-Variance Tradeoff

#### 7.1 Learning Outcomes

A successful student shall be able to:

- Explain the danger of overfitting.
- Explain the bias -variance trade off.
- *Explain the purpose of training and test data sets.*

#### 7.2 Lecture/Demonstration

- Degrees of Freedom and Restrictive versus flexible models
- Over-fitting
- Bias-Variance Trade off

• Validating and models by splitting data.

#### 7.3 Exercises

- Exercise on using splines to explore degrees of freedom and restrictive vs flexible models.
- *Exercise on using training and test data for linear regression.*

## Session 8. Evaluating and Tuning Models

#### **8.1 Learning Outcomes**

A successful student shall be able to:

- *Explain the need to evaluate and tune models.*
- Consider strategies for variable/feature selection Regularization.
- Explain strategies for evaluating model performance.

#### 8.2 Lecture/Demonstration

- Optimizing Loss Functions
- Evaluation Metrics
- Variable Selection and Partial F-Tests
- ROC Curves in Classification

## 8.3 Exercises

• Exercise on variable selection and tuning in regression and classification.

# Friday May 23

## Session 9. Neural Networks

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#### 9.1 Learning Outcomes

A successful student shall be able to:

• *Explain the basic architecture of a neural network.* 

- Explain the role of back-propagation and gradient descent.
- Build a basic neural network using R.

# 9.2 Lecture/Demonstration

- Convergence of big data (cheap storage and memory), big processing power (GPUs), and algorithmic advances.
- Neural Network architecture
- Role of Activation Functions and gradient descent and back-propagation.
- Challenges in Over fitting and convergence

## 9.3 Exercises

• *Exercise on building a neural network (neural net library)* 

# Session 10. Generative AI and Prompt Engineering

## **10.1 Learning Outcomes**

A successful student shall be able to:

- Explain what makes generative AI different.
- Explain the concepts of tokens and embeddings.
- *Explain the concept of semantic similarity and prediction.*
- *Apply concepts in prompt engineering.*

## **10.2 Lecture/Demonstration**

- Generative AI compared to predictive models.
- Tokenization and Embedding
- Semantic similarity for prediction.
- Dangers of overfitting and the need for regularization.
- Elements of prompt engineering.

## **10.3 Exercises**