# **2022 JPMC Intro to Data Science Agenda**

Duration: 3 days

# **Prerequisites:**

- Familiarity with Excel, basic knowledge of Python and data tables
- Pre-assessment consisting of 15-20 questions that will test capability of using Python (e.g. knowledge of if statements, looping, functions, data types, string methods, etc.)

#### <u>Day 1 – Intro to Data Analysis and Visualization</u>

# 1. Introduction to Data Analysis and Manipulation with Pandas

- Brief refresher on basics of Python and using the proposed IDE (Spyder, Jupyter or Athena)
- Intro to Python packages
- Intro to Pandas package (importing data, cleaning & exploring, slicing data, sorting & filtering)
- Advanced uses of Pandas (calculated fields, apply & rolling methods, data merging)

#### 2. Basic Statistics and Visualization

- Basic statistics terminology and functions (measures of central tendency, percentiles, population vs sample, etc.)
- Visualizations with pandas, matplotlib and seaborn
- Histograms and boxplots with seaborn
- Detection and removal of outliers

# Day 2 - Linear and Logistic Regressions

# 3. Regression Analysis with OLS of statsmodels

- Overview of simple regression models and OLS
- Explanation of correlation, R-squared, p-tests and error terms
- Multiple regression and multicollinearity
- Case study with single factor and multiple factors regressions (e.g. CAPM model, Fama French 3
  Factor Model)

# 4. Logistic Regression with sklearn

- Overview and validation of logistic regression models
- Multicollinearity in logistic regressions
- Individual impact of variables
- Confusion matrix
- Case study (e.g. credit card approval or investor classifier)

# **Day 3 – Decision Trees and Model Selection**

# 5. Decision Trees with sklearn

- Overview of decision trees and key terms (segmentation, entropy, information gain)
- Building and validating decision trees
- Pruning, fine tuning and prediction
- Decision trees for regression vs. classifiers
- Case study (similar case studies to Linear and Logistic Regression to show difference in model predictions)

# 6. Model Selection and Cross Validation

- How to validate and determine best model
- Discussion of overfitting/underfitting data
- Types of errors
- Training and tuning models with hyperparameters
- Splitting data into train/test
- Cross validation
- Case study (validating previous models covered in training program)