# Phase 3 Project Launch

**Machine Learning Classification** 

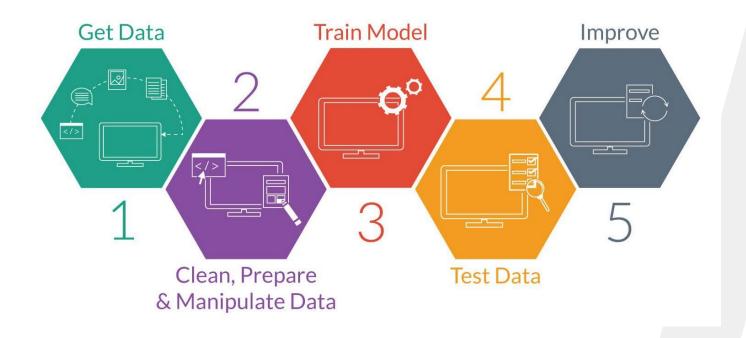


#### **Classification Project**

#### **Project Assignment**

- Phase 3 Project Rubric
  - Note: Rubric is similar, with a few changes
- Phase 3 Project Checklist

# **CRISP-DM**



# Remember the Workflow Steps!

- 1. Identify your business problem/stakeholder
- 2. EDA/Data Prep (class imbalance, visualizations, summary stats, correlations)
- 3. Train/Test Split, Cross-Validation, Transformations (OneHotEncoder, StandardScaler, MinMaxScaler, Regularization, SimpleImputer, etc.)
- 4. Baseline Model (Dummy Classifier)
- 5. Evaluate Baseline
- Run a model
- 7. Evaluate model confusion matrix, log loss, accuracy, precision, recall, ROC-AUC, recall, etc. Talk about class imbalance
- 8. Decide if you need another model, then repeat steps 6 & 8

#### **Business Understanding**

- Select a Stakeholder
  - Discuss cost of different errors.
  - Decide on an appropriate metric.



- How will your model be used?
  - More interpretable or better performance?

#### Data Understanding/EDA

- Do not forget to describe your data! Where did it come from? How big is it? What features does it contain?
- How well does the sample represent the population?
- Thorough EDA! Address class imbalance, missing values, correlations. Visualize!
- Especially if your final model is less interpretable, demonstrating good EDA will help earn the **trust** of your stakeholder



#### **Train/Test Split, Transformations, Cross-Validation**

- Train-test-split BEFORE transforming anything
- Fit and transform the training set, then <u>just transform</u> the test set exactly how the training set was transformed
- Many of the same tools as Phase 2, but now with additional practice with cross-validation
- Pipelines aren't required, but are convenient!

### **Iterative Model Building:**

#### Should you ever fit on the test set?



#### **Iterative Model Building: Train Test Split**

- Decide an approach to model validation
  - Untouched hold out set recommended
- Cross-validation is recommended for tuning
  - Minimize data leakage



#### **Iterative Model Building: Use All the Models!**

- Apply multiple models! Decide between model types based on not only the scores but also your business and data understanding!
- Document your iterative progress include just a selection of important models in final notebook

# **Iterative Model Building: Hyperparameter Tuning**

- Use cross-val to select best hyperparameters
- Find optimal complexity to balance bias/variance and maximize validation scores

- Use grid-search and pipelines to streamline process
- Tune to a metric aligned to business problem

#### **Iterative Model Building: Final Model**

- Select final model
- Fit on entire training dataset, score on test/holdout
- Inspect feature importances or parameters
- Inspect patterns of errors
- What did your model do well?
- What did your model not do well?
- Did your model perform as expected on the test/holdout set?

#### **Project Deliverables**

#### There are 3 Deliverables for this Project:

- Github repository
  - README.md
  - Clear commit history (good commit messages!)
  - Organized repository
- Final Notebook
  - One clean notebook
- Non-Technical Presentation
  - Aim for 5 minutes total

# Questions?