# Techniques in Predicting Income

Jan C. Bierowiec

## **Pre-Processing**

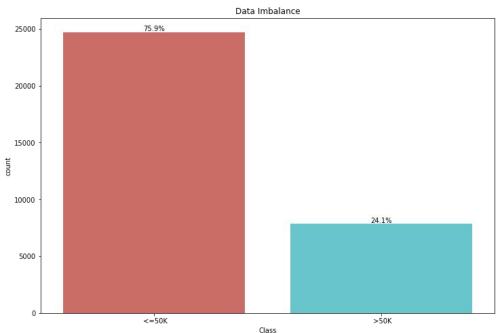
- Encoding
- Imbalanced Dataset
- Missing Values

## **Encoding**

- 14 variables
  - 6 continuous age, fnlwgt, education-num, capital-gain, capital-loss, hours-per-week
  - 8 discontinuous workclass, education, marital status, occupation, relationship, race, sex, native country
- Data encoded using dummy encoding
  - k-1 binary variables

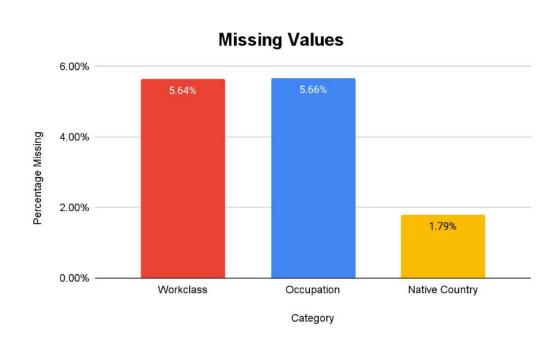
#### **Imbalanced Dataset**

- 75.9% <=50,000
- 24.1% >50,000
- Bagging method
  - Select random variables in training set and replace them with either 0, 1, >1 copies



#### **Missing Values**

- Three categories with missing values:
  - Workclass
  - Occupation
  - Native-Country
- Two methods chosen and compared:
  - KNN based imputation
  - Random Forest based imputation



# **Algorithm**

- Ensemble
- Imputation Results
- Normalization Results

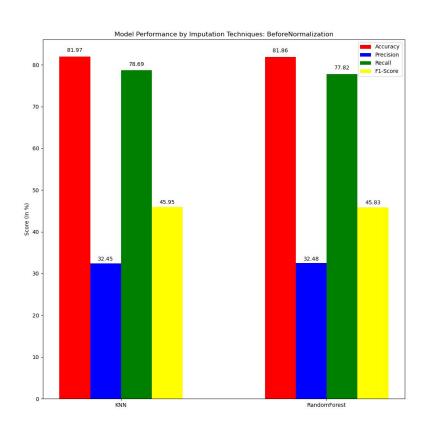
#### **Ensemble Learning**

- Ensemble classifier built from three supervised learning models:
  - KNN
  - Naive Bayes
  - Random Forest
- Ensemble learning is less biased towards any one individual learning algorithm.
- Z-score normalization results compared with non-normalization results

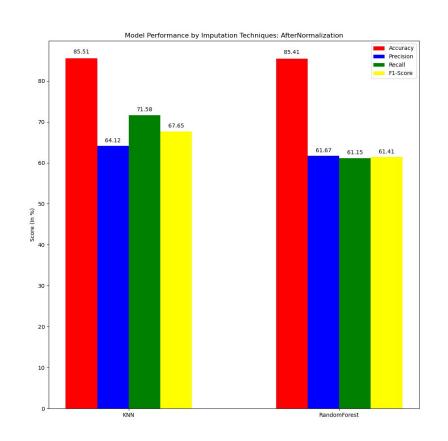
#### **Imputation**

- Model Based Approach :- Decrease variance, Decrease bias
  - o KNN
    - Grower's similarity coefficient to find the K nearest neighbor
    - Aggregation of the K values to impute the missing values
    - Aggregation is based on most frequent occurrences.
  - Random Forest
    - Build random forest for each individual missing value feature
    - Generalize ensembles of D-trees through bagging
    - Performs iterations to reduce *Out-Of-Bag* error

## **Imputation Methods Without Normalization**



#### **Imputation Methods with Normalization**



## **Results Comparison**

Method	Accuracy	Precision	Recall	F1-Score
KNN	81.97%	32.45%	78.69%	45.95%
Random Forest	81.86%	32.48%	77.82%	45.83%
Normalized KNN	85.51%	64.12%	71.58%	67.65%
Normalized Random Forest	85.41%	61.67%	61.15%	61.41%

#### **Future Steps**

- Parameter Optimization
  - Are there "best" parameters for the algorithms in each ensemble classifier for each data set
- Feature Selection
  - Determine if removing attributes improves predictive capabilities of learning algorithm
  - Try the Wrapper Method of Feature Selection
- Find out if there is a more optimal algorithm
- Try implementing a Neural Network

#### **Sources**

- Pantanowitz, Adam, and Tshilidzi Marwala. "Evaluating the impact of missing data imputation through the use of the random forest algorithm." arXiv preprint arXiv:0812.2412 (2008).
- Ganganwar, Vaishali. "An overview of classification algorithms for imbalanced datasets." International Journal of Emerging Technology and Advanced Engineering 2.4 (2012): 42-47.
- Stekhoven, Daniel J. "Using the missForest package." *R package* (2011): 1-11.
- Kowarik, Alexander, and Matthias Templ. "Imputation with the R Package VIM."
  *Journal of statistical software* 74
  (2016): 1-16.
- https://towardsdatascience.com/having-an-imbalanced-dataset-here-is-how-you-can-solve-it-1640568947eb