



PHISHGUARD: A MULTI-LAYERED ENSEMBLE MODEL FOR OPTIMAL PHISHING WEBSITE DETECTION

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CS 700
Fall 2023



Introduction



- Escalating sophistication and frequency in cyber attacks
- Financial and data losses from phishing incidents
- Traditional methods inadequate against new phishing tactics
- Multi-layered ensemble model for accurate, efficient detection

Techniques and System Analysis

■ Dataset

- Phishing website dataset *
(11055 x 32)

■ Machine Learning Models

- SVM
- Random Forest
- XGBoost

■ Data Preprocessing

- Data Sampling
- Feature Selection
- Mean, Median, Quartiles - Box Plots

■ Performance Analysis

- Hypothesis Testing
- Confidence Intervals
- ANOVA Test
- Contrast Approach

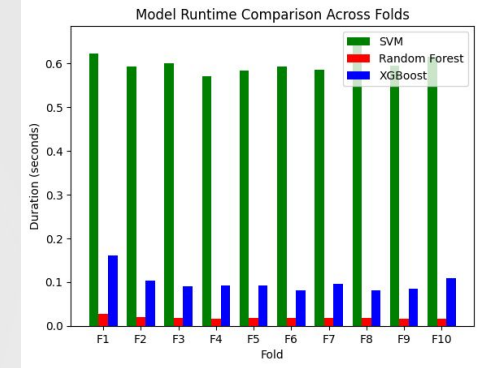
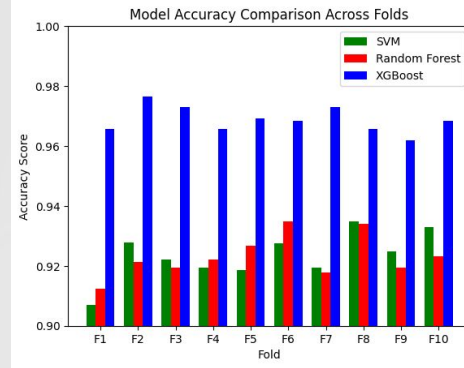
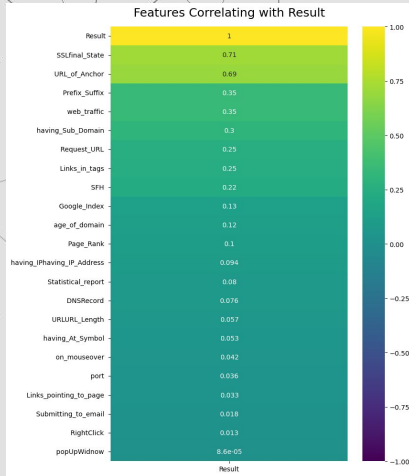


Methodological Approach



- ❑ Balanced dataset sampling followed by feature selection using Recursive Feature Elimination (RFE).
- ❑ Utilized box plots to represent data distribution, outliers, and feature relationships.
- ❑ Trained SVM, Random Forest, and XGBoost models using 10-fold cross-validation, focusing on accuracy and training/testing duration.
- ❑ Utilized hyperparameter tuning (grid search) to optimize model performance, alongside metrics like precision, recall, and F1-score.
- ❑ Conducted hypothesis testing and calculated confidence intervals.
- ❑ Applied ANOVA and contrast approach to evaluate and compare model performances. (significance level = 5%)

Preliminary Results



F-Statistic	154.15584
P-Value	1.7e-15
Decision	Significant differences
CI of SVM	(0.91769, 0.92927)
CI of Random Forest	(0.91808, 0.92814)
CI of XGBoost	(0.96560, 0.97180)

Table: ANOVA Test

Comparison	Confidence Interval	Decision
<i>SVM vs Random Forest</i>	(-0.00441, 0.00514)	No significant difference
<i>Random Forest vs XGBoost</i>	(-0.05037, -0.04081)	XGBoost better
<i>SVM vs XGBoost</i>	(-0.05000, -0.04045)	XGBoost better

Table: Contrast Approach