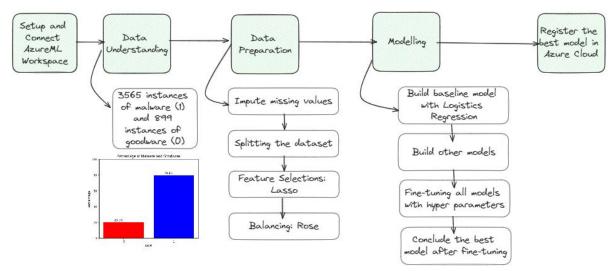


Methodology and Workflow

- The methodology applied for this dataset is based on the CRISP-DM Model.
- In the data understanding step, one row with a missing value is identified. This row is subsequently deleted in the data preparation step
- The dataset then undergoes feature selection using the Lasso method and balancing with the ROSE method due to the imbalance in the dataset (3565 instances of malware (1) and 899 instances of goodware (0)).





Primary Evaluation

• In total, 10 models are built to predict the dataset. Here are the final results in terms of Accuracy and AUC indicators.

		Accuracy	AUC
baseline	Logistic Regression	0.9697	0.9582
	SVM	0.9944	0.9944
	MLPClassifier	0.9922	0.9998
	AdaBoostClassifier	0.9877	0.9784
best mode	KNeighborsClassifier	0.9821	0.9964
	RandomForestClassifier	0.9966	0.9958
	DecisionTreeClassifier	0.9922	0.9860
	GaussianProcessClassifier	0.9944	0.9995
	HistGradientBoostingClassifier	0.9933	0.9999
	GradientBoostingClassifier	0.9933	0.9999
	GaussianNB	0.9698	0.9787

- As can be seen, the baseline model does not perform very well compared to the other models.
- All the models achieve notably high results, raising concerns that overfitting might occur during modeling.
- In a preliminary evaluation, the Random Forest model emerges as the best-performing model with an Accuracy score of 0.9966 and an AUC score of 1.0.
- Subsequently, fine-tuning will be performed to assess whether the models can achieve even higher scores.



Final Evaluation

• After fine-tuning, here are the new table results:

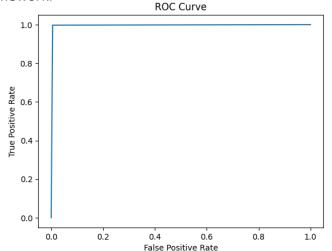
		Accuracy	AUC
baseline	Logistic Regression	0.9899	0.9909
	SVM	0.9955	0.9951
	MLPClassifier	0.9944	0.9996
	AdaBoostClassifier	0.9843	0.9889
best mode	KNeighborsClassifier	0.9855	0.9942
	RandomForestClassifier	0.9966	0.9958
	DecisionTreeClassifier	0.9933	0.9888
	GaussianProcessClassifier	0.9944	0.9995
	HistGradientBoostingClassifier	0.9922	0.9999
	GradientBoostingClassifier	0.9933	0.9995
	GaussianNB	0.9698	0.9787

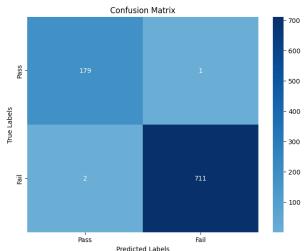
- Most of the models yield higher results after fine-tuning.
- Although the results for Random Forest remain the same, it is still considered the best model among all.
- In conclusion, Random Forest will be chosen as the optimal model for detecting malware for the company.



About the best model – Random Forest

- The model excels in malware and goodware detection with an exceptional accuracy of 99.66%.
- The confusion matrix indicates only 3 misclassifications out of 893 instances, showcasing robust performance in identifying both malware and goodware.
- Furthermore, the AUC Score of 99.58% underscores the model's excellent discrimination ability. In summary, the model's outstanding performance makes it a highly reliable choice for deployment in the company's security framework.







Reference

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- https://chat.openai.com/share/86bd6096-7b2d-4fed-b1d6-397c56dd2eed
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