Mecanismo para
Detecção de Transações
de Cartão de Crédito
Fraudulentas

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Objetivo

Efetuar a classificação de transações de cartão de crédito com foco na detecção de **Transações Fraudulentas** (recall).

Base de Dados

Credit Card Fraud Detection

Anonymized credit card transactions labeled as fraudulent or genuine

kaggle

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Variáveis

- Time
- Amount
- Class
- v1-v28

Base de Dados

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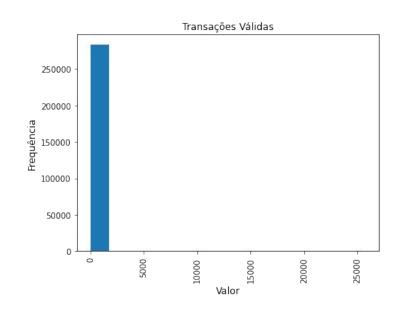
Variáveis

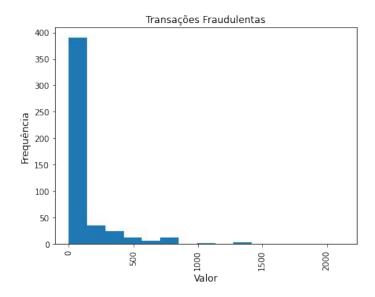
- Time
- Amount
- Class
- v1-v28

Distribuição

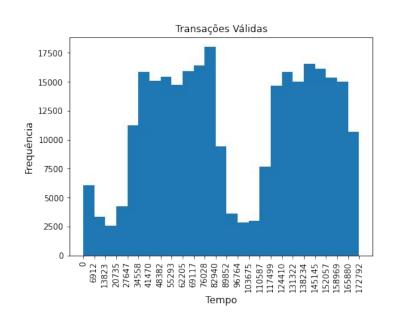
- Total de Amostras: 284.807
- Classe 0 (Válidas): 284.315 (99,82%)
- Classe 1 (Fraude): 492 (0,17%)

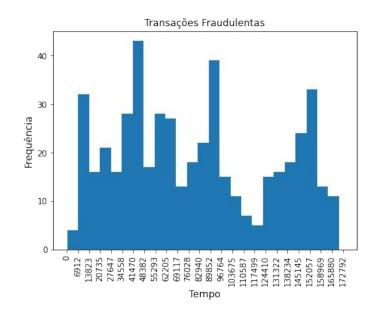
Análise de Dataset





Análise de Dataset





Experimentos

- KNeighborsClassifier (KNN)
- Multi-layer Perceptron (MLP)
- CatBoostClassifier (Random Forest)

Experimentos

Escolha de Classificador

- KNeighborsClassifier (KNN)
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Tuning de Classificador (CatBoostClassifier)

- learning_rate: 0.0947
- max_depth: 7
- n_estimators: 700



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Balanceamento de Bases de Treino

 Synthetic Minority Over-sampling (SMOTE)

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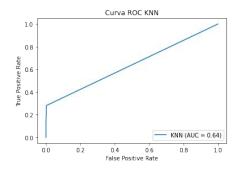
Balanceamento de Bases de Treino

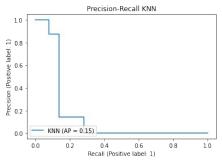
 Synthetic Minority Over-sampling (SMOTE)

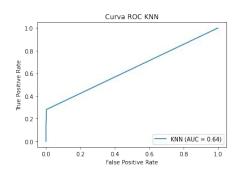
Considerações

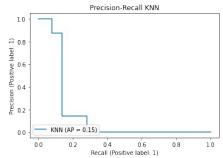
- Kfold
- Outliers
- CatBoostClassifier GPU

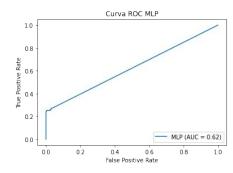


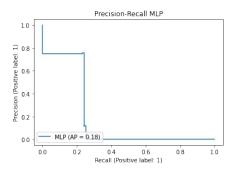




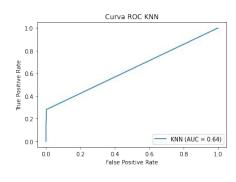


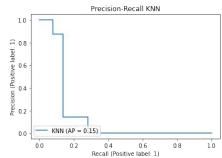


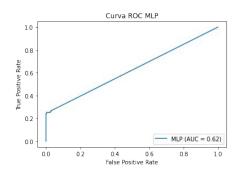


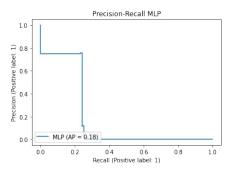


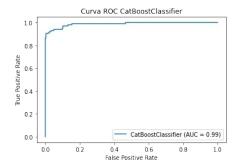


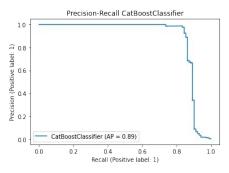




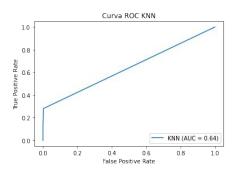


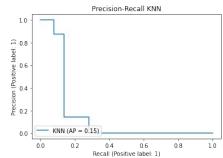


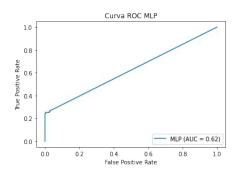


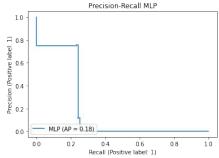


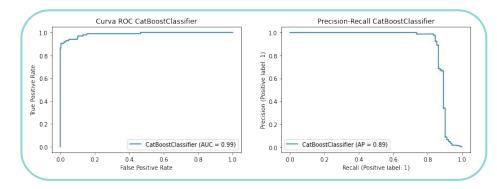




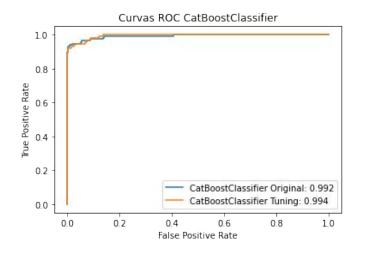


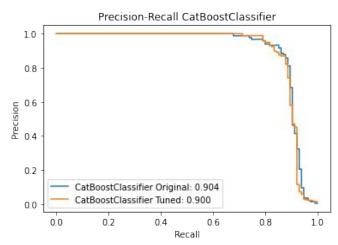




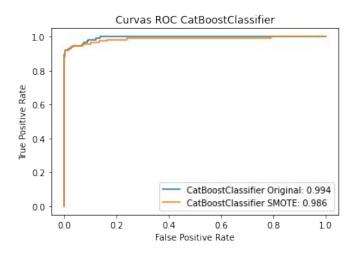


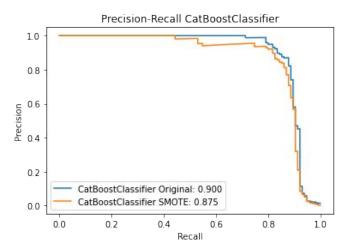
Tuning de Classificador

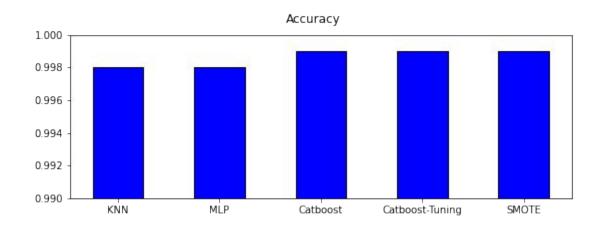


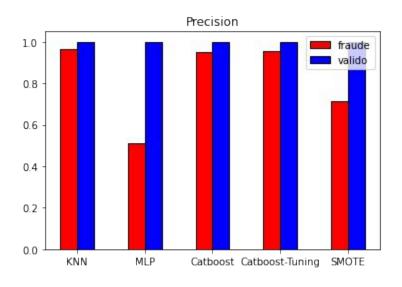


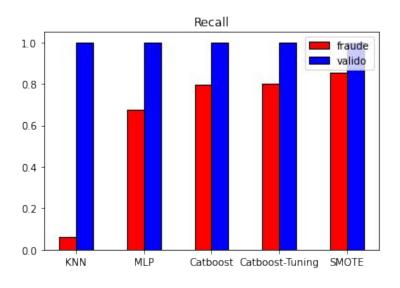
Over-sampling (SMOTE)

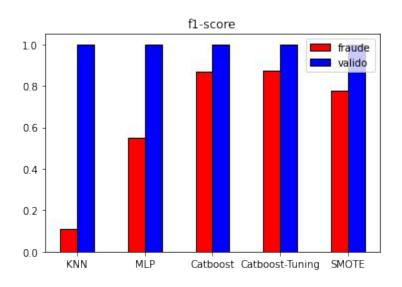






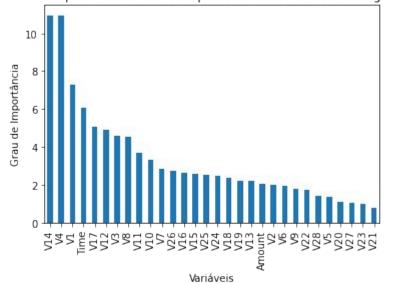






Importância de Variáveis (CatBoostClassifier-Tuning + SMOTE)

Análise de Importância de Variável para CatBoostClassifier-Tuning + SMOTE



Conclusão

- Classificador CatBoostClassifier foi o que atingiu as melhores métricas gerais e para a Classe 1 (Fraudes).
- Apesar de pouco, foi possível obter desempenho a partir do tuning do classificador CatBoostClassifier.
- Utilização da técnica de oversampling SMOTE contribuiu para a melhor classificação da transacções fraudulentas.
- De forma geral, as técnicas empregadas cooperaram com o objetivo principal do trabalho, aumentar a taxa de Verdadeiros positivos para Classe 1, Fraudes.

Obrigado

CatBoost. Catboostclassifier.

https://catboost.ai/en/docs/concepts/python-reference_catboostclassifier.

imbalanced-learn developers. Smote.

https://imbalanced-learn.org/stable/references/generated/imblearn.over_sampling.SMOTE.html.

scikit-learn developers. sklearn.model selection.kfold.

https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.KFold.html#examples-using-sklearn-model-selection-kfold.

scikit-learn developers. sklearn.neighbors.kneighborsclassifier.

https://scikit-learn.org/stable/modules/generated/sklearn.neighbors.KNeighborsClassifier.html.

scikit-learn developers. sklearn.neighbors.mlpclassifier.

https://scikit-learn.org/stable/modules/generated/sklearn.neural_network.MLPClassifier.html.

ULB, M. L. G. Credit card fraud detection.

https://www.kaggle.com/mlg-ulb/creditcardfraud/.