





Impact of Unusual Features in Credit Scoring Problem

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Before



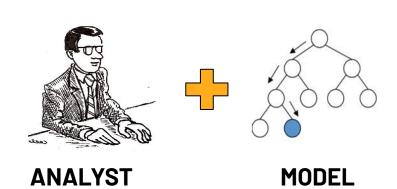








After









Features

Hard features











INCOME

SAVINGS

MARITAL

EMPLOYMENT

AGE

Soft features











GEOLOCATION

WEB

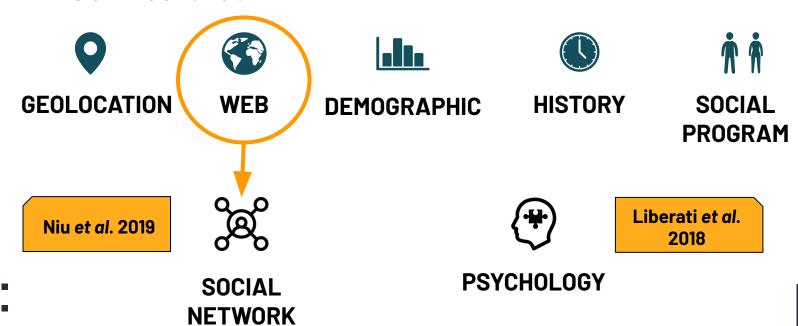
DEMOGRAPHIC

HISTORY

SOCIAL PROGRAM

Previous Works

Soft features



Our Contribution

Soft features

Vercosa et al. 2020



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GEOLOCATION

WEB

DEMOGRAPHIC

HISTORY

SOCIAL PROGRAM

Niu et al. 2019



SOCIAL NETWORK



Liberati et al. 2018

PSYCHOLOGY

Soft features



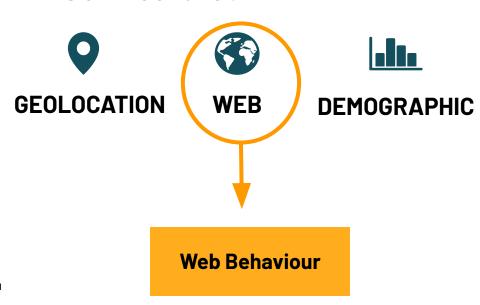






Region Behaviour

Soft features





HISTORY



PROGRAM

Soft features



GEOLOCATION



WEB



DEMOGRAPHIC



HISTORY



Purchase Power

Soft features



GEOLOCATION



WEB



DEMOGRAPHIC





SOCIAL PROGRAM





CREDIT CARD GRANTING

CAR INSURANCE

Experiment 1



Identify most promising features

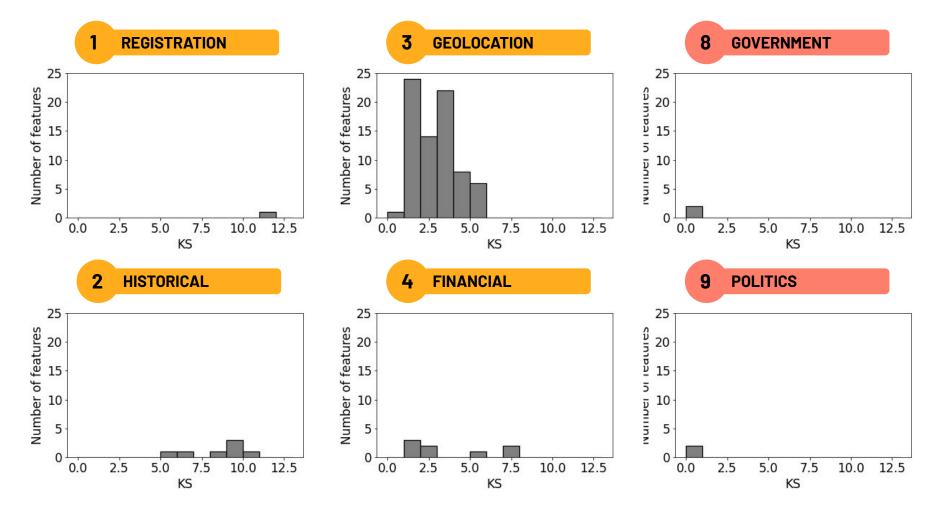


Kolmogorov-Smirnov (KS)



Bivariate Analysis

Results



Experiment 1: part 2



Identify most promising features groups

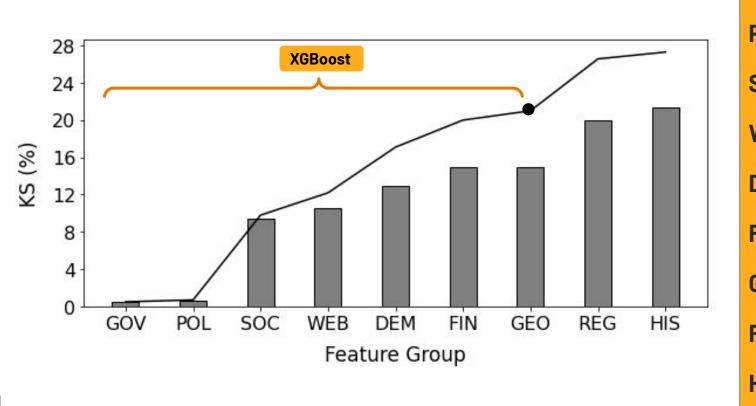


Kolmogorov-Smirnov (KS)



Extreme Gradient Boosting (XGBoost)

Results



GOV: Government

POL: Politics

SOC: Social Prog

WEB: Web

DEM: Demographic

FIN: Financial

GEO: Geolocation

REG: Registration

HIS: Historical

Experiment 2



Identify best performance



KS, Area Under Curve (AUC), MSLE, Lift



XGBoost, AdaBoost, Multi-layer Perceptron

- Experiment 2
 - Lift metric

- Models fine-tuning
 - GridSearchCV

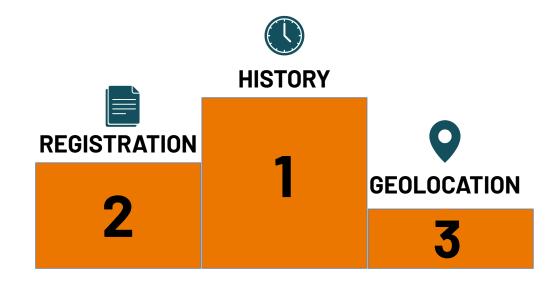
Results

Experiment 2

Models	Lift	MSLE	AUC (%)	KS (%)
MLP	0.6044	0.12	66.93	24.02
AdaBoost	0.6308	0.1240	68.08	26.34
XGBoost	0.1099	0.1099	68.81	27.24
Company	0.6296	0.1095	68.30	26.61

Conclusions

Promising feature groups



Conclusions

Promising models



Limitation

Available
 Registration
 features

Future Work

 Correlation of features

Feature selection