## LOAN PREDICTION

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Data processing

Exploratory data analysis

Building model

Model evaluation

### ISSUES WITH DATA

## Problems with "X":

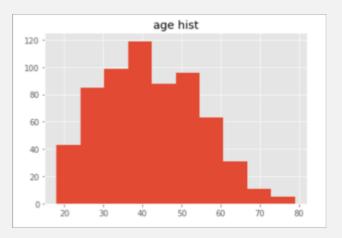
- Slightly corrupted
- Missing Value
- Several id unrecognized
- "Useless" variable

## Problems with "Y"

- Not matched with "X"
- Clients with more than two loans

### DATA PROCESSING

- Create "age" variable
- Extract "email" type
- Missing value imputation
- Data binning
- Create dummy variables



```
40s 170

30s 170

50s 135

20s 106

60s 41

70s 12

10s 6

Name: age, dtype: int64
```

```
hotmail 133
aol 132
bing 130
yahoo 130
gmail 115
Name: email, dtype: int64
```

## **EXPLORATORY DATA ANALYSIS**

## After data cleaning:

#### 640 rows × 55 columns

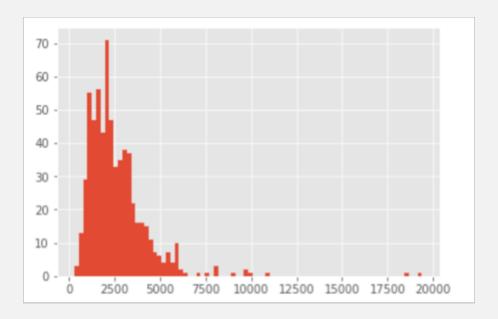
```
0 351
1 289
Name: flgGood, dtype: int64
```

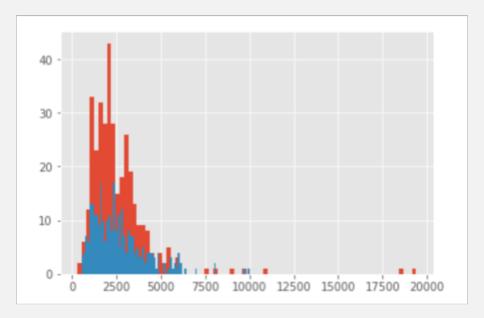
amount_requested	0
monthly_rent_amount	0
loan_duration	0
num_payments	0
payment_amount	0
amount_approved	0
duration_approved	0
payment_amount_approved	0
address_zip	0
bank_routing_number	0
monthly_income_amount	0
raw_12c_score	0
raw_FICO_telecom	0
raw_FICO_retail	0
raw_FICO_bank_card	0
raw_FICO_money	0
flgGood	0
age_20s	0
age_30s	0
ane 40s	0

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# MONTHLY\_INCOME \_AMOUNT

count	640.00000		
mean	2616.026563		
std	1712.436920		
min	300.00000		
25%	1503.000000		
50%	2245.000000		
75%	3200.000000		
max	19392.000000		
Name:	monthly_income_amount,	dtype:	float64

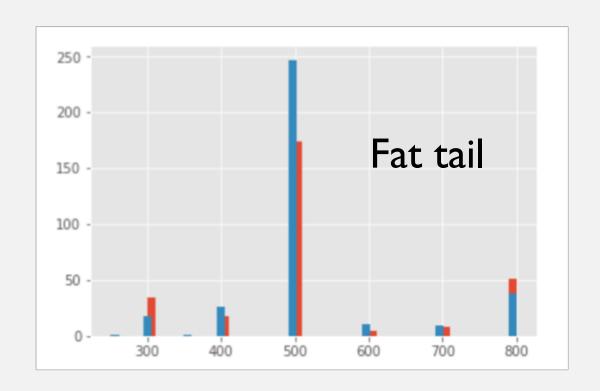


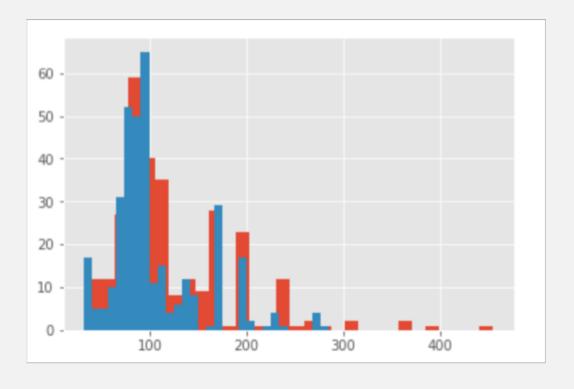


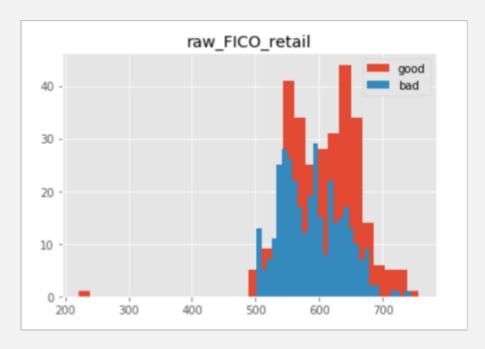
## AMOUNT\_APPROVED

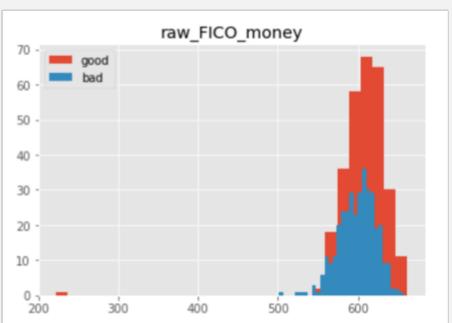
PAYMENT\_AMOUNT\_APPROVED

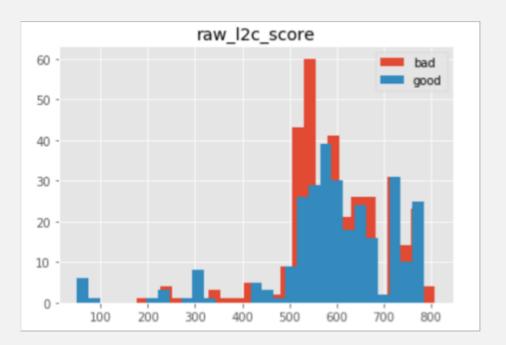
## "Loss distribution"

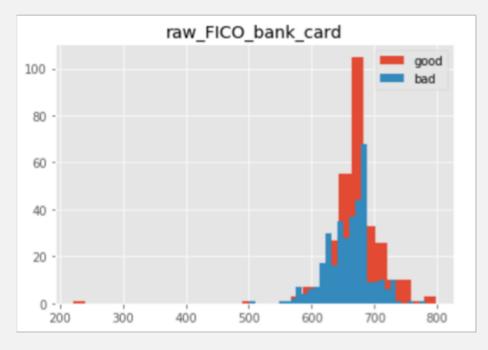












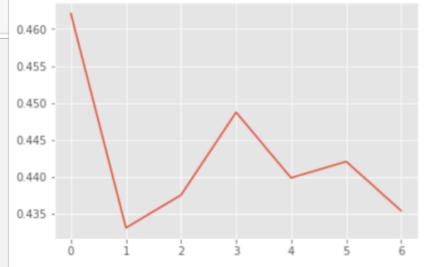
## LI-REGULARIZED LOGISTIC REGRESSION

```
# k-fold: choosing best hyperparameter
log_reg_Cs = [1, 2, 3, 4, 5, 6, 7]
CVs = []

for i in log_reg_Cs:
    tmp_lr = LogisticRegression(penalty = "l1", solver = "liblinear", max_iter=1000, C = i)
    tmp_cv = cross_val_score(tmp_lr, train[predictors], train["flgGood"], cv=4, scoring="accuracy")
    CVs.append(1-np.mean(tmp_cv))

plt.plot(CVs)|
plt.show()
```

Choose hyperparameter C=I



#### **EVALUATION**

#### Final model:

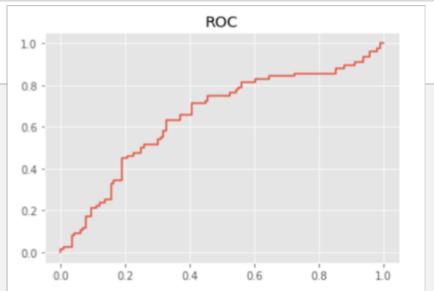
```
log_reg = LogisticRegression(penalty = "11", solver = "liblinear", max_iter=200, C = 1)
```

```
log_reg = LogisticRegression(penalty = "l1", solver = "liblinear", max_iter=200, C = 1)
log_reg.fit(train[predictors], train["flgGood"])
print("Accuracy: ", np.mean(log_reg.predict(test[predictors]) == test["flgGood"]))
fpr, tpr, thresholds = metrics.roc_curve(test["flgGood"], log_reg.predict_proba(test[predictors])[:, 1])
print("AUC: ", metrics.auc(fpr, tpr))
print("F-1 score: ", metrics.fl_score(test["flgGood"], log_reg.predict(test[predictors])))
```

Accuracy: 0.640625

AUC: 0.6493874773139745

F-1 score: 0.5660377358490565



### GRADIENT BOOSTING

```
{'max_depth': 7,
  'max_features': 15,
  'min_samples_leaf': 20,
  'min_samples_split': 60,
  'n_estimators': 50}
```

```
gsearch2.best_params_
{'learning_rate': 0.01, 'subsample': 0.6}
```

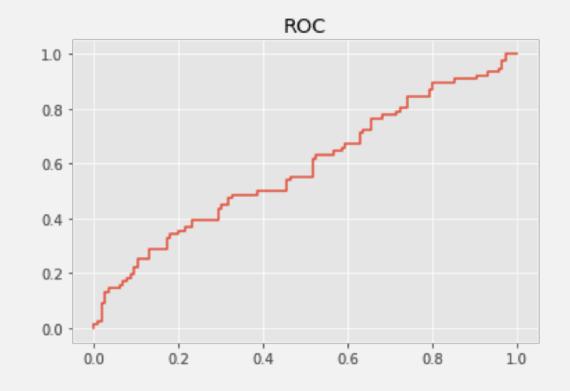
### **EVALUATION**

#### Final model:

Accuracy: 0.59895833333333334

AUC: 0.5835980036297641

F-1 score: 0.43795620437956206



## THANK YOU!