Credit Risk Classification

Shashin Sharan

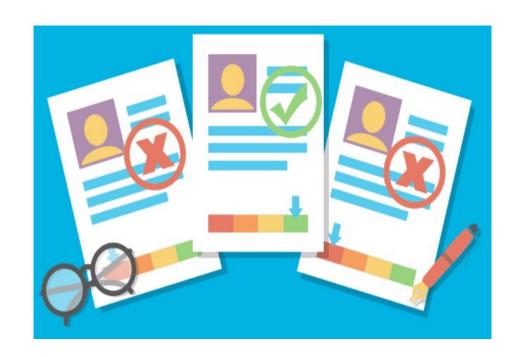
Problem Identification

Every year many customers approach banks or financial institutions for loan

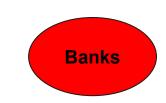


Problem Identification

Banks need to know credibility of customers before approving loans



Who might care







And many more....

What factors might affect?

- Number of Jobs
- Duration of loan
- Type of housing
- Purpose of loan
- Type of account
- Gender of customers



Data Information

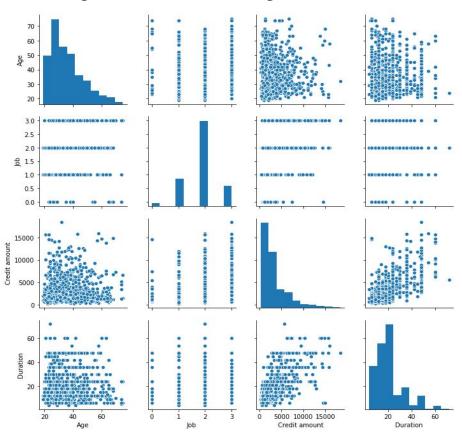
Source: UCI ML Repository

Number of customers: 1000

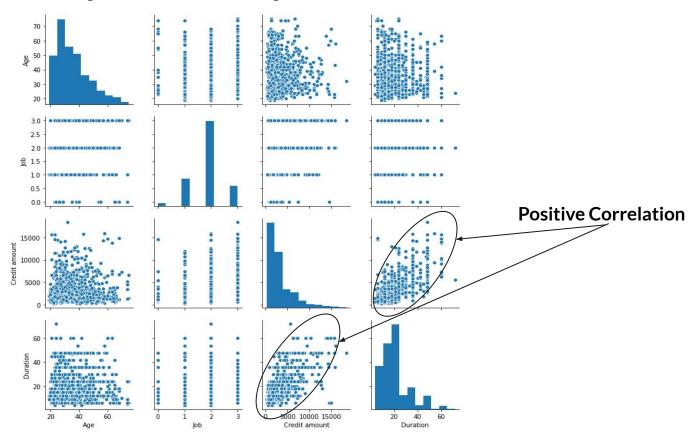
Number of fields per customer: 10

| Age | Sex | Job | Housing | Saving accounts | Checking account | Credit amount | Duration | Purpose | Risk |
|-----|--------|-----|---------|-----------------|------------------|---------------|----------|---------------------|------|
| 67 | male | 2 | own | No Account | little | 1169 | 6 | radio/TV | good |
| 22 | female | 2 | own | little | moderate | 5951 | 48 | radio/TV | bad |
| 49 | male | 1 | own | little | No Account | 2096 | 12 | education | good |
| 45 | male | 2 | free | little | little | 7882 | 42 | furniture/equipment | good |
| 53 | male | 2 | free | little | little | 4870 | 24 | car | bad |

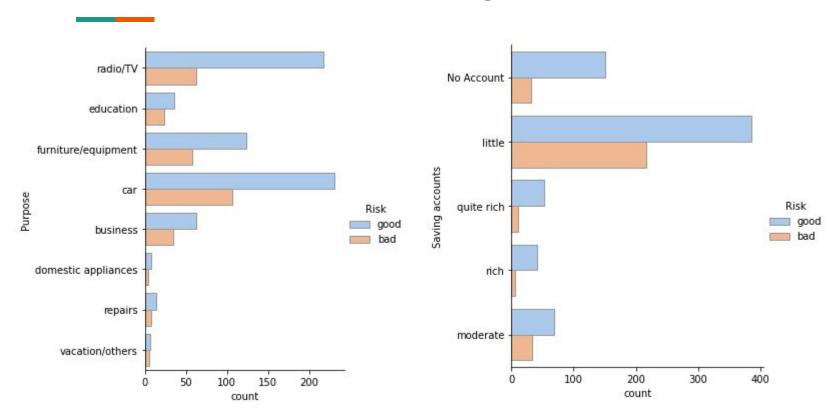
Exploratory Data Analysis



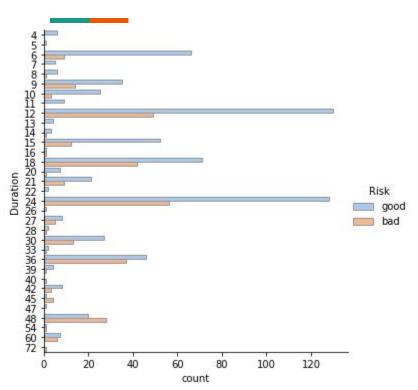
Exploratory Data Analysis

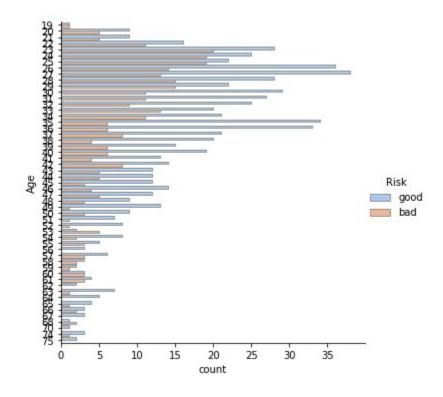


Different factors affecting Risk

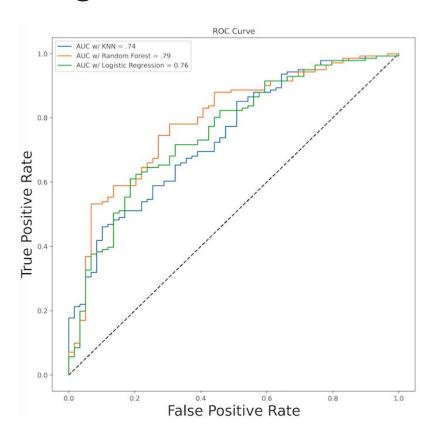


Different factors affecting Risk





Data Modeling



Data Modeling

| Model | Accuracy | AUC Score |
|---------------------|----------|-----------|
| KNN | .76 | .74 |
| Random Forest | .74 | .79 |
| Logistic Regression | .74 | .76 |

Recommendations and key findings

- Random Forest seems to outperform other models
- Currently accuracy is around 75%
- More data can improve the accuracy