#### Loan Default Prediction

Michal Topinka

Peking HSBC Business School 1602010585@sz.pku.edu.cn

April 20, 2017

### Data Source

- The data come from a three-year old competition on *Kaggle.com* hosted by the Imperial College London
- The goal was to determine whether a loan will default as well as to estimate the loss incurred if it does
- More specifically, they had to predict the loss for each row in the test set (next slide)
- The competition is closed now, winners were announced and the three people who were ITM had to expose their code

2 / 5

## **Data Description**

- The dataset correspond to a set of financial transactions associated with individuals
- The data have been standardized\*, de-trended, and anonymized. There are 769 features labeled  $f_1$  to  $f_{769}$
- 2 datasets:
  - Training set 105471 samples, including dependent variable 'loss'
  - Test set 210944 samples, without the variable 'loss'
- The 'loss' variable ranges from 0 to 100
  - 0 = no default
  - 70 = only 30% of the loan was reimbursed
  - 100 = 100% of the loan was not repaid

## **Project**

 Since the data for the 'loss' variable for the test set was not published, there is no way for me to check the accuracy when using the test set → I will only work with the training set

#### Project outline

- Data preprocessing (checking for duplicate rows, duplicated and constant columns, imputing NA values, etc.)
  - ullet  $\rightarrow$  removed 40 columns, 0 rows
- Feature Selection
- Feature Extraction
- Classification (Linear regression, SVM)
- Regression analysis to predict the size of the loss

# Methods (already implemented)

- Logistic Regression
- SVM (computational problems)
- Random Forrest Classifier