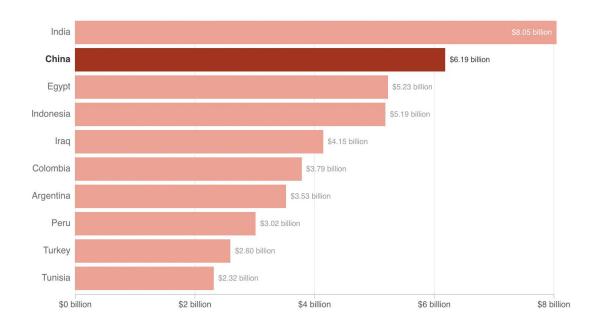
Team Bear Qarik Corporate Project

Kashif Bari, Kevin Bombardier, Attilio Castano, Bingjin Liu

The World Bank





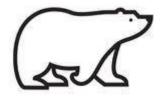
Loans given by the World Bank from 2016 to 2018

Our Goal





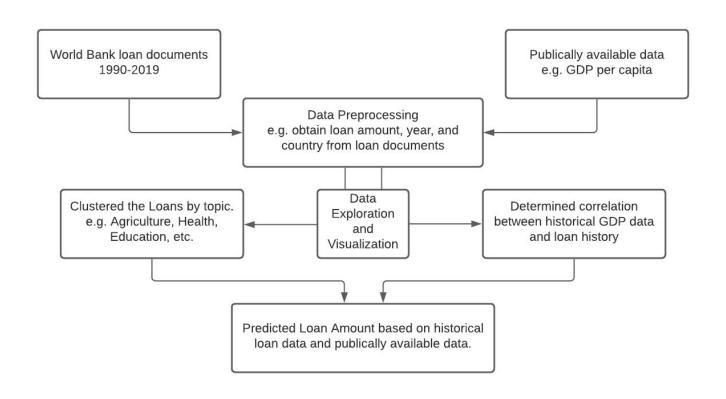




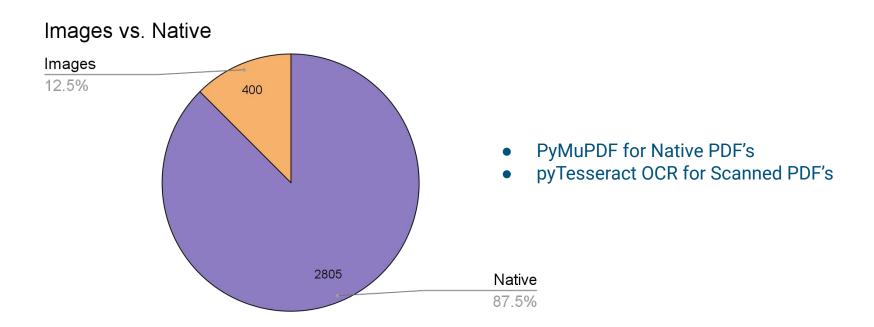




Project Execution Overview



Loan Documents: from pdf to text



Collect Publicly Available Data











GDP

Stability (Absence of Violence)

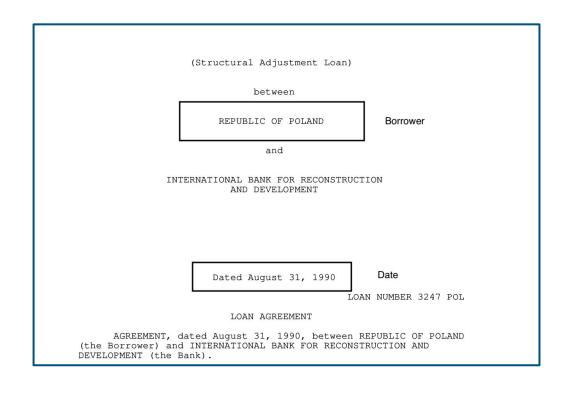
Literacy Rate

Electricity Usage

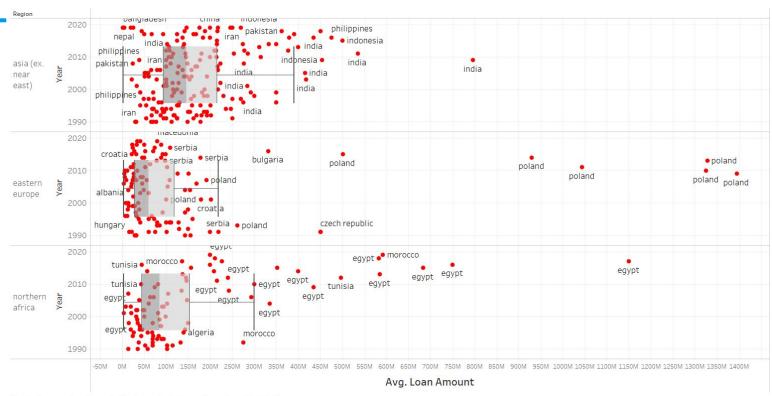
Income Inequality

Data Preprocessing

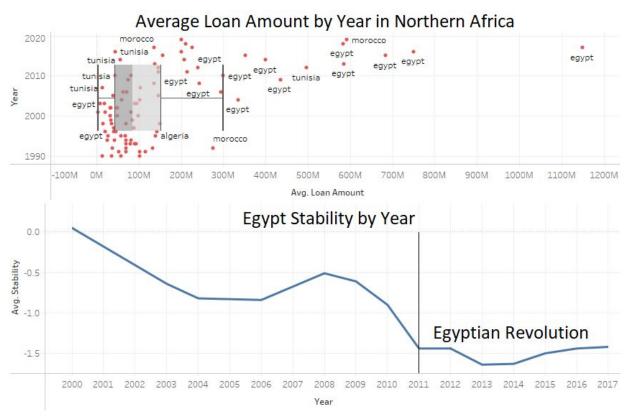
- Loan Amount
- Loan Date
- Borrower (Country)
- Project Name
- Project Description



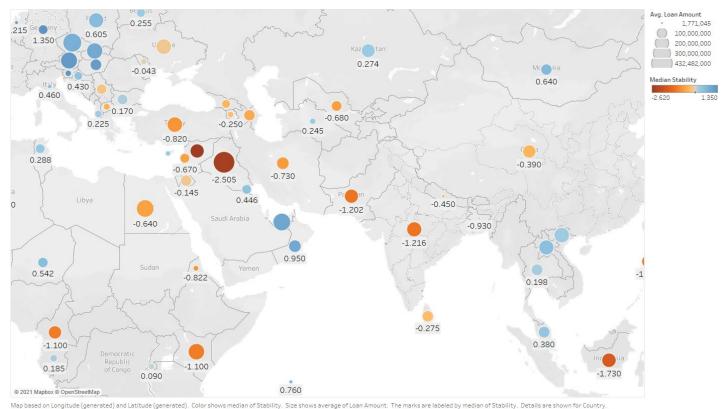
Average Loan Amount by Year and Region



Comparison of Loan Amount and Stability



Average Loan Amounts by Country and Stability



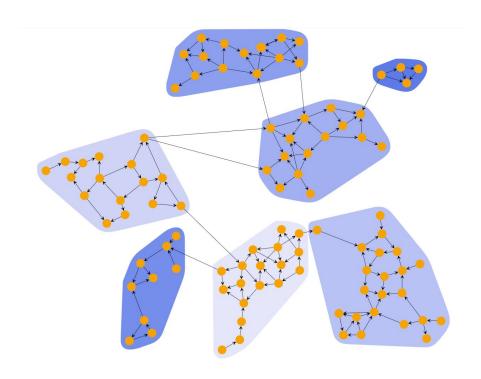
Two Approaches to Topic Analysis Clustering

Latent Dirichlet Allocation (LDA)

- Generative model
- 7 topics/clusters
- Evaluation of model
 - Topic Coherence: 0.586
 - o Perplexity: -6.718

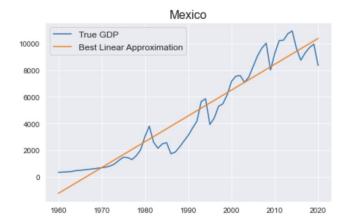
GloVe & KMeans

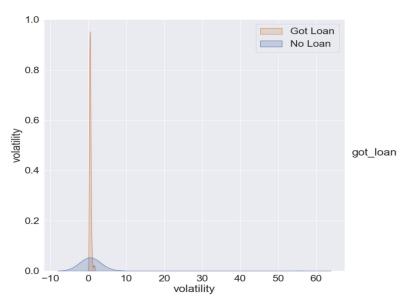
- GloVe: Transforms words into vectors
- KMeans: Clustering technique for vectors
- 11 topics/clusters
- Evaluation of model
 - Silhouette score: 0.112



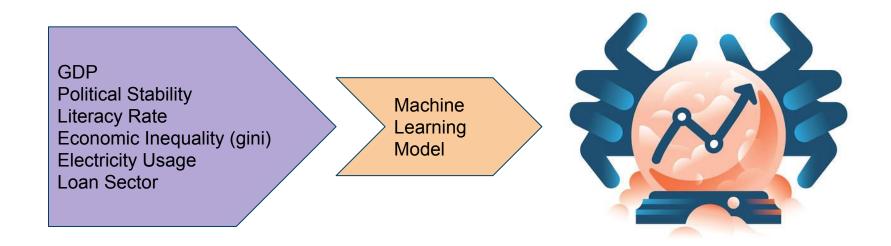
Predicting if a country got a loan from GDP Analysis

- Quantified correlation between
 - Historical GDP data
 - GDP Volatility = Normalized RMSE
 - Loan history
- Techniques
 - Linear Regression for feature eng.
 - Random Forest for classification.
- Evaluation of model
 - ROC AUC score ~0.75
 - Accuracy ~70%

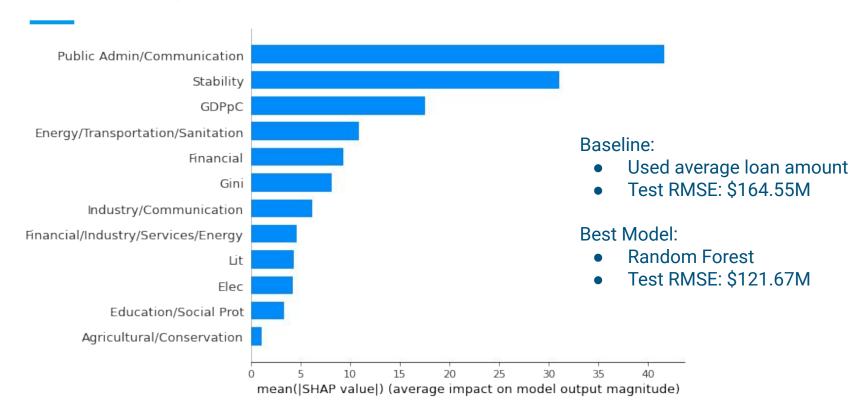




Predicting Loan Amount



Ranking of features by mean SHAP values



Conclusions

- Low GDP volatility is correlated to more loans.
- Political instability is not a barrier to apply for loans.
- Sector of the loan is correlated with the magnitude of the loan, as shown below.



For more details: https://github.com/DoctorBomb/Qarik-Team-Bear

Future Work

- Explore in more detail sector-specific trends.
- Obtain more data relevant to predicting loan amounts.

thank