

Group 19:

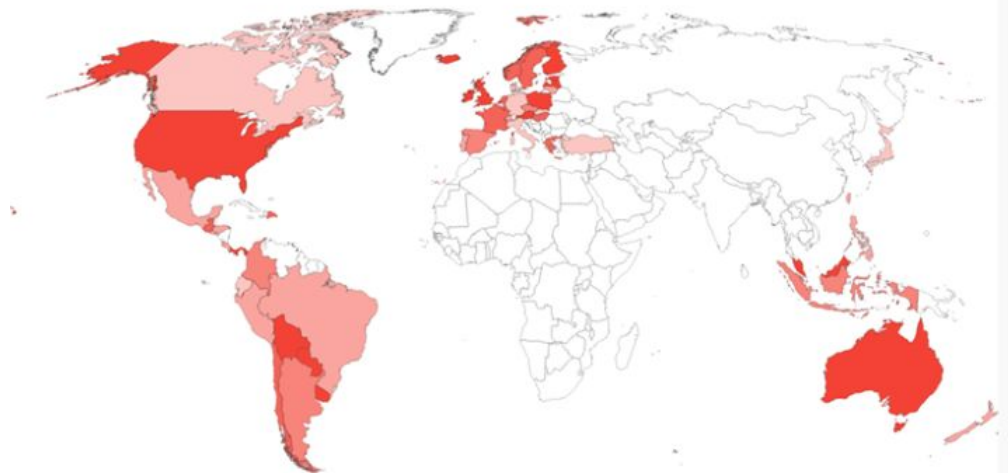
Song Success Prediction:

Ricky Pan, Martin Ziran Xu, Liam Shi, Deep
Mistry, Brian Long



Project Summary:

- **Song success prediction**
based on its musical features
in different regions of the
world.
- Current benchmarks:
 - **University of
Antwerpen:** 65% test
accuracy on overall
popularity



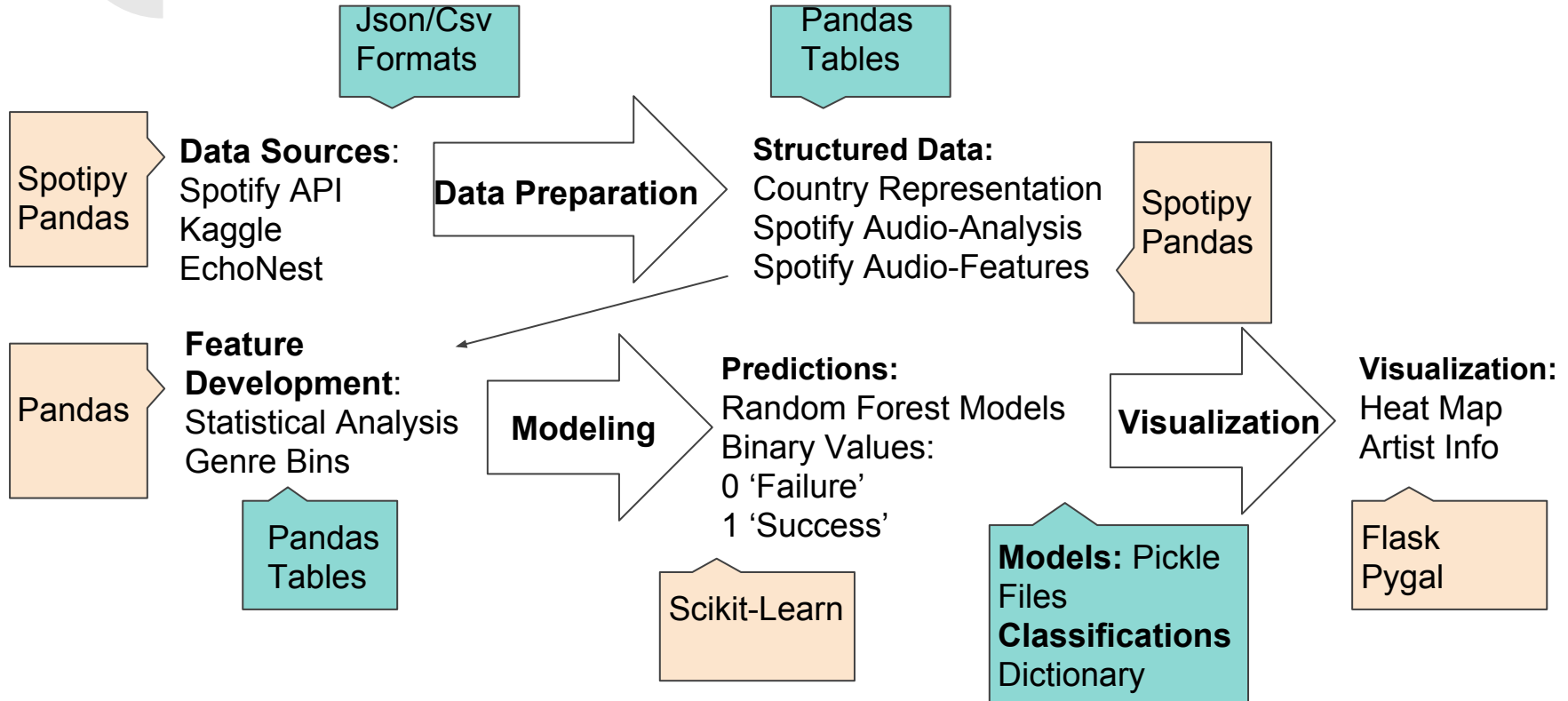


Product

Demonstration

[Link to GitHub containing all code!](#)

System Architecture





Learning Path and Biggest Challenges

- Regional output tags
- Cleaning and processing of big real world datasets
- Aggregating and balancing training set
- Feature extraction beyond spotify APIs standard features
- Working with various machine learning methods



Regional output

- Scraped top 50 regional charts on spotify
- Kaggle dataset tracked regional charts over one year
- Took unique songs and cross referenced them with the spotify API
- Defined binary tag for each region

550 Songs (Spotify)



72291 Songs (Kaggle)



1592 Songs (Subset)

```
['ee', 'br', 'my', 'cz', 'pe', 'gb', 'ie', 'se', 'sk', 'co', 'sg', 'ec', 'es', 'no', 'dk', 'be', 'lu', 'do', 'nz', 'de', 'hu',  
'cr', 'jp', 'hk', 'lt', 'pl', 'it', 'hn', 'mx', 'ph', 'cl', 'is', 'ca', 'ar', 'at', 'ch', 'tr', 'py', 'tw', 'gt', 'sv', 'us',  
'fr', 'global', 'nl', 'bo', 'lv', 'gr', 'pa', 'uy', 'au', 'pt', 'fi', 'popularity']
```

Feature extraction

Accousticness
Danceability
Duration_ms
Energy
Instrumentalness
Key
Liveness
Loudness
Mode
Speechiness
Tempo
Valence
Genre

Timbre Vector 1-12

Timbre: *The characteristic quality of sound that distinguishes one voice or musical instrument from another or one vowel sound from another: it is determined by the harmonics of the sound and is distinguished from the intensity and pitch.*

Extracted 120 Timbre Features

```
['acousticness', 'danceability', 'duration_ms', 'energy', 'instrumentalness', 'key', 'liveness', 'loudness', 'mode', 'speechiness', 'tempo', 'valence', 'mean_timbrel1', 'median_timbrel1', 'std_timbrel1', 'min_timbrel1', 'max_timbrel1', 'range_timbrel1', '80Percentile_timbrel1', 'kurtosis_timbrel1', 'skewness_timbrel1', 'mean_timbrel2', 'median_timbrel2', 'std_timbrel2', 'min_timbrel2', 'max_timbrel2', 'range_timbrel2', '80Percentile_timbrel2', 'kurtosis_timbrel2', 'skewness_timbrel2', 'mean_timbrel3', 'median_timbrel3', 'std_timbrel3', 'min_timbrel3', 'max_timbrel3', 'range_timbrel3', '80Percentile_timbrel3', 'kurtosis_timbrel3', 'skewness_timbrel3', 'mean_timbrel4', 'median_timbrel4', 'std_timbrel4', 'min_timbrel4', 'max_timbrel4', 'range_timbrel4', '80Percentile_timbrel4', 'kurtosis_timbrel4', 'skewness_timbrel4', 'mean_timbrel5', 'median_timbrel5', 'std_timbrel5', 'min_timbrel5', 'max_timbrel5', 'range_timbrel5', '80Percentile_timbrel5', 'kurtosis_timbrel5', 'skewness_timbrel5', 'mean_timbrel6', 'median_timbrel6', 'std_timbrel6', 'min_timbrel6', 'max_timbrel6', 'range_timbrel6', '80Percentile_timbrel6', 'kurtosis_timbrel6', 'skewness_timbrel6', 'mean_timbrel7', 'median_timbrel7', 'std_timbrel7', 'min_timbrel7', 'max_timbrel7', 'range_timbrel7', '80Percentile_timbrel7', 'kurtosis_timbrel7', 'skewness_timbrel7', 'mean_timbrel8', 'median_timbrel8', 'std_timbrel8', 'min_timbrel8', 'max_timbrel8', 'range_timbrel8', '80Percentile_timbrel8', 'kurtosis_timbrel8', 'skewness_timbrel8', 'mean_timbrel9', 'median_timbrel9', 'std_timbrel9', 'min_timbrel9', 'max_timbrel9', 'range_timbrel9', '80Percentile_timbrel9', 'kurtosis_timbrel9', 'skewness_timbrel9', 'mean_timbrel10', 'median_timbrel10', 'std_timbrel10', 'min_timbrel10', 'max_timbrel10', 'range_timbrel10', '80Percentile_timbrel10', 'kurtosis_timbrel10', 'skewness_timbrel10', 'mean_timbrel11', 'median_timbrel11', 'std_timbrel11', 'min_timbrel11', 'max_timbrel11', 'range_timbrel11', '80Percentile_timbrel11', 'kurtosis_timbrel11', 'skewness_timbrel11', 'mean_timbrel12', 'median_timbrel12', 'std_timbrel12', 'min_timbrel12', 'max_timbrel12', 'range_timbrel12', '80Percentile_timbrel12', 'kurtosis_timbrel12', 'skewness_timbrel12']
```

Spotify API



Machine Learning Methods

Input

- Feature importance
- Feature selection
- Normalizing
- Gap Creation
- Grouping into genres

Models

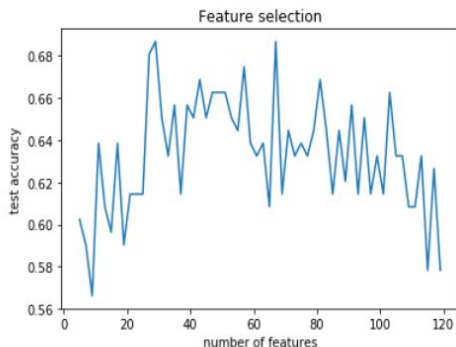
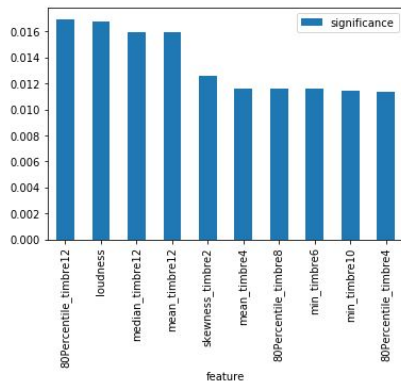
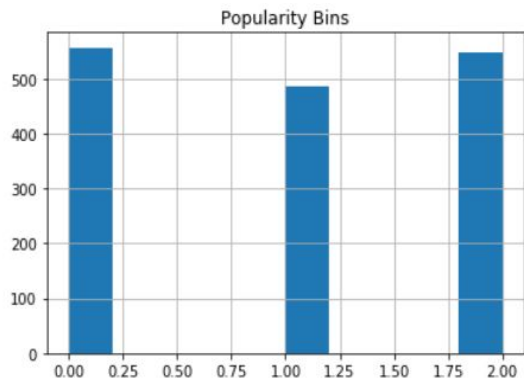
- Binary classification problem
- SVM, Random Forest, Gradient Boosting, Logistic Regression, KNN, Perceptron
- Hyperparameter tuning using stratified K-fold

Performance

- Training and test accuracy
- ROC curve and AUC score
- Confusion matrix



ML Performance on overall popularity



Random Forest: 67.50%
Test Accuracy
No timbre = 50%

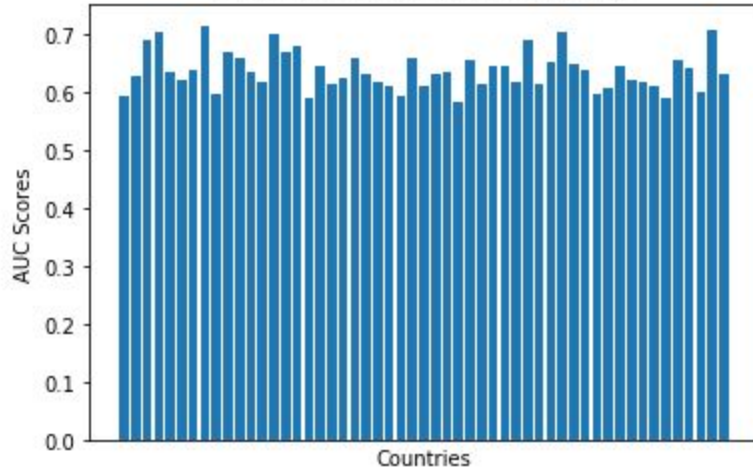
Top 10 Features:

```
[('loudness', 0.01786902630361211), ('80Percentile_timbre12', 0.01760271801219514), ('median_timbre12', 0.016583438399828395), ('mean_timbre12', 0.015389716219674653), ('80Percentile_timbre4', 0.01313153205058648), ('skewness_timbre2', 0.012189930767476579), ('80Percentile_timbre8', 0.011957885326952833), ('mean_timbre4', 0.011617575335665611), ('kurtosis_timbre7', 0.011146743936069552), ('min_timbre10', 0.01084267661855836)]
```

ML Performance on regional data



AUC Score for Different Countries

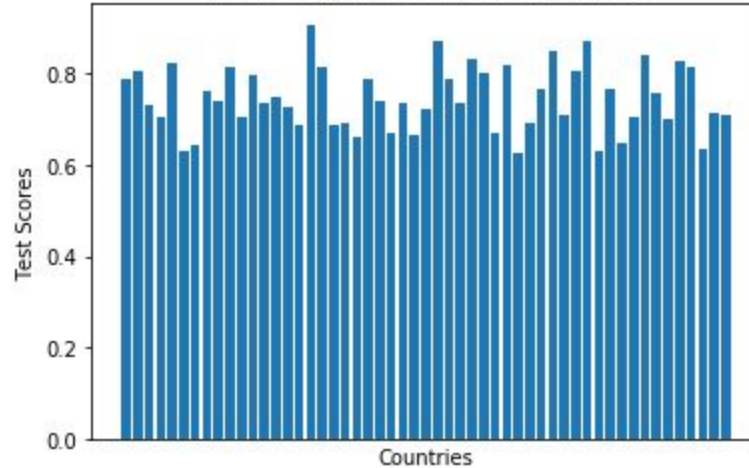


Average auc Score: 0.6391501585937819

Best auc Score: 0.7164389664389663

Worst auc Score: 0.5838422181878197

Test Accuracy for Different Countries



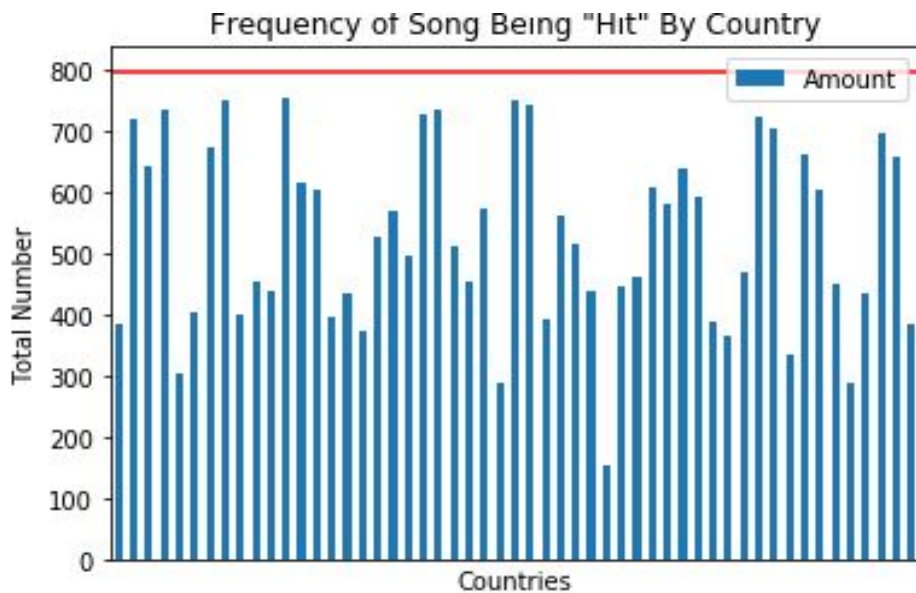
Average test accuracy: 0.7459540538406885

Best test accuracy: 0.9079497907949791

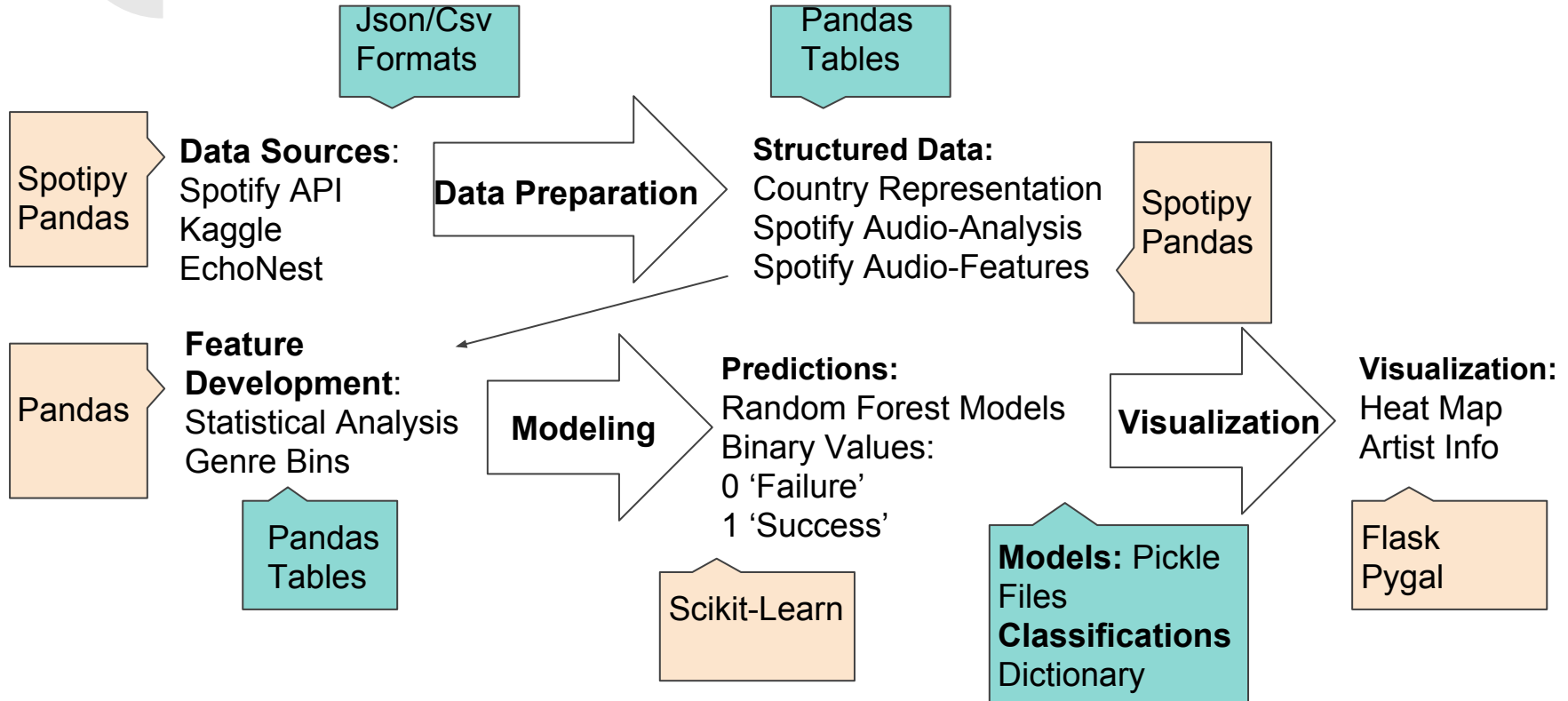
Worst test accuracy: 0.6276150627615062



Balanceness of dataset

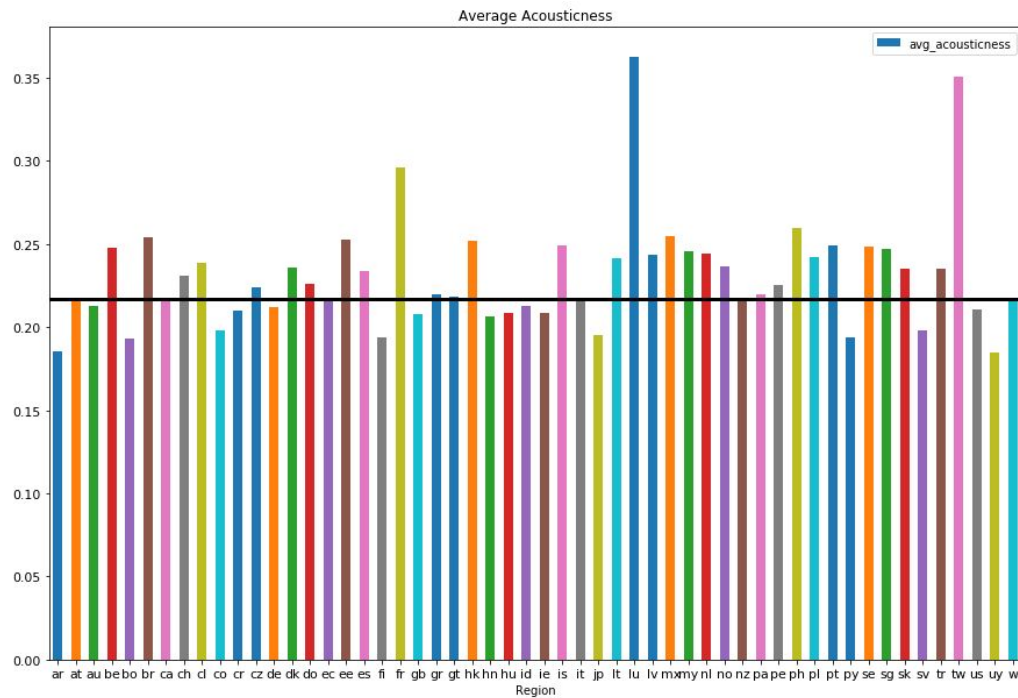


System Architecture



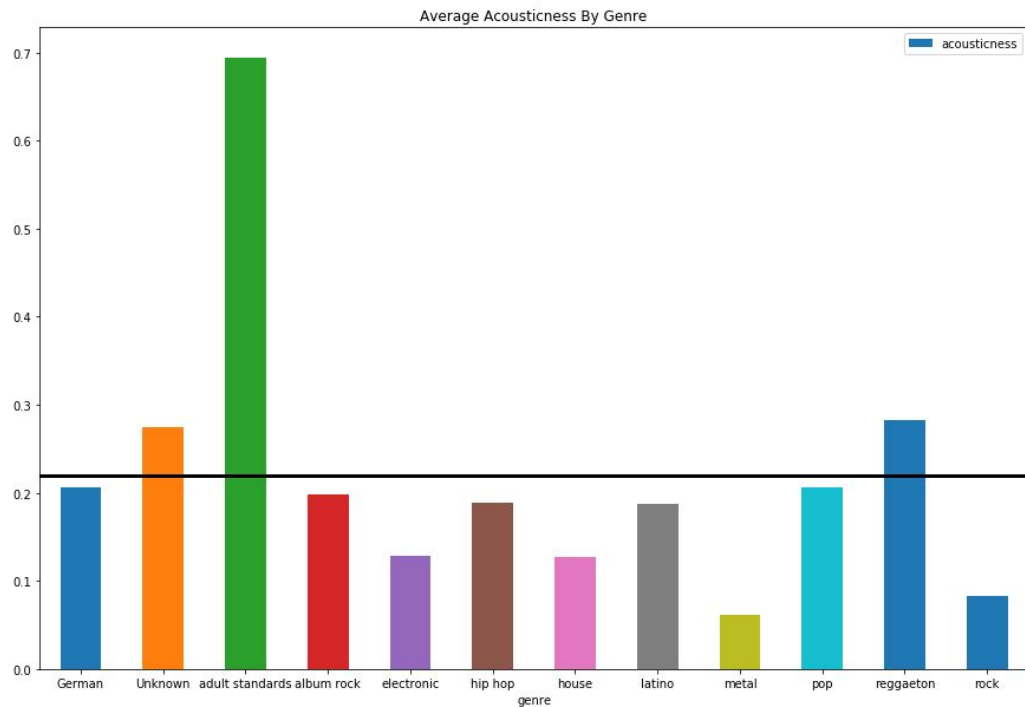


Data Exploratory



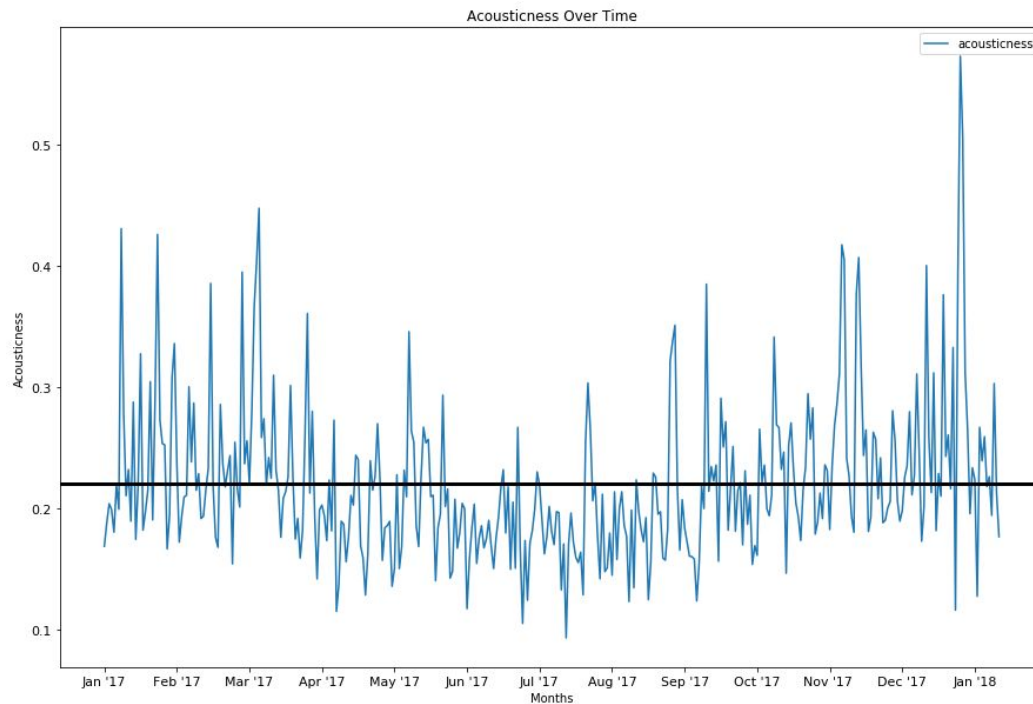


Data Exploratory





Data Exploratory





Future steps:

- Perform training on the whole dataset: 15x more songs
- Include meta-data as features: artist hotness, lyrics (bag of words)
- Train neural network with raw timbre vectors

Thank you!

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