

Predicting Popularity Of A Song using Spotify Data

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Data

- **Rows** - 174389 , **Features** – 18
- **Features** - ['key', 'mode', 'explicit', 'acousticness', 'danceability', 'duration_ms', 'energy', 'instrumentalness', 'liveness', 'loudness', 'speechiness', 'tempo', 'valence', 'year', 'artists' , 'release_date', 'name', 'id']
- **Target** - 'Popularity'

Question?

- Determine how well the features explain the popularity of songs on Spotify and Predict the popularity of a song based on these features

Pre-Processing

Categorical Features:

- Missing values
- One Hot Encoding

Continuous Features:

- Missing Values
- Standard Scaler

Algorithms

- **Ridge** - ['alpha','solver']
- **Extra Trees Regressor** - ['n_estimators', 'max_depth', 'min_samples_leaf', 'max_features', 'bootstrap']
- **Random Forest Regressor** - ['n_estimators', 'max_depth', 'min_samples_leaf', 'max_features', 'bootstrap']

Best Model

- RandomForestRegressor(max_depth=20, min_samples_leaf=5, n_estimators=30)

Interpretation

- R-Squared value - 0.68 and it implies that model explains 68% of the fitted data and that gives an indication that the features in the Spotify data could be used to interpret the popularity of the songs.

Summary

- Worked because the data set has diverse and predictive features.
- Useful in business setting when trying to understand which songs could become popular based on features such as tempo, danceability etc.,
- Could also be used further to recommend songs to users based on the common features in the songs they listen to.

Next Steps

- Improving the model fit by using other exhaustive methods like Grid Search.
- Finding if there is multicollinearity between the features and dropping those features.



Thank You

