MY 4 WEEKS PROGRESS REPORT

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Features:

- 1. Song Recommendation (minimalistic feature)
- 2. Recommendation on the basis of Genre and Year of Release (old or new)
- 3. Depicting the Importance of Acousticness, Loudness, Tempo, Liveness, danceability and valence using Feature Correlation
- 4. Used Spotify API to play songs on the WebApp
- 5. Redirecting to the recommended Songs on the Spotify app with one click.
- 6. Predicting the Popularity of songs after their recommendation using Linear Regression, Decision Tree Classifier, Support Vectorization, KNN and Random Forest
- 7. Calculated the Accuracy, F1 scores and AUC of each algorithm to see which is best for Prediction and Recommendation

Week 1: (4th May - 10th May)

Decided to explore all 3 given projects, and within 1-2 days finalized to work on Recommendation engine.

a) Product Building

- i. Researched about the shortcomings of recommendation engine.
- ii. Drafted how I am going to build it and which algorithms should be preferred while building this product.

Reason to choose recommendation engine -: I was recently building an Ecommerce platform and I was curious to know how are they able to recommend products so similar to our searches.

Within recommendation I decided to work on either music or job recommendation engine.

Week 2: (11th May – 17th May)

Finalized Music Recommendation Engine.

a) Gathering resources and Analyzing

- Read a few research papers on Recommendation Engine and their working.
 Link:
 - https://www.researchgate.net/publication/277714802 A Survey of Music Recommendation Systems and Future Perspectives
- ii. Analyzed product requirements, divided engine among various filtering methods and defined a clear data flow.

b) Searching Required Dataset

- i. Analyzed many dataset and worked upon required columns.
- ii. Finally allocated Spotify Gen Track Dataset and filtered it's content.

Week 3: (18th May - 24th May)

a) Training ML Model

- i. Configured the dataset and started with Training ML Model
- ii. Used KNN Algorithm prior and achieved Accuracy of 0.77633

b) Enhancing Algorithms for Recommendation and Prediction

- Trained the model using Random Forest Classifier and got great accuracy of 0.93573
- Decided to explore other Algorithms as well Decision Tree Classifier, Support
 Vector, Linear Regression and Support Vector.

The MOST efficient algorithm with 0.93573 accuracy on 0-1 scale was **RANDOM FOREST**

c) Building UI for deploying the ML Model

- i. Picked Streamlit for designing the Web App
- ii. Loaded ML model onto the page
- iii. Loaded custom widgets and graphs using streamlit Ploty, Express, Pydeck etc.

Week 4: (25th May - 29th May)

a) Graphs and Data Allocation

- i. Pushed graphs and Performance Summary Model.
- ii. Linked Spotify to the list of recommended songs.

b) Deployment and Github setup

i. Deployed the app on Streamlit Host

Link: https://share.streamlit.io/uditkatyal/songfitt /main/app.py

ii. To run and deploy in local, check for README.md file on Github.