Hitachi 2 User's Manual



About our Product

Our team has created a pipeline that takes a list of songs and their respective artists, and outputs a series of visualizations based around song metadata, pulled from Spotify's API, and song lyrics, pulled from Genius's API and interpreted through Pinecone. All of the code, data, and API pulling needed to produce our final visualizations can be achieved by following a few simple steps.

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Requirements

- Python IDE or Interpreter
 - We used Microsoft Visual Code, as it has compatibility with .ipynb files as well as standard .py files, but other IDEs will also work.
- lyricsgenius
 - o lyricsgenius is a Python library that has versatile tools for using the Genius API, and it can be directly pip installed to the terminal, as seen below:

C:\Users\Daniel Day>pip install lyricsgenius

- spotipy
 - o Spotipy is a Python library that helps harness spotify metadata. If your environment does not have spotipy pre-installed, you can also pip install it directly into the terminal:

C:\Users\Daniel Day>pip install spotipy

- Acceptable Format for Songs:
 - o CSV containing columns for song title and artist name
 - o Data from tsort.info, a website that charts top 100 billboard songs from every year.
- Code from GitHub repository
 - o Our code can be accessed from this link: https://github.com/darthjuju1/data_capstone-399

Data Extraction and Cleaning Pipeline

Extraction from CSV:

- In order to fetch the metadata for songs in a given CSV file, open run_df_fetcher.py.
- Replace variable given_df with your respective CSV file.
- Make sure that the generate_csv function has all of the variables set how you want them, though the default settings will output the product as desired. Below is the documentation for this function:

• Run the file.

Additional Explanation



This function will not create a copy of the csv by default, but if you set "create_copy" to True it will, in case you would like to inspect the data before merging. If you set this, be sure to use function merge_csv in oder to merge your copy back into the main dataframe.

Regular Expressions are also being run for all added lyrics and artists that clean the data for later usage in Pinecone.

• Genius's API will expire after around an hour. As a result, the function will only search for 55 minutes before stopping. This will typically gather data for 800-1000 songs, and the function can be run again after an hour.

Extraction from Website

- Open run_scraper.py
- Replace url with your address to website, and file_path with the desired output csv.



Notice

We currently only have the web scraper built around the website tsort.info, which has songs from the top 100 billboard charts on an annual basis dating back to the mid-20th century.

- Run the file.
- Like the previous function, Genius's access token will expire after around an hour, so this file will break the loop after 55 minutes. This will typically cycle through around 8-10 years, and the function can be run again after an hour.

End Result

You will now have a csv file with cleaned lyrics and a total of 9 fields called song_lyrics.csv. This will be used in the next part of the process: Pinecone.

Pinecone Vectorization

Index Creation:

• Used Pinecone's default code snippet for creating an index with integrated text embeddings through the llama-text-embed-v2 model.

Upserting:

Using a try/except block to skip over songs that failed to upsert, we were able to push our initial dataset of 8000 songs to our index's namespace over the course of 20 minutes. However due to some total of 7,752 upserted songs. We were able to develop a more efficient batch upsert function, for use with our final 12,500 song data frame.

Fetching:

Used Pinecone's query() to pull back vectors for semantic lookups—whether for search or similarity scoring. They were then saved to a CSV for analysis and visualization.

Data Visualization



Visualizations: What to Know

All of our visualizations can be generated by running a single file, but we will break down the details of each one below, as well as how to access them.

How to Access

- Open data_capstone_399/src/visualization/run_vis.py
- All visualizations rely on only two variables: df (DataFrame), and filepath. Fill in those two variables at the top of the file.

```
import vis_functions as vis
import pandas as pd

df = pd.read_csv("C:/Users/Daniel Day/Downloads/School/Data-399/data_capstone_399/data/Improved_Bucket_Accuracy_balanced_60s_up.csv")
filepath = "C:/Users/Daniel Day/Downloads/School/Data-399/data_capstone_399/src/Visualizations"

decades_list = vis.get_decades(df)
vis.decades_scatter(decades_list, filepath=filepath)
vis.decades_radial(decades_list, 'C:/Users/Daniel Day/Downloads/School/Data-399/data_capstone_399/Visualizations')
vis.plotly_scatter_explicit(df)
vis.plot_avg_intra_genre_cosine_by_decade(df, filepath=filepath)
vis.plot_cos_sim_bubble_chart(df, filepath=filepath)
vis.plot_decade_bar_chart(df, filepath=filepath)
vis.decades_genre_similarity(decades_list, filepath=filepath)
vis.decades_genre_similarity(decades_list, filepath=filepath)
```

• Run the file.

Your visualizations will now be saved as HTML files within your chosen directory! You are now free to explore your interactive graphs and charts. In the case of our team, we chose to make an interactive application and website:



User Interface and Website







