# **Assignment**

**Objective:** Find and compile a list of at least 100,000 music artists with their name, genre, profile picture, and location, store this information in a database, and build a search backend and interface that allows users to search for artists with auto-suggestions for highly matching names.

## Dataset selected - MusicBrainz Database

The MusicBrainz Database is built on the PostgreSQL relational database engine and contains all of MusicBrainz' music metadata. This data includes information about artists, release groups, releases, recordings, works, and labels, as well as the many relationships between them. The database also contains a full history of all the changes that the MusicBrainz community has made to the data.

Dataset Downloaded - <u>JSON Data dumps</u> Contained many relational tables including 'Artists'.

- 1) Extracted the tar.xz zip files
- 2) The original JSON file was very bulky 12.8 GB so much that it couldn't be even opened as a normal .json file. Hence used Python for Data processing.
- 3) Removed unnecessary fields and compiled into artist\_truncated.json

```
ARTIST
> mbdump
artist_api.py
{} artist_truncated.json
artists_data.csv
{} artists_data.json
{} artists_url_10k.json
{} artists url threading.json
R COPYING
■ JSON_DUMPS_SCHEMA_NUMBER
main.py
new.ipynb
README
■ REPLICATION SEQUENCE
■ SCHEMA_SEQUENCE
≡ TIMESTAMP
```

```
file_path = 'mbdump/artist.json'
truncated_file_path = 'artist_truncated.json'

c=0

in with open(file_path, 'r', encoding='utf-8') as input_file, open(truncated_file_path
for line in input_file:
    try:

    record = json.loads(line)

    output_file.write(json.dumps(record) + '\n')
    c+=1
    if c%10000==0:
        print(c,"/",500000)
    if(c>=500000):
        break
    except json.JSONDecodeError as e:

    v 4m 3.9s

... 10000 / 500000
    300000 / 500000
    300000 / 500000
    300000 / 500000
    500000 / 500000
    500000 / 500000
    600000 / 500000
    800000 / 500000
    800000 / 500000
    800000 / 500000
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    800000 / 5000000
    800000 / 5000000
    800000 / 5000000
    800000 / 5000000
```

- Cleaned the data, Removed duplicates and any entries that do not have 'Location' and compiled into artists\_data.json
- 5) Used Beautiful Soup to get the profile picture of artists. So, wikimedia (wikipedia) too has a massive database of artists that has a very specific and structured way of storing profile pictures of artists.

```
"https://en.wikipedia.org/wiki/{artist name.replace(' ', ' ')}"
```

- 6) Hence, querying over the artists names and making API requests by replacing the url with the artist name would give the profile picture. In some cases however, there was no profile picture found.
- 7) A major constraint would be **rate limiting**, as too many API requests would lead to IP blocking. Hence I used batch processing. Created **110 batches** where each batch would process 1000 artists with a time.sleep of 0.5 seconds. Used **multi-threading** with reduced worker count
- 8) The output file at this stage artists\_url\_threading.json contains an additional field called 'pp\_url' that contains the link to the profile picture. The final dataset has 1,10,000 artists
- 9) Finally, I used MongoDB as it is very convenient to work with React + Node with MongoDB especially with .json files

```
Processing Batch 89/110: 100%
                                          1000/1000 [00:11<00:00, 90.60it/s]
Processing Batch 90/110: 100%
                                          1000/1000 [00:00<00:00, 2897.85it/s]
Processing Batch 91/110: 100%
                                          1000/1000 [00:00<00:00, 4161.68it/s]
                                          1000/1000 [01:03<00:00, 15.70it/s]
Processing Batch 92/110: 100%
                                         1000/1000 [00:48<00:00, 20.51it/s]
Processing Batch 93/110: 100%
                                         1000/1000 [00:50<00:00, 19.96it/s]
Processing Batch 94/110: 100%
Processing Batch 95/110: 100%
                                          1000/1000 [00:43<00:00, 22.81it/s]
Processing Batch 96/110: 100%
                                          1000/1000 [00:55<00:00, 17.98it/s]
                                          1000/1000 [00:47<00:00, 20.90it/s]
Processing Batch 97/110: 100%
Processing Batch 98/110: 100%
                                         1000/1000 [00:32<00:00, 30.93it/s]
Processing Batch 99/110: 100%
                                        | 1000/1000 [00:36<00:00, 27.03it/s]
Processing Batch 100/110: 100%
                                          1000/1000 [00:40<00:00, 24.39it/s]
Processing Batch 101/110: 100%
                                           1000/1000 [00:36<00:00, 27.28it/s]
Processing Batch 102/110: 100%
                                          1000/1000 [00:40<00:00, 24.92it/s]
Processing Batch 103/110: 100%
                                           1000/1000 [00:35<00:00, 28.44it/s]
Processing Batch 104/110: 100%
                                           1000/1000 [00:38<00:00, 25.92it/s]
Processing Batch 105/110: 100%
                                           1000/1000 [00:45<00:00, 22.01it/s]
Processing Batch 106/110: 100%
                                           1000/1000 [00:43<00:00, 23.10it/s]
Processing Batch 107/110: 100%
                                           1000/1000 [00:43<00:00, 22.95it/s]
Processing Batch 108/110: 100%
                                           1000/1000 [01:00<00:00, 16.48it/s]
Processing Batch 109/110: 100%
                                           1000/1000 [00:39<00:00, 25.40it/s]
Processing Batch 110/110: 100%
                                           1000/1000 [00:51<00:00, 19.50it/s]
Updated dataset saved as 'artists url threading.json'
```

```
{
    "id": "49c43c49-328d-4b14-8a1d-be99cafaec14",
    "name": "Seeed",
    "genres": [
```

```
"count": 4,
    "disambiguation": "",
    "id": "6c3503e3-bae3-42de-9e4c-7861f2053fbe",
    "name": "dancehall"
},
{
    "name": "hip hop",
    "id": "52faa157-6bad-4d86-a0ab-d4dec7d2513c",
    "disambiguation": "",
    "count": 2
},
{
    "name": "reggae",
    "id": "02b2f720-d06e-42ce-85c2-lecd4191ffcb",
    "disambiguation": "",
    "count": 4
}
],
    "location": "Germany",
    "pp_url":
"https://upload.wikimedia.org/wikipedia/commons/thumb/9/9c/1_Live_Krone_2013_Seeed_2.jpg"
},
```

```
new > local > artists
  Documents 110.0K
                     Aggregations Schema Indexes 1
  _id: ObjectId('6708eb6ec6ff1d124f8b52bd')
         id: "49c43c49-328d-4b14-8a1d-be99cafaec14"
        name : "Seeed"
       ▼ genres : Array (3)
         ▼ 0: Object
            count : 4
             disambiguation: ""
            id: "6c3503e3-bae3-42de-9e4c-7861f2053fbe"
name: "dancehall"
         ▶ 2: Object
        location: "Germany
        pp_url: "https://upload.wikimedia.org/wikipedia/commons/thumb/9/9c/1_Live_Krone..."
        _id: ObjectId('6708eb6ec6ff1d124f8b52be')
id: "ea809832-657d-4335-b492-1daa3d46a322"
        name: "桜庭統"
        genres: null
        location: "Japan"
        pp_url : null
         _id: ObjectId('6708eb6ec6ff1d124f8b52bf')
         id: "7b9ada5f-8008-4dcf-bacd-a87e365a854a"
         name . UNiveleas Leevill
```

Further using MERN Stack I implemented the Search Task.

## **Displaying artists:**

Primary goal was to connect the backend with database and fire queries that would search for the artist name in the ison file. The artist details would be displayed.

#### Recommendations:

Secondly, I used the prefix tree (Trie DS) approach to give suggestions for search. A prefix tree is generated at the time of search. For example -

While searching for 'Coldplay'

- Initially we would have a tree with 37 nodes ( numbers from 0-9 and alphabets from A-Z)
- Now instead of parsing the entire database we would directly move to a node that contains 'C'.
- When the user enters C, that particular branch is selected eliminating a major portion of the dataset making it very efficient.
- Further the next alphabet is entered i.e. 'o'. So the second level tree having the branch containing 'o' is selected.
- In the similar I have allowed the algorithm to move up to 4 alphabets (as it is sufficient)

So when C is entered, all the words most related to C are suggested, which is rather vague But when Co is entered, there is a great chance that the word we want would already appear

# Overcoming drawback of Trie DS

One major drawback of the prefix tree would be dependence on the preceding word. If I write Boldplay instead of Coldplay, most ideally I should be able to see Coldplay by suggestion but with the prefix tree I won't be able to. Hence we can use the Fuzzy logic by Fuse.js

Fuse.js is a lightweight, JavaScript-based fuzzy-search library that helps perform fast and efficient searches, especially when working with a large amount of text data. Unlike standard search algorithms, which rely on exact matches, Fuse.js enables you to perform "fuzzy" searches—meaning it can find approximate matches even when the search input is misspelled or incomplete.

One key method Fuse.js uses is "Levenshtein distance," which is a metric for the minimum number of edits needed to turn one string into another. If you type "Boldplay," Fuse.js will calculate the distance between this and other artists in the dataset, determining that "Coldplay" is the closest.

## **Abbreviations**

Simply convert all the words into abbreviations and store them along with the Trie. Add those to fuzzy logic as well











Boldplay





MJ



