# WSM-Project3

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## TOC

- 1. TFIDF
- 2. ItemCF
- 3. N-Gram
- 4. Discussions & Feedbacks

# **TFIDF**

### **TFIDF**

Port project1's code

```
# Build dimension dictionary, indicates key's index
    index = 0
     for word in self.parser.parse(" ".join([doc["doc"] for doc in docs]), language):
      if word not in self.dimensions:
         self.dimensions[word] = index
        index += 1
     self.tf = np.zeros((len(docs), len(self.dimensions)))
 8
     # Count tf, df, idf, and tfidf by using numpy functions, and store them for further usage.
     for i, doc in enumerate(tqdm.tqdm(docs, postfix="Building Index")):
10
       processedDoc = self.parser.parse(doc["doc"], language)
11
      doc["parsed"] = processedDoc
12
      for word in processedDoc:
13
         self.tf[i][self.dimensions[word]] += 1
14
```

14k sessions \* 1m songs = 140,000,000 ints / shorts / bit

### Concept

Predict user preferences based on user's historical behavior, instead of attributes of the item.

#### **Formula**

• Basic

$$W_{\mu v} = rac{\mid N(\mu) \cap N(v) \mid}{\mid N(\mu) \mid}$$

Adding penalty term

$$W_{\mu v} = rac{\mid N(\mu) \cap N(v) \mid}{\sqrt{\mid N(\mu) \mid\mid N(v) \mid}}$$

### **Implementation**

Use dict to create co-occurance matrix.

```
      Song1 Song2 ... Song10000

      Song1 1000 512 ... 201

      Song2 512 1 ... 345

      ... ... ...

      Song10000 201 345 ... 2000
```

- Do cosine similarity search based on the matrix
- Randomly fill with top 20 songs if recommendation less than 5 songs.
- Basic Score: 0.14537

### **Consider Listening Time**

- Give weights based on listening time.
- Use unix\_played\_at to get listening time.

```
song1 <-diff-> song2 <-diff-> song3
```

- Our weight:
  - < 10 sec: 0.1
  - 11 ~ 150 sec: 0.7
  - > 150 sec: 1
- Score: 0.10981 (worse)

### **Consider Repeat Time**

- If last 3 songs are the same song or more than 8 repeated songs among top 20 songs
  - Recommend the same song 5 times
- **Score: 0.12708** (worse)

#### Conclusion

#### **Listening Time**

- < 10 sec: 0.1, 11 ~ 150 sec: 0.7, > 150 sec: 1 => **Score: 0.10981**
- < 10 sec: 0.1, > 11 sec: 1 => **Score: 0.115**
- Predict what KKbox predicts, instead of what the user would like.

#### Repeat Time

- more than 8 repeated songs among top 20 songs => Score: 0.12708
- more than 20 repeated songs among top 20 songs => Score: 0.12755
- Rating Criteria: Recommend a variety of unique songs get higher scores.

### **Prepare Data**

• Join tables together for furthur usage

```
andyjjrt > ~/c/wsm-project3 > python main.py
                                                                                                        wsm-project3 Py < andyjjrt@nccucs108 <
                                                                                                                                                master ?8
                        song id unix played at
          session id
                                                                    lid
                                                                                                       pid
                                                                                                                               title text id
                       s 354122
                                      1660012505
                                                                                                            c1079ef109db2aba72f78c632ab73803
                                                                    NaN
                      s 1030665
                                     1660012730
                                                                    NaN
                                     1660015113
                                                                                                            f9b7f48dbd07a9979e64ccf88af181aa
                       s 642781
                                                               l 440385
                                                                                                  p 288113
                       s 280722
                                     1660015289
                                                               l 169111
                                                                                                            c1079ef109db2aba72f78c632ab73803
                        s 90294
                                     1660015841
                                                               l 196433
                                                                                                            9f25d97515a9e19da4c7ad6dba6d8776
                                     1664707185
                                                                    NaN
14306470
              715323
                       s 355651
                                                                                                            0025132fbe92679e3472ba869b60af35
                       s 170450
                                     1664707201
                                                                         p 71934.p 72220.p 72837.p 314936
14306471
              715323
                                                                                                            c1079ef109db2aba72f78c632ab73803
14306472
              715323
                       s 233844
                                     1664707400
                                                       l 29174.l 257878
                                                                                                            5bb36dbeb12e9e4149eed8166a60fcf8
14306473
                       s 931390
                                     1664707449
                                                               l 284779
              715323
                                                                                                            859a2ae6d55da7cdfc57bc85d2008a99
14306474
              715323
                       s 935948
                                     1664707668 ...
                                                               l 294002
                                                                                                            0d42e410deb1c569568ba132e738d6aa
```

Trigram

```
1  n = 3
2  train_data, padded_sents = padded_everygram_pipeline(n, allWords)
```

#### model.score

```
result = list(model.score(words[-2:]))
result = [r for r in result if r[0] != "<s>" and r[0] != "</s>"]
result.sort(key=lambda x: x[1])
length = len(result)
return [session_id] + result[:5]
```

```
20 [|||||||100.0%]
         ||||||||98.7%]
                                                                                ||||||100.0%] 21 [|||||||||||||||||||||||||||||||||100.0%]
   11 [||||||||||
                                                            Tasks: 411, 2087 thr; 23 running
                                                            Load average: 22.66 23.00 23.11
                                                            Uptime: 3 days, 16:19:16
           PRI NI VIRT RES SHR S CPU% MEM% TIME+ Command
2985048 andyjjrt   20   0 7184M 2322M 17476 R 101.   2.4 11h27:41 python main.py
2985041 andyjjrt    20    0  7184M 2322M 17552 R 100.    2.4 <mark>11h</mark>27:49 python main.py
2985043 andyjjrt    20    0  7184M 2322M 17552 R 100.    2.4  11h27:45  python main.py
2985049 andyjirt    20    0  7184M 2322M 17408 R 100.    2.4  11h27:48 python main.py
2985031 andyjjrt    20    0  7184M 2322M 17552 R 99.8   2.4 <mark>11</mark>h27:47 python main.py
2985032 andyjjrt   20   0  7184M 2322M 17552 R 99.8   2.4 <mark>11</mark>h27:48 python main.py
2985033 andyjjrt    20    0  7184M 2322M 17552 R 99.8   2.4  11h27:46 python main.py
2985034 andyjjrt    20    0  7184M 2322M 17552 R 99.8   2.4  11h27:46 python main.py
```

• 12hr per run with 22 threads, **Basic Score: 0.20894** 

#### model.count

```
if words[-1] == words[-2]: return [session_id] + [songIdMapping[words[-1]]] * 5

result = list(model.counts[words[-2:]].items())

result = [r for r in result if r[0] != "<s>" and r[0] != "</s>"]

result.sort(key=lambda x: x[1])

length = len(result)

if length >= 5:

return [session_id] + [songIdMapping[r[0]] for r in result[:5]]

else:

tmp = [songIdMapping[r[0]] for r in result]

for i in range(5 - len(tmp)):

tmp.append(random.choice(list(songIdMapping.items()))[1])

return [session_id] + tmp
```

• 6hr per run with 22 threads, 30% replacement to **0.22006** 

#### What we should do

- Randomly choose songs from same genre, titletext, etc.
- Use model.generate() instead.
- Try 4-gram.
- Reciprocal Rank Fusion

$$RRFScore(d \in D) = \sum_{r \in R} rac{1}{k + r(d)}, k = 60$$

• Data validation (important).

# Discussions & Feedbacks

### **Discussions & Feedbacks**

Data validation

```
PROBLEMS 1 OUTPUT DEBUG CONSOLE PORTS TERMINAL

andyjjrt ~/codes/wsm-project3 python main.py

13%|
17
13%|
andyjjrt ~/codes/wsm-project3
```

```
for i in tqdm.tqdm(range(recordLength)):
  words = list()
  for j in range(25):
    words.append(recordTable.loc[i*25 + j, "song_id"])
  allWords.append(words)
```

# Thanks for listening