Simple, Open, Music Recommendations with Python

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FOSDEM 2023

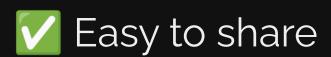
About me

- Systems software developer @ Codethink
- Musician and music fan
- Former teacher

Playlists (1990s)



X Difficult to make



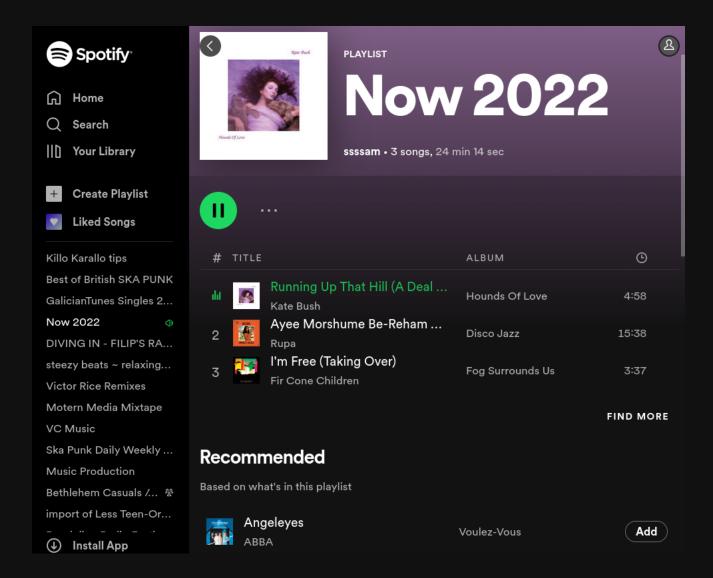
Playlists (2000s)







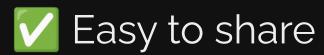
Playlists (2010s)



Playlists (2010s)







🔖 Can generate the playlist for you

Spotify philosophy

- Grow as big as possible ("blitzscaling")
- Pay artists as little as possible
- Optimize for passive listener engagement
- Apply user surveillance and machine-learning to every problem
- All hail the Algorithm

What would the opposite look like?

- Not for profit / DIY
- Encourage building a local music collection
- Link to artist-controlled websites
- Work with open data

Let's get experimenting!



What can we learn from ... Dynamicland?



What can we learn from **Git**?



Git's core ideas were implemented in a month

- 1. Well-defined data model: blobs, trees, commits, refs.
- 2. Multi-call binary: small programs that work together
- 3. "Porcelain" and "Plumbing" layers

Git's design allows...

- a "polyglot" codebase
- easy extensions
- popular websites built around it

Calliope: the same principle for playlists.

- Data model: everything is a playlist
- Multi-call binary cpe (also has a Python API)
- Build recommendation pipelines as shell pipelines
- Optimized for ease of maintenance over ease of use.

pip3 install calliope-music

Core data model

Playlist item

```
{ "creator": "Artist 1", "title": "Great Song" }
```

Playlist

```
{ "creator": "Artist 1", "title": "Great Song" }
{ "creator": "Artist 2", "title": "Banging Tune" }
{ "creator": "Artist 3", "title": "Unpleasant Noise" }
```

This is JSON Lines data so it can be processed one line at a time.

...based on XPSF

What is XSPF?

- A playlist format like M3U
- XML like RSS
- · Pronounced spiff
- MIME type application/xspf+xml

What does XSPF look like?

A very simple document looks like this:

Calliope's playlist format is documented here.

Demo: playlist manipulation

```
{ "creator": "Artist 1", "title": "Great Song" }
{ "creator": "Artist 2", "title": "Banging Tune" }
{ "creator": "Artist 3", "title": "Unpleasant Noise" }
```

- Shuffle: cpe shuffle
- Export: cpe export
- Line-based shell processing
- Data-oriented shell processing

What's next?

Content resolution

XSPF is an intermediate format. We expected a new kind of software called a **content resolver** to do the job of converting XSPF to a plain old list of files or URIs.

-- XSPF spec

Demo: content resolution

Three songs:

```
{"creator": "Kate Bush", "title": "Hounds of Love"}
{"creator": "Madonna", "title": "Holiday"}
{"creator": "Ana Frango Elétrico", "title": "Saudade"}
```

- Resolve locally: cpe tracker resolve-content
- Resolve remotely: cpe spotify resolve-content

What's next?

Recommendations

big playlist \rightarrow algorithm \rightarrow small playlist

Case study: Special Mix

Special Mix generates a 1 hour playlist of discoveries from a specific year.

python3 -m calliope_examples.special_mix

Ingredients:

- 1. Listening history: pylistenbrainz
- 2. Content resolution: beets
- 3. Track selection: simpleai

1. Listening history

- Use Listenbrainz to track music you listen to
- Use Web Scrobbler browser extension to submit listens
- Use pylistenbrainz and cpe listenbrainz to access the data



1. Listening history

cpe listenbrainz listens

```
> cpe listenbrainz-history --user samthursfield2 listens \
     from json --objects | first
Updating listens from Listenbrainz server
  listenbrainz.listened_at
                                 1675368832
  listenbrainz.recording_msid
                                 306525cd-74d3-4acb-b292-8bf300ba6
  listenbrainz.artist_msid
  listenbrainz.release_msid
  creator
                                 Knobs
  title
                                WIW
  album
                                Stipple
                                 https://knobs.bandcamp.com/album/
  listenbrainz.origin_url
```



1. Listening history

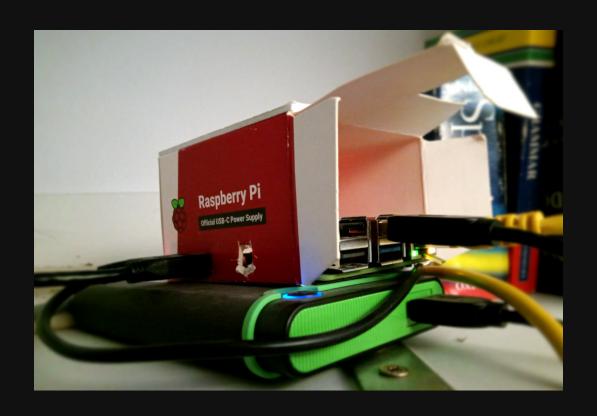
cpe listenbrainz-history --no-sync --user samthursfield2 \
histogram --bucket year | from json | last 5

| # | bucket | count |
|---|---------------------|-------|
| 0 | 2019-01-01 00:00:00 | 6014 |
| 1 | 2020-01-01 00:00:00 | 5990 |
| 2 | 2021-01-01 00:00:00 | 4239 |
| 3 | 2022-01-01 00:00:00 | 6721 |
| 4 | 2023-01-01 00:00:00 | 208 |

...choose a year, select by first_listen_date: now we have a playlist



2. Content resolution



2. Content resolution

Beets is the media library management system for obsessive music geeks.

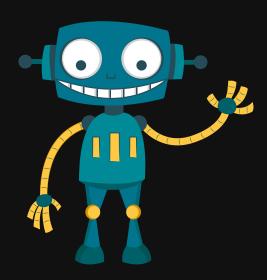
Content resolvers are pluggable and Special Mix can use any...

...now we have a playlist with track URLs and durations.

3. Track selection

The cpe select module wraps the Python simpleai package.

You define **constraints** for the playlist, then run a **local search** algorithm to try and find a suitable combination of tracks.



No neural network required.

3. Track selection

Music playlist generation by adapted simulated annealing

Steffen Pauws, Wim Verhaegh, Mark Vossen 1

Philips Research, Prof. Holstlaan 4, 5656 AA Eindhoven, The Netherlands

Abstract

We present the design of an algorithm for use in an interactive music system that automatically generates music playlists that fit the music preferences of a user. To this end, we introduce a formal model, define the problem of automatic playlist gen-

3. Track selection

Table 4. Constraint set 'typical'.

| desemintion | constraint |
|---------------------------|---|
| description | |
| All different songs | $pairs-global(1, n_{max}, 1, d(v) = \{x \mid x \neq v\})$ |
| Release in $1980-2001$ | each-global $(1, n_{\text{max}}, 7, [1980, 2001])$ |
| $\geq 20\%$ Stevie Wonder | fraction-global $(1, n_{\text{max}}, 3, \{\text{Stevie Wonder}\}, .2, 1)$ |
| $\geq 10\%$ Seal | fraction-global $(1, n_{\text{max}}, 3, \{\text{Seal}\}, .1, 1)$ |
| $\geq 10\%$ Peter Gabriel | fraction-global $(1, n_{\text{max}}, 3, \{\text{Peter Gabriel}\}, .1, 1)$ |
| $\geq 10\%$ Janet Jackson | fraction-global $(1, n_{\text{max}}, 3, \{\text{Janet Jackson}\}, .1, 1)$ |
| $\geq 10\%$ Mariah Carey | fraction-global $(1, n_{\text{max}}, 3, \{\text{Mariah Carey}\}, .1, 1)$ |
| $\geq 20\%$ Phil Collins | fraction-global $(1, n_{\text{max}}, 3, \{\text{Phil Collins}\}, .2, 1)$ |
| $\geq 40\% \text{ R\&B}$ | fraction-global $(1, n_{\text{max}}, 5, \{R\&B\}, .4, 1)$ |
| $\geq 40\%$ Popular | fraction-global $(1, n_{\text{max}}, 5, \{\text{Popular}\}, .4, 1)$ |
| 2-3 different genres | cardinality-global $(1, n_{\text{max}}, 5, 2, 3)$ |
| Different succ. genres | $chain-global(1, n_{\max}, 5, d(v) = \{x \mid x \neq v\})$ |
| Similar succ. tempi | chain-global $(1, n_{\text{max}}, 8, d(v) = \{x \mid sim(x, v) \in [0, 0.1]\})$ |

Each constraint defines a *function* to score a playlist from 0 to 1.

cpe select searches for the playlist with the highest score given the constraints.

Using **local search** to find a solution

Example:

- All songs must be 2 to 4 minutes long.
- The playlist must be 10 minutes long.

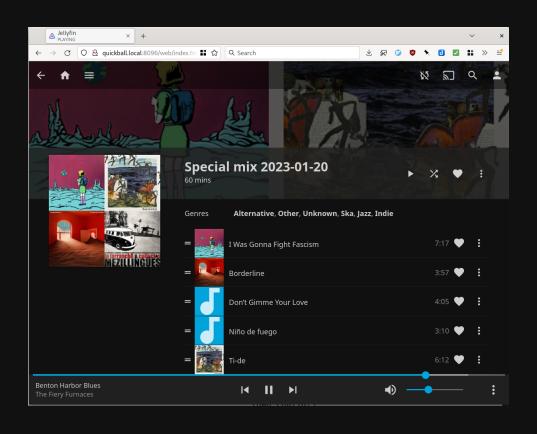
```
from simpleai.search.viewers import ConsoleViewer, WebViewer
from calliope.playlist import Playlist, PlaylistItem
from calliope.select import ItemDurationConstraint, PlaylistDurationConstraint
import calliope.playlist, calliope.select, calliope.shuffle
import sys
MINUTES = 60
constraints = [
    ItemDurationConstraint(vmin=2 * MINUTES, vmax=4 * MINUTES),
    PlaylistDurationConstraint(vmin=10 * MINUTES, vmax=10 * MINUTES),
corpus = Playlist([
    PlaylistItem({"calliope.id": """, "title": "Amazing Tune", "duration": 2 * MINUTES}),
    PlaylistItem({"calliope.id": "

"
", "title": "Punk Classic", "duration": 1 * MINUTES}),
    PlaylistItem({"calliope.id": "♬", "title": "Lengthy Opus", "duration": 12 * MINUTES}),
    PlaylistItem({"calliope.id": "", "title": "Ambient Noise", "duration": 7 * MINUTES}),
])
viewer = WebViewer()
input_playlist = calliope.shuffle.shuffle(corpus)
output_playlist = calliope.select.select(input_playlist, constraints, viewer=viewer)
calliope.playlist.write(output_playlist, sys.stdout)
sys.stderr.write(f"Total duration: {sum(item['duration'] for item in output_playlist)}\n")
```

Export to music player

```
> head 'Special mix 2023-01-20.m3u'
#EXTM3U
#PLAYLIST:Discoveries of 2020
../../Music/Soccer96 - Tactics EP [2020]/01 I Was Gonna Fight Fas
../../Music/Tame Impala - The Slow Rush [2020]/03 Borderline.mp3
../../Music/Vic Ruggiero - On the Ragtime [2009]/09 Don't Gimme Y
../../Music/Echte Übersee Records_ Finest Latino Ska and Punk Fro
../../Music/KOKOROKO - KOKOROKO [2019]/02 Ti-de.mp3
```

Export to music player



Recap: Special Mix

python3 -m calliope_examples.special_mix

Ingredients:

- 1. Listening history: pylistenbrainz
- 2. Local music collection: beets
- 3. Track selection: simpleai

What's next?

Discussion

Project:

- Code: https://gitlab.com/samthursfield/calliope
- Package: pip install calliope-music
- Documentation: https://calliope-music.readthedocs.io

Forums:

- Beets forum: "Calliope antisocial music recommendations"
- Metabrainz forum: "Commandline tool for working with Listenbrainz data"

Keep it simple!



