

INTRODUCTION TO DIGITAL MUSICOLOGY

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Preface

This page contains material for the course **Introduction to Digital Musicology**, held at Julius-Maximilians-Universität, Würzburg (Germany) in Fall 2025.

 Warning

This is work in progress.

Test citation Knuth (1984).

Part I.

INTRODUCTION

1. What is Digital Musicology?

Introduction and terminology

Goal

Understanding what “digital musicology” means.

- Introduction
- Overview of the field
- Terminology
 - e.g. digital vs computational; the latter in 2nd semester
 - digital vs empirical vs quantitative
 - how does DM relate to “traditional” subdivisions of musicology?

Exercise

Read.

2. Digital Musicology today

Goal

Acquiring and overview of current activities in Digital Musicology.

- Current research topics
- Important institutions and people (also, e.g. NFDI4Culture)
- Central journals and conferences

3. The history of Digital Musicology

Goal

Knowing the beginnings and the major stages of DM.

Part II.

DATA ABOUT MUSIC

4. RISM metadata

Goal

Learn what metadata are and how to search for music sources on RISM Online.

- What is RISM?
- What is RISM Online?

Exercise

Understand basic SPARQL and design queries via prompting.

5. Spotify and MusicBrainz metadata

Goal

Understand the kind of metadata provided by Spotify vs MusicBrainz.

6. Music and the streaming industry

Goal

Gain first insights into the music market and its workings.

Exercise

Work with sales data.

7. Analyzing song survival

In this session, we will analyze songs from the Billboard 100 charts and trace their ‘course of life’ in the charts.

The data was obtained from Kaggle, a large community website for data analysis challenges.

As before, we first import the `pandas` library for data analysis and load the data using the `read_csv` function that takes as its main argument the path to the data file, in our case `charts.csv`.

```
import pandas as pd
import matplotlib.pyplot as plt

df = pd.read_csv("data/charts.csv")
```

Inspecting the first 5 lines with the `.head()` method of pandas DataFrames, we obtain an understanding of the structure of the data.

```
df.head()
```

	date	rank	song	artist	last-week	peak-rank	weeks-on-bc
0	2021-11-06	1	Easy On Me	Adele	1.0	1	3
1	2021-11-06	2	Stay	The Kid LAROI & Justin Bieber	2.0	1	16
2	2021-11-06	3	Industry Baby	Lil Nas X & Jack Harlow	3.0	1	14
3	2021-11-06	4	Fancy Like	Walker Hayes	4.0	3	19
4	2021-11-06	5	Bad Habits	Ed Sheeran	5.0	2	18

Think: What do the columns represent? Provide verbal descriptions of their meaning and write it down.

After this general overview, we might want to achieve a slightly deeper understanding. For instance, it is not difficult to interpret the `date` column, but from only the first few entries, we cannot know the temporal extend of our data.

Let’s find out what the earliest and latest dates are using the `.min()` and `.max()` methods, respectively.

7. Analyzing song survival

```
df["date"].min(), df["date"].max()
```

```
('1958-08-04', '2021-11-06')
```

This tells us that the data stored in `charts.csv` runs from August 1958 to November 2021 and thus allows us to trace the movement of songs in the Billboard charts across more than 60 years.

```
# Top artists
df.artist.value_counts()
```

```
artist
Taylor Swift                1023
Elton John                  889
Madonna                     857
Drake                       787
Kenny Chesney               769
...
YoungBoy Never Broke Again Featuring Sherhonda Gaulden    1
Drake Featuring Chris Brown                               1
Kehlani Featuring Jhene Aiko                             1
DaBaby Featuring A Boogie Wit da Hoodie & London On Da Track 1
The Shins                                                  1
Name: count, Length: 10205, dtype: int64
```

```
# Longest in charts
df.sort_values(by="weeks-on-board", ascending=True).iloc[50_000:]
```

	date	rank	song	artist	last-week	peak-rank	we
213768	1980-11-22	69	Turn And Walk Away	The Babys	79.0	69	2
106440	2001-06-16	41	Fill Me In	Craig David	69.0	41	2
106443	2001-06-16	44	Bootylicious	Destiny's Child	66.0	44	2
106448	2001-06-16	49	All Or Nothing	O-Town	60.0	49	2
213674	1980-11-29	75	My Mother's Eyes	Bette Midler	85.0	75	2
...
39148	2014-05-10	49	Radioactive	Imagine Dragons	48.0	3	87
1215	2021-08-14	16	Blinding Lights	The Weeknd	17.0	1	87
1117	2021-08-21	18	Blinding Lights	The Weeknd	16.0	1	88
1020	2021-08-28	21	Blinding Lights	The Weeknd	18.0	1	89
919	2021-09-04	20	Blinding Lights	The Weeknd	21.0	1	90


```
df["date"] = pd.to_datetime(df["date"])
```

```
df[df.artist=="Drake"].song.value_counts()
```

```
song
Hotline Bling      36
God's Plan         36
Controlla          26
Fake Love          25
Nice For What      25
..
Trust Issues       1
Too Much           1
Own It             1
Tuscan Leather     1
Come Thru          1
Name: count, Length: 108, dtype: int64
```

```
df[df.artist=="Elton John"].song.value_counts()
```

```
song
Candle In The Wind 1997/Something About The Way You Look Tonight      42
Can You Feel The Love Tonight (From "The Lion King")                  26
I Guess That's Why They Call It The Blues                             23
The One                                                         22
Candle In The Wind                                                         21
Little Jeannie                                                         21
The Last Song                                                         20
Recover Your Soul                                                         20
Believe                                                         20
Circle Of Life (From "The Lion King")                                  20
Blessed                                                         20
Sad Songs (say So Much)                                                19
I Don't Wanna Go On With You Like That                               18
Nikita                                                         18
Bennie And The Jets                                                  18
Mama Can't Buy You Love                                              18
Blue Eyes                                                         18
You Can Make History (Young Again)                                    17
Sacrifice                                                         17
Empty Garden (Hey Hey Johnny)                                         17
```

7. Analyzing song survival

Crocodile Rock	17
Goodbye Yellow Brick Road	17
I'm Still Standing	16
Club At The End Of The Street	16
Simple Life	16
Someday Out Of The Blue	15
Don't Let The Sun Go Down On Me	15
Island Girl	15
Daniel	15
Rocket Man	15
Healing Hands	15
The Bitch Is Back	14
Who Wears These Shoes?	14
Wrap Her Up	14
Sorry Seems To Be The Hardest Word	14
Lucy In The Sky With Diamonds	14
Your Song	14
Nobody Wins	13
A Word In Spanish	13
Someone Saved My Life Tonight	13
You Gotta Love Someone	13
In Neon	13
Chloe	13
(Sartorial Eloquence) Don't Ya Wanna Play This Game No More?	12
Kiss The Bride	12
Saturday Night's Alright For Fighting	12
Grow Some Funk Of Your Own/I Feel Like A Bullet (In The Gun Of Robert Ford)	11
Made In England	10
Levon	10
Victim Of Love	10
Part-Time Love	10
Honky Cat	10
Friends	9
Heartache All Over The World	8
Ego	8
Tiny Dancer	7
Bite Your Lip (Get up and dance!)	6
Border Song	5
Name: count, dtype: int64	

```
def chart_performance(artist, song):  
    data = df[(df["artist"] == artist) & (df["song"] == song)]  
    data = data.sort_values(by="date").reset_index(drop=True)
```

```
data["date_rel"] = pd.to_timedelta(data["date"] - data["date"][0]).dt.days
return data
```

```
test_cases = {
    "Taylor Swift": "You Belong With Me",
    "Drake": "God's Plan",
    "Elton John": "Candle In The Wind 1997/Something About The Way You Look Tonight",
    "The Weeknd": "Blinding Lights",
    "Elvis Presley": "Please Don't Stop Loving Me"
}
```

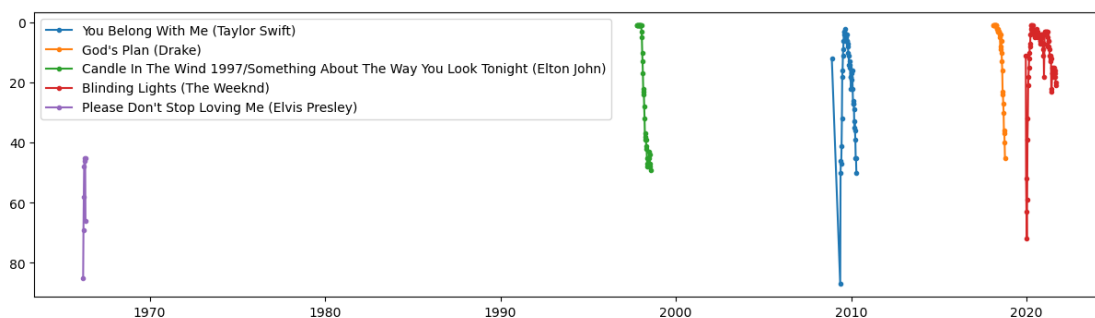
```
taylor = chart_performance("Taylor Swift", "You Belong With Me")
drake = chart_performance("Drake", "God's Plan")
elton = chart_performance("Elton John", "Candle In The Wind 1997/Something About The Way You Look Tonight")
weeknd = chart_performance("The Weeknd", "Blinding Lights")
elvis = chart_performance("Elvis Presley", "Please Don't Stop Loving Me")
```

```
_, ax = plt.subplots(figsize=(15,4))

for artist, song in test_cases.items():
    data = chart_performance(artist, song)
    x = data["date"].values
    y = data["rank"].values

    ax.plot(x, y, marker=".", label=f"{song} ({artist})")

plt.gca().invert_yaxis()
plt.legend()
plt.show()
```



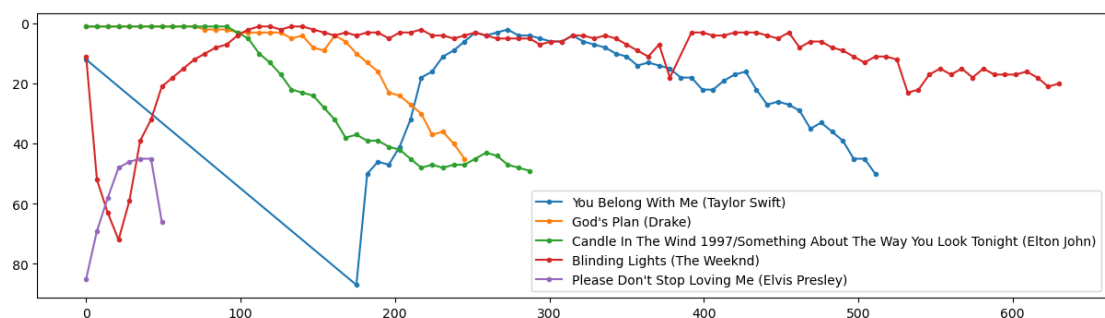
7. Analyzing song survival

```
_, ax = plt.subplots(figsize=(15,4))

for artist, song in test_cases.items():
    data = chart_performance(artist, song)
    x = data["date_rel"].values
    y = data["rank"].values

    ax.plot(x, y, marker=".", label=f"{song} ({artist})")

plt.gca().invert_yaxis()
plt.legend()
plt.show()
```



```
# TODO: remove lines for missing weeks (gaps in curves)
# add two cases:
# - short duration but high peak
# - long duration but low peak
```

```
# Q: can we predict a song's survival using the features given in the data?
# --> at least introduce notion of training/test data and discuss the epistemologic
# sources for explanation
```

```
# Try other data: https://www.kaggle.com/datasets/thedevastator/billboard-hot-100-a
```

```
df_charts = pd.read_csv("Hot Stuff.csv", index_col=0)
df_charts["WeekID"] = pd.to_datetime(df_charts["WeekID"])
```

```
df_charts.head()
```

	url	WeekID	Week Position	Song
index				
0	http://www.billboard.com/charts/hot-100/1965-0...	1965-07-17	34	Don't Just Stand TI
1	http://www.billboard.com/charts/hot-100/1965-0...	1965-07-24	22	Don't Just Stand TI
2	http://www.billboard.com/charts/hot-100/1965-0...	1965-07-31	14	Don't Just Stand TI
3	http://www.billboard.com/charts/hot-100/1965-0...	1965-08-07	10	Don't Just Stand TI
4	http://www.billboard.com/charts/hot-100/1965-0...	1965-08-14	8	Don't Just Stand TI

```
df_audio = pd.read_csv("Hot 100 Audio Features.csv", index_col=0)
```

```
df_audio.head()
```

	SongID	Performer	Song
index			
0	-twistin'-White Silver SandsBill Black's Combo	Bill Black's Combo	-twistin'-White Silver Sands
1	¿Dónde Está Santa Claus? (Where Is Santa Claus...	Augie Rios	¿Dónde Está Santa Claus?
2And Roses And RosesAndy Williams	Andy WilliamsAnd Roses And Roses
3	...And Then There Were DrumsSandy Nelson	Sandy Nelson	...And Then There Were Dr
4	...Baby One More TimeBritney Spears	Britney Spears	...Baby One More Time

```
d = df_charts.merge(df_audio)
```

```
d.shape
```

```
(330208, 29)
```

```
d["WeekID"] = pd.to_datetime(d["WeekID"])
```

```
d.sample(10)
```

	url	WeekID	Week Position	Song
275610	http://www.billboard.com/charts/hot-100/1964-0...	1964-06-20	85	My Dreams
189145	http://www.billboard.com/charts/hot-100/2014-0...	2014-02-08	97	Radio
245340	http://www.billboard.com/charts/hot-100/1969-0...	1969-08-02	92	Let's Call It A Day
141579	http://www.billboard.com/charts/hot-100/2009-0...	2009-08-08	3	Knock You Down
117570	http://www.billboard.com/charts/hot-100/1960-0...	1960-02-20	93	Sleepy Lagoon
150750	http://www.billboard.com/charts/hot-100/1966-0...	1966-09-17	36	Flamingo

7. Analyzing song survival

	url	WeekID	Week Position	Song
23582	http://www.billboard.com/charts/hot-100/1985-1...	1985-10-05	63	Soul K
21493	http://www.billboard.com/charts/hot-100/2003-0...	2003-04-12	21	Rock '
71236	http://www.billboard.com/charts/hot-100/2008-1...	2008-12-13	52	My Li
208184	http://www.billboard.com/charts/hot-100/1984-1...	1984-12-15	60	Missin

```
## BOOTSTRAP!
```

```
# d = d.sample(500_000, replace=True)
```

```
d.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 330208 entries, 0 to 330207
Data columns (total 29 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   url                                   330208 non-null  object
1   WeekID                               330208 non-null  datetime64[ns]
2   Week Position                         330208 non-null  int64
3   Song                                 330208 non-null  object
4   Performer                           330208 non-null  object
5   SongID                               330208 non-null  object
6   Instance                             330208 non-null  int64
7   Previous Week Position                298048 non-null  float64
8   Peak Position                         330208 non-null  int64
9   Weeks on Chart                       330208 non-null  int64
10  spotify_genre                         315700 non-null  object
11  spotify_track_id                     287066 non-null  object
12  spotify_track_preview_url            169915 non-null  object
13  spotify_track_duration_ms            287066 non-null  float64
14  spotify_track_explicit                287066 non-null  object
15  spotify_track_album                  287004 non-null  object
16  danceability                         286508 non-null  float64
17  energy                               286508 non-null  float64
18  key                                   286508 non-null  float64
19  loudness                             286508 non-null  float64
20  mode                                 286508 non-null  float64
21  speechiness                          286508 non-null  float64
22  acousticness                         286508 non-null  float64
23  instrumentality                      286508 non-null  float64
24  liveness                             286508 non-null  float64
```

```

25 valence                286508 non-null float64
26 tempo                  286508 non-null float64
27 time_signature         286508 non-null float64
28 spotify_track_popularity 287066 non-null float64
dtypes: datetime64[ns](1), float64(15), int64(4), object(9)
memory usage: 73.1+ MB

```

```
from IPython.display import Audio, HTML
```

```
Audio(url=d.loc[1000,"spotify_track_preview_url"])
```

```
<IPython.lib.display.Audio object>
```

```

def curves(performer, song):
    data = d[(d.Performer == performer) & (d.Song == song)].sort_values(by="WeekID").reset_index()
    data["date_rel"] = pd.to_timedelta(data["WeekID"] - data["WeekID"][0]).dt.days
    x = data["date_rel"].values # or date_rel or WeekID
    y = data["Week Position"].values
    return x,y

```

```

test_cases2 = {
    "Patty Duke": "Don't Just Stand There",
    "Ace Of Base": "Don't Turn Around",
    "Dan + Shay": "Speechless",
    "YoungBloodZ Featuring Lil Jon": "Damn!",
    "K-Ci & JoJo": "All My Life",
    "Trevor Daniel": "Falling"
}

```

```

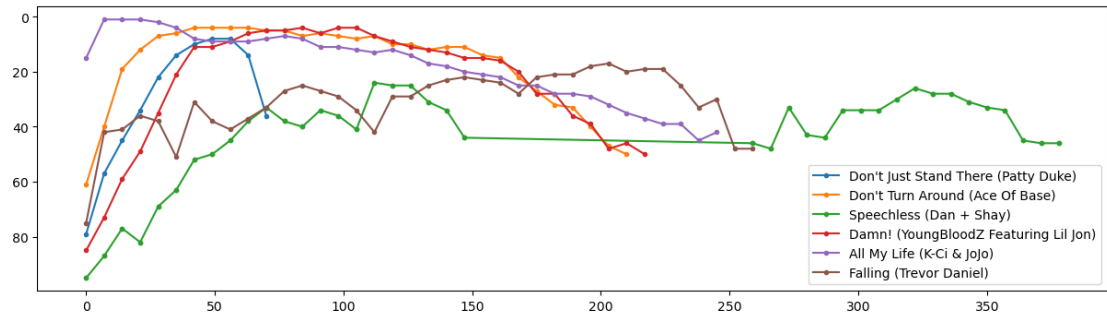
_, ax = plt.subplots(figsize=(15,4))

for performer, song in test_cases2.items():
    x,y = curves(performer, song)
    ax.plot(x, y, marker=".", label=f"{song} ({performer})")

plt.gca().invert_yaxis()
plt.legend()
plt.show()

```

7. Analyzing song survival



Modeling the life of a song in the Top 100:

We assume that once a song has left the Top 100, it is impossible to re-enter (even though that does happen, of course)

1. Each song has a starting rank r_0 .
2. For each following week, there is a bernoulli dropout probability θ that determines whether a song remains in the charts.
- 3.

Observation: Genres tend to leave the Top 100 higher than they entered them

```
entrances = []
peaks = []
exits = []

for _, group in d.groupby("SongID"):
    weeks = group.sort_values(by="WeekID")["Week Position"].values
    entrances.append(weeks[0])
    peaks.append(weeks.min())
    exits.append(weeks[-1])
```

```
import numpy as np
```

```
# from matplotlib.collections import LineCollection
```

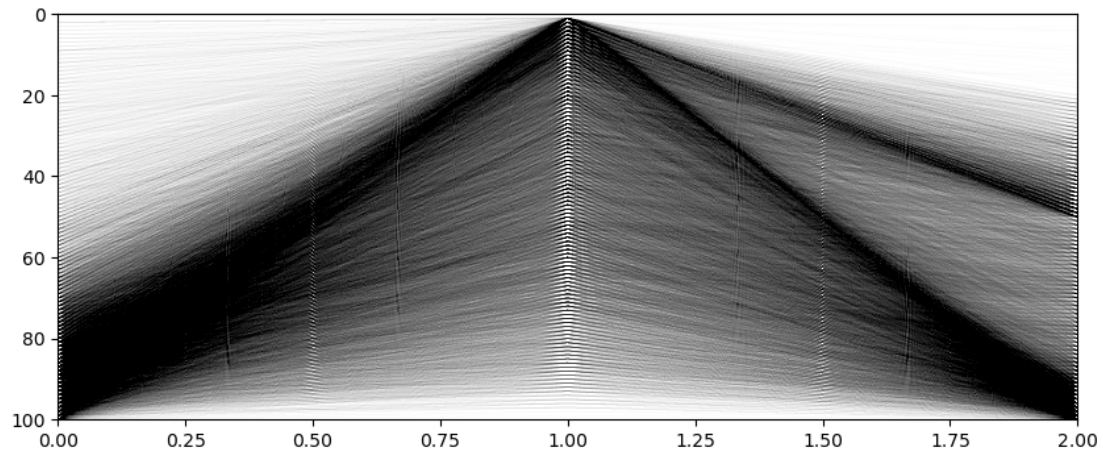
```
_, ax = plt.subplots(figsize=(10,4))

K = len(entrances) + 1

for a, b, c in zip(entrances[:K], peaks[:K], exits[:K]):
    if a != b != c: # remove constants
        ax.plot([0, 1, 2], [a, b, c], c="k", lw=.5, alpha=.01)
```



```
plt.xlim(0,2)
plt.ylim(0,100)
plt.gca().invert_yaxis() # smaller is better
plt.savefig("img/rise-decline.png", dpi=600)
plt.show()
```



OBSERVATION: At least 3 types:

- constants
- low in, peak, low out
- low in, peak, mid out

Try to disentangle what causes the difference

Part III.

MUSIC AS DATA

8. Audio

Goal

Understand what an audio signal is and how it is represented digitally .

- Waveform to spectrogram
- Harmonics
- Timbre
- Audible range and volume
- reading melodies from a spectrogram
- digital audio: sampling

9. MIDI

Goal

Be able to name use cases for MIDI. Translate MIDI numbers to pitches.

10. MEI - header

Goal

Understand basic XML encoding and the skeleton structure of MEI.

- mei friend

11. MEI - the body

Goal

Understand the relation between CWMN and the MEI music element.

- MuseScore export
- mei friend

Part IV.

WORKING WITH MUSIC DATA

12. Digital music analysis: harmony

Goal

Understand what labeling is and why labels can be useful.

- further MuseScore practice
- segmentation and labeling
- Counting chords, finding cadences

13. Digital music analysis: melody

Goal

Understand how melodic pattern matching works in principle.

- Pattern finding in melodies (Non-Western)

Part V.

CRITICAL DIGITAL MUSICOLOGY

14. Copyright

Goal

Know a few famous copyright infringement cases and why data analysis is important here.

- Plagiarism cases and copyright

15. Representation and representativeness

Goal

Understand the difference between representativeness and representation. Obtain a critical understanding of biases relevant for data selection.

- Representation and the canon
- Representing means modeling means abstraction (what is “music” in “music encoding”?)
- biases: how to recognize them, how to deal with them, and when biases are a good thing.
- FAIR and CARE

16. Discussion

 Goal

References

Knuth, Donald E. 1984. “Literate Programming.” *Comput. J.* 27 (2): 97–111. <https://doi.org/10.1093/comjnl/27.2.97>.

