Discogs Dataset trend of vinyl and CD with prediction. Decision Trees.

COL Data Science Bootcamp Course

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1. **Data Set Selection and Understanding**

- The Discogs database (discogs.com) is the biggest and most comprehensive music database online. It hosts a catalogue of over 13 million recordings through all genres and formats, as well as an immense database where everyone can buy and sell music.

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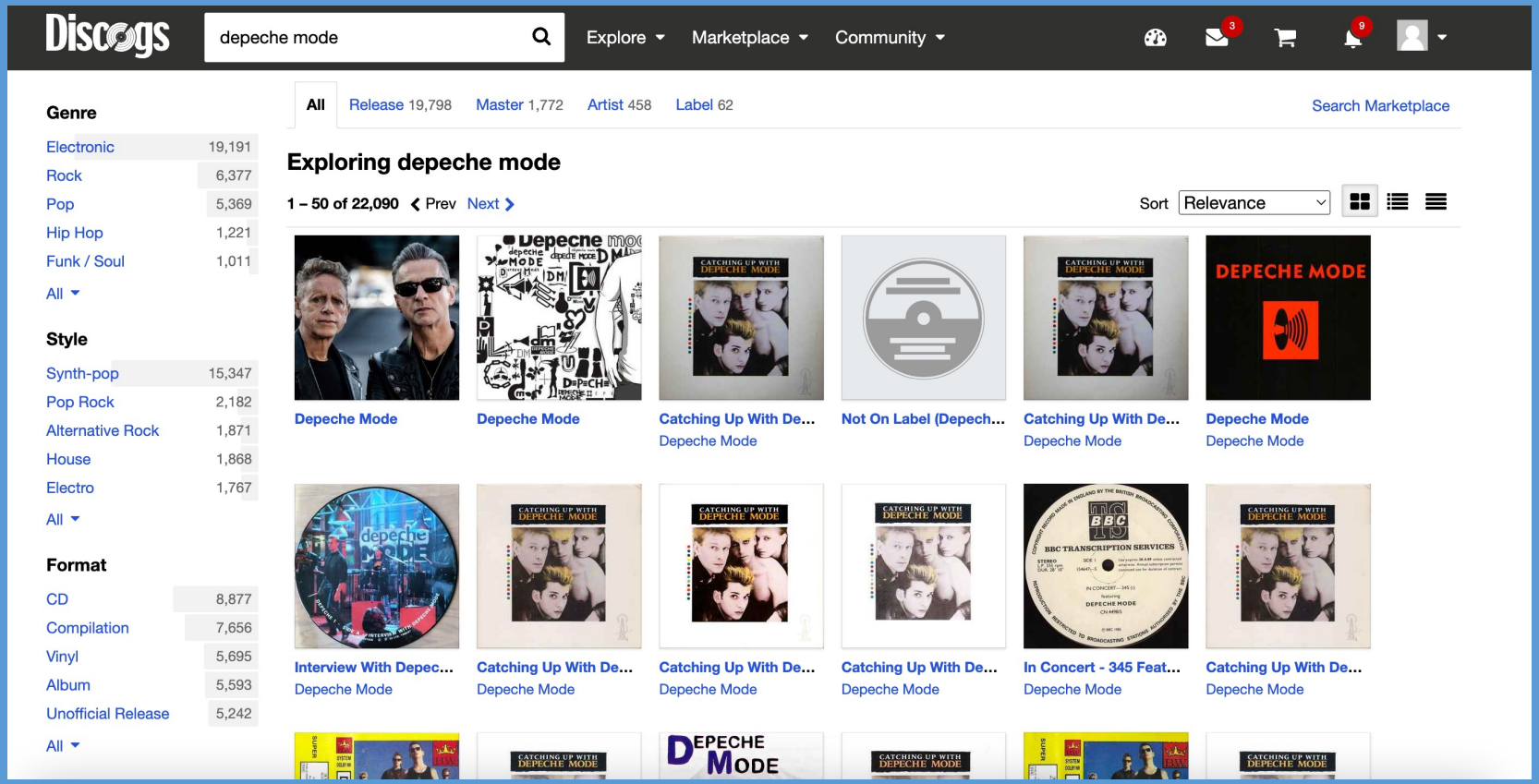
- The dataset has the following attributes: artist, title, label, country, format, release date , genre, styles, have, want, num_ratings, average_rating, lowest_price, median_price, highest_price.

1. Data Set Selection and Understanding



2. Problem Definition

Objectives

In order to figure out the market of turntables and CD, looking into the trend of vinyl and CD sales is an idea.

1. How many vinyl and CD were released per year from 2000 to 2019?

2. Is there a correlation between vinyl records and CD?

3. What is the forecast for the vinyl and CD released through 2025?

4. Decision Tree

3. Exploratory Data Analysis (EDA)

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 315072 entries, 0 to 315071
Data columns (total 6 columns):
 #   Column      Non-Null Count   Dtype
---  ------      --------------   -----
 0   release_id  315072 non-null  int64
 1   country     314426 non-null  object
 2   year        311041 non-null  float64
 3   genre       315072 non-null  object
 4   style       313909 non-null  object
 5   format      315071 non-null  object
dtypes: float64(1), int64(1), object(4)
memory usage: 14.4+ MB
```

df.head()

	release_id	country	year	genre	style	format
0	1	Sweden	1999.0	Electronic	Deep House	Vinyl
1	2	Sweden	1998.0	Electronic	Broken Beat	Vinyl
2	2	Sweden	1998.0	Electronic	Techno	Vinyl
3	2	Sweden	1998.0	Electronic	Tech House	Vinyl
4	3	US	1999.0	Electronic	Techno	CD

```
[ ] df["year"].value_counts()
```

year	count
2016.0	827806
2012.0	814333
2017.0	805624
2015.0	805219
2013.0	801459
...	...
2021.0	4
1889.0	2
1860.0	1
1894.0	1

1. Exploratory
2. Clean data
3. Filter CD &Vinyl
4. Combining per year

```
[ ] df["format"].value_counts()
```

format	count
CD	10699846
Vinyl	10651356
File	3171396
Cassette	2110064
DVD	341762
Shellac	234792
Box Set	232879
All Media	138300
VHS	66013

3. Exploratory Data Analysis (EDA)

Remove duplicates values and combining num realese_id / year

	release_id	country	year	genre	style	format
83	56	US	2000	Electronic	House	CD
84	56	US	2000	Electronic	Garage House	CD
452	284	US	2000	Electronic	House	CD
453	284	US	2000	Electronic	Techno	CD
454	284	US	2000	Electronic	Downtempo	CD
492	297	UK	2000	Electronic	Abstract	CD
493	297	UK	2000	Electronic	IDM	CD
494	297	UK	2000	Electronic	Ambient	CD
519	313	US	2000	Electronic	Techno	CD
520	313	US	2000	Electronic	Electro	CD

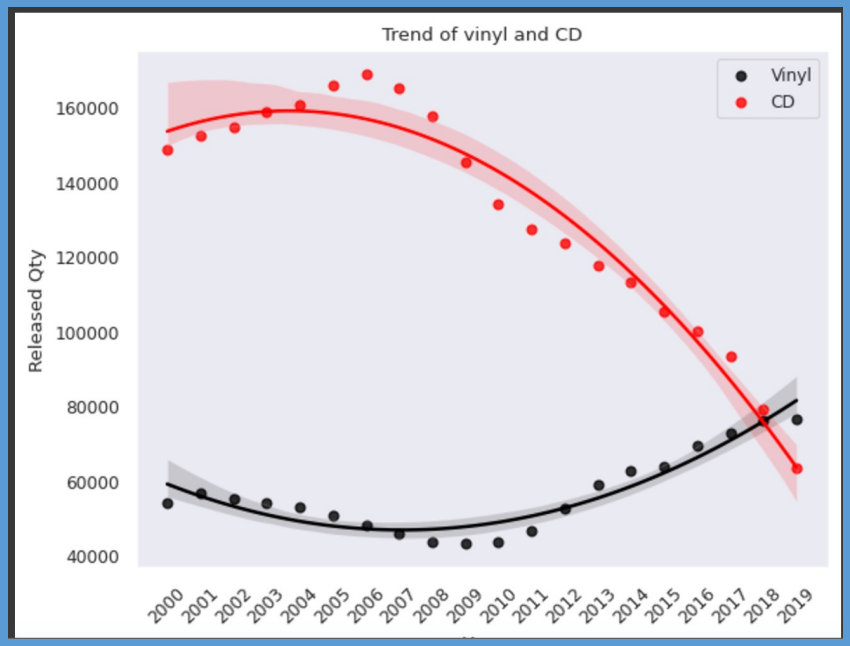
	release_id	country	year	genre	style	format
15	9	US	2000	Electronic	House	Vinyl
16	9	US	2000	Electronic	Deep House	Vinyl
23	15	US	2000	Electronic	House	Vinyl
24	15	US	2000	Electronic	Deep House	Vinyl
25	16	US	2000	Electronic	Techno	Vinyl
26	16	US	2000	Electronic	Tech House	Vinyl
30	18	US	2000	Electronic	Techno	Vinyl
31	18	US	2000	Electronic	Tech House	Vinyl
32	19	Canada	2000	Electronic	Progressive House	Vinyl
33	19	Canada	2000	Electronic	House	Vinyl

	release_id	country	year	genre	style	format
57	36	US	2000	Electronic	Trance	CD
83	56	US	2000	Electronic	House	CD
452	284	US	2000	Electronic	House	CD
474	293	UK	2000	Electronic	Leftfield	CD
492	297	UK	2000	Electronic	Abstract	CD
519	313	US	2000	Electronic	Techno	CD
521	314	Germany	2000	Electronic	Abstract	CD
524	315	Belgium	2000	Electronic	Abstract	CD
551	324	UK	2000	Electronic	Abstract	CD
554	325	UK	2000	Electronic	Leftfield	CD

	release_id	country	year	genre	style	format
13	7	US	2000	Electronic	Deep House	Vinyl
14	8	US	2000	Electronic	Deep House	Vinyl
15	9	US	2000	Electronic	House	Vinyl
20	13	US	2000	Electronic	Deep House	Vinyl
23	15	US	2000	Electronic	House	Vinyl
25	16	US	2000	Electronic	Techno	Vinyl
30	18	US	2000	Electronic	Techno	Vinyl
32	19	Canada	2000	Electronic	Progressive House	Vinyl
35	20	US	2000	Electronic	Tech House	Vinyl
36	21	UK	2000	Electronic	Deep Techno	Vinyl

4. Model Selection and Evaluation

Plot the data to see the released vinyl and CD were changing

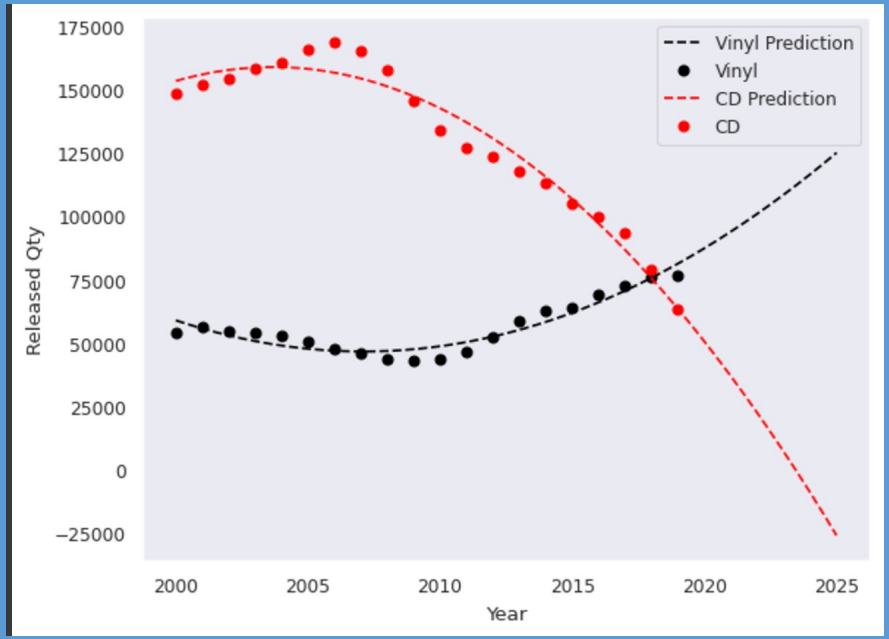



```
[ ] data_v_c=pd.merge(data_v_sorted, data_c_sorted, how='inner')
    data_v_c['year']=data_v_c['year'].astype('int')
    print(data_v_c)
```

```
    year  vinyl      CD
0   2000  53970  148730
1   2001  56417  152220
2   2002  54920  154569
3   2003  53890  158510
4   2004  52868  160395
5   2005  50617  165640
6   2006  47845  168719
7   2007  45816  165118
8   2008  43600  157706
9   2009  43002  145387
10  2010  43360  133979
11  2011  46286  127099
12  2012  52312  123323
13  2013  58630  117669
14  2014  62614  112915
15  2015  63657  105032
16  2016  69196  100017
17  2017  72601   93325
18  2018  76070   78914
19  2019  76406   63316
```

4. Model Selection and Evaluation

Regression model polynomial In order to forecast how many vinyl and CD may be released in the next 5 years.




```
[28] corr_v_c = data_v_c[['vinyl', 'CD']].corr()
     print(corr_v_c)

            vinyl        CD
vinyl  1.000000 -0.839542
CD    -0.839542  1.000000
```

The number of records for vinyls are increasing, although the number of CDs are decreasing.

There's an inverse relationship between the variables vinyls and CDs.

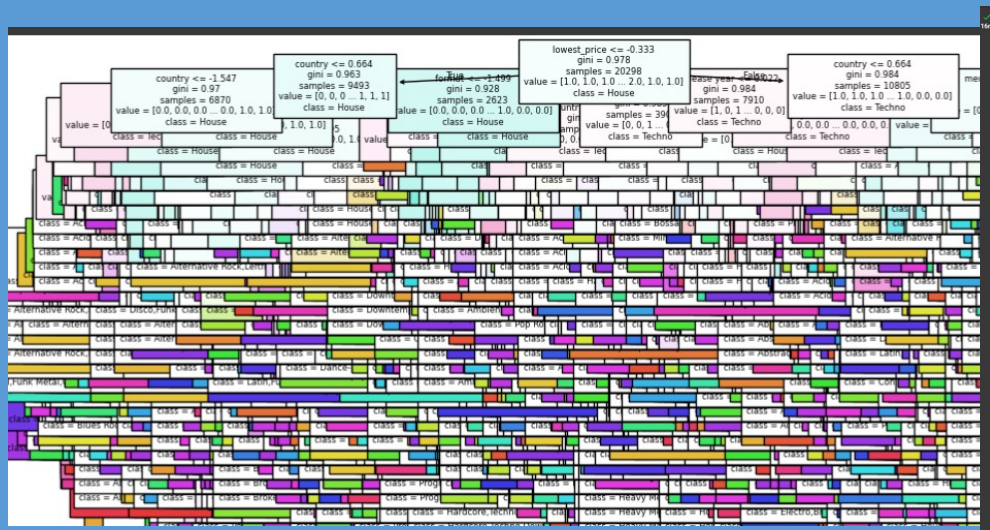
4. Model Selection and Evaluation

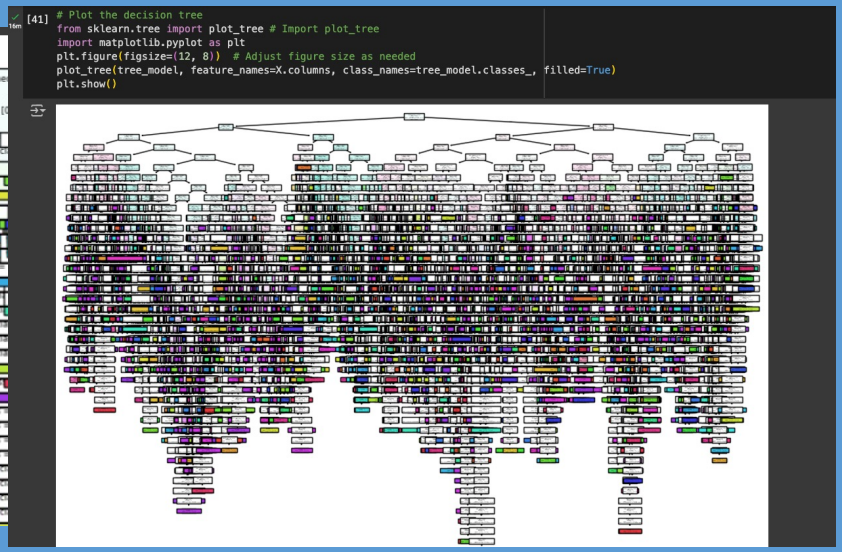
Decision Tree

The 'styles' feature in the dataset likely represents a list of music styles associated with each record. This is a categorical feature that could be valuable for predicting other aspects of the record, such as genre, popularity, or price. Possible decision tree tasks with 'styles'.

Conditions: we only take top 5 countries.

Due to the quantities of matrix the decision tree is not a model selection to be considered.





5. Feedback and Summarizing

The number of records for vinyls are increasing, although the number of CDs are decreasing.

There's an inverse relationship between the variables vinyls and CDs.

Vinyls is increasing the quantities released exponential.

CDs are not expecting to be selling anymore.

People are preferring vinyls collectors or media than CDs.