

4_pandas

October 17, 2023

1 Data Wrangling with Pandas

- A major part of computational social science is the storing, manipulation and reporting of data.
- Pandas is a powerful data management library specifically built for these kinds of tasks.
- It can handle very large amounts of data whilst remaining quick and responsive.
- We will be using Pandas throughout our practical sessions as a general purpose data management tool but this week we will focus on learning its features.

Pandas Documentation

The first thing we need to do is `import` the pandas library. This ensures it is available for us to use in this environment.

Here we import the `pandas` module. We could simply use `import pandas` however `as` allows us to use a shorter name.

As social convention many modules are referred to with these short names.

```
[ ]: import pandas as pd
```

1.1 Loading the Data

Today we will be using data gathered from Spotify, the popular music streaming service. Spotify provides access to some of its data through their public API. This data has been collected and pre-prepared by your instructors.

```
[ ]: filename = 'test_spotify_top_songs.csv'
songs_df = pd.read_csv(filename)
type(songs_df)
```

```
[ ]: pandas.core.frame.DataFrame
```

```
[ ]: songs_df
```

```
[ ]:
```

| | track_id | track_name | artists \ |
|---|------------------------|--------------------|-------------|
| 0 | 5mjYQaktjmjcMKcUIcqz4s | Strangers | Kenya Grace |
| 1 | 56y1j0TK0XSvJzVv9vHQBK | Paint The Town Red | Doja Cat |
| 2 | 1reEeZH9wNt4z1ePYLyC7p | greedy | Tate McRae |

| | | | |
|------|------------------------|------------------------|----------------|
| 3 | 59NraMJSLaMCVtwXTSia8i | Prada | cassö |
| 4 | 5aIVCx5tnkOntmdiinnYvw | Water | Tyla |
| ... | ... | ... | ... |
| 1275 | 07GtD0Cxmye5KDwsTSACPk | Chantilly Lace | The Big Bopper |
| 1276 | 3SQhmctWreNMOX6Zkm2K5R | Maybellene | Chuck Berry |
| 1277 | 6o5RbQYhmF4yRRw7ZArxGL | Autumn In New York | Billie Holiday |
| 1278 | 1xV0ttVNT27FBTD8iHj0fU | It's Only Make Believe | Conway Twitty |
| 1279 | 3nUrhP3KuK4R1qdxRk2Kgo | Stupid Cupid | Connie Francis |

| | genre | release_year | release_date | explicit | popularity \ |
|------|-----------------------|--------------|--------------|----------|--------------|
| 0 | singer-songwriter pop | 2023 | 2023-09-01 | False | 97 |
| 1 | dance pop | 2023 | 2023-09-20 | True | 87 |
| 2 | alt z | 2023 | 2023-09-13 | True | 31 |
| 3 | ***00PS!*** | 2023 | 2023-08-11 | True | 94 |
| 4 | ***00PS!*** | 2023 | 2023-07-28 | False | 91 |
| ... | ... | ... | ... | ... | ... |
| 1275 | doo-wop | 1958 | 1958-01-01 | False | 57 |
| 1276 | blues | 1959 | 1959-07-01 | False | 57 |
| 1277 | adult standards | 1956 | 1956-01-01 | False | 57 |
| 1278 | arkansas country | 1959 | 1959-01-01 | False | 57 |
| 1279 | adult standards | 2005 | 2005-06-14 | False | 57 |

| | duration_ms | playlist_name | danceability | loudness \ |
|------|-------------|-------------------------|--------------|------------|
| 0 | 172964 | Top 50 - United Kingdom | 0.628 | -8.307 |
| 1 | 230480 | Top 50 - United Kingdom | 0.864 | -7.683 |
| 2 | 131872 | Top 50 - United Kingdom | 0.750 | -3.190 |
| 3 | 132359 | Top 50 - United Kingdom | 0.638 | -5.804 |
| 4 | 200255 | Top 50 - United Kingdom | 0.673 | -3.495 |
| ... | ... | ... | ... | ... |
| 1275 | 145266 | All Out 50s | 0.489 | -6.054 |
| 1276 | 143240 | All Out 50s | 0.756 | -10.701 |
| 1277 | 224333 | All Out 50s | 0.587 | -15.196 |
| 1278 | 132026 | All Out 50s | 0.461 | -9.627 |
| 1279 | 133746 | All Out 50s | 0.609 | -4.739 |

| | speechiness | playlist_type |
|------|-------------|-----------------|
| 0 | NaN | mixed_pop |
| 1 | 0.1940 | mixed_pop |
| 2 | 0.0322 | mixed_pop |
| 3 | 0.0375 | mixed_pop |
| 4 | 0.0755 | mixed_pop |
| ... | ... | ... |
| 1275 | 0.0858 | all_out_decades |
| 1276 | 0.1120 | all_out_decades |
| 1277 | 0.0607 | all_out_decades |
| 1278 | 0.0598 | all_out_decades |
| 1279 | 0.0389 | all_out_decades |

[1280 rows x 14 columns]

```
[ ]: # .head() shows us the top 5 rows
songs_df.head()
```

```
[ ]:
      track_id      track_name      artists \
0  5mjYQaktjmcMKcUIcqz4s      Strangers  Kenya Grace
1  56y1j0TKOXsvJzVv9vHQBK  Paint The Town Red      Doja Cat
2  1reEeZH9wNt4z1ePYLyC7p      greedy      Tate McRae
3  59NraMJsLaMCVtwXTSia8i      Prada      cassö
4  5aIVCx5tnk0ntmdiinnYvw      Water      Tyla

      genre  release_year  release_date  explicit  popularity \
0  singer-songwriter pop      2023    2023-09-01      False      97
1      dance pop      2023    2023-09-20      True      87
2      alt z      2023    2023-09-13      True      31
3  ***OOPS!***      2023    2023-08-11      True      94
4  ***OOPS!***      2023    2023-07-28      False      91

      duration_ms      playlist_name  danceability  loudness  speechiness \
0      172964  Top 50 - United Kingdom      0.628    -8.307      NaN
1      230480  Top 50 - United Kingdom      0.864    -7.683      0.1940
2      131872  Top 50 - United Kingdom      0.750    -3.190      0.0322
3      132359  Top 50 - United Kingdom      0.638    -5.804      0.0375
4      200255  Top 50 - United Kingdom      0.673    -3.495      0.0755

      playlist_type
0      mixed_pop
1      mixed_pop
2      mixed_pop
3      mixed_pop
4      mixed_pop
```

```
[ ]: # .tail() shows us the last 5 rows
songs_df.tail()
```

```
[ ]:
      track_id      track_name      artists \
1275  07GtD0Cxmye5KDWsTSACPk  Chantilly Lace  The Big Bopper
1276  3SqhmctWreNMOX6Zkm2K5R      Maybellene      Chuck Berry
1277  6o5RbQYhmF4yRRw7ZArxGL  Autumn In New York  Billie Holiday
1278  1xV0ttVNT27FBTD8iHj0fU  It's Only Make Believe  Conway Twitty
1279  3nUrhP3KuK4R1qdxRk2Kgo  Stupid Cupid  Connie Francis

      genre  release_year  release_date  explicit  popularity \
1275      doo-wop      1958    1958-01-01      False      57
```

| | | | | | |
|------|------------------|------|------------|-------|----|
| 1276 | blues | 1959 | 1959-07-01 | False | 57 |
| 1277 | adult standards | 1956 | 1956-01-01 | False | 57 |
| 1278 | arkansas country | 1959 | 1959-01-01 | False | 57 |
| 1279 | adult standards | 2005 | 2005-06-14 | False | 57 |

| | duration_ms | playlist_name | danceability | loudness | speechiness | \ |
|------|-------------|---------------|--------------|----------|-------------|---|
| 1275 | 145266 | All Out 50s | 0.489 | -6.054 | 0.0858 | |
| 1276 | 143240 | All Out 50s | 0.756 | -10.701 | 0.1120 | |
| 1277 | 224333 | All Out 50s | 0.587 | -15.196 | 0.0607 | |
| 1278 | 132026 | All Out 50s | 0.461 | -9.627 | 0.0598 | |
| 1279 | 133746 | All Out 50s | 0.609 | -4.739 | 0.0389 | |

| | playlist_type |
|------|-----------------|
| 1275 | all_out_decades |
| 1276 | all_out_decades |
| 1277 | all_out_decades |
| 1278 | all_out_decades |
| 1279 | all_out_decades |

```
[ ]: # You can specify the number of rows to return
```

```
songs_df.head(10)
```

```
[ ]:
      track_id      track_name \
0  5mjYQaktjmcMKcUIcqz4s      Strangers
1  56y1jOTK0XSvJzVv9vHQBK  Paint The Town Red
2  1reEeZH9wNt4z1ePYLyC7p      greedy
3  59NraMJsLaMCVtwXTSia8i      Prada
4  5aIVCx5tnk0ntmdiinnYvw      Water
5  2FDTHlrBguDzQkp7PVj16Q      Sprinter
6  1BxfuPKGuaTgP7aMOBbdwr      Cruel Summer
7  3vkCue0mm7xQDoJ17W1Pm3  My Love Mine All Mine
8  1kuGVB7EU95pJ0bxwvfwKS      vampire
9  2ZWmmrWUgDBcPSLihBMvhg  Baddadan (feat. IRAH, Flowdan, Trigga & Takura)
```

| | artists | genre | release_year | release_date | explicit | \ |
|---|----------------|-----------------------|--------------|--------------|----------|---|
| 0 | Kenya Grace | singer-songwriter pop | 2023 | 2023-09-01 | False | |
| 1 | Doja Cat | dance pop | 2023 | 2023-09-20 | True | |
| 2 | Tate McRae | alt z | 2023 | 2023-09-13 | True | |
| 3 | cassö | ***OOPS!*** | 2023 | 2023-08-11 | True | |
| 4 | Tyla | ***OOPS!*** | 2023 | 2023-07-28 | False | |
| 5 | Dave | uk hip hop | 2023 | 2023-06-01 | True | |
| 6 | Taylor Swift | pop | 2019 | 2019-08-23 | False | |
| 7 | Mitski | brooklyn indie | 2023 | 2023-09-15 | False | |
| 8 | Olivia Rodrigo | pop | 2023 | 2023-09-08 | True | |
| 9 | Chase & Status | dancefloor dnb | 2023 | 2023-07-27 | False | |

| | popularity | duration_ms | playlist_name | danceability | loudness \ |
|---|------------|-------------|-------------------------|--------------|------------|
| 0 | 97 | 172964 | Top 50 - United Kingdom | 0.628 | -8.307 |
| 1 | 87 | 230480 | Top 50 - United Kingdom | 0.864 | -7.683 |
| 2 | 31 | 131872 | Top 50 - United Kingdom | 0.750 | -3.190 |
| 3 | 94 | 132359 | Top 50 - United Kingdom | 0.638 | -5.804 |
| 4 | 91 | 200255 | Top 50 - United Kingdom | 0.673 | -3.495 |
| 5 | 94 | 229133 | Top 50 - United Kingdom | 0.916 | -8.067 |
| 6 | 99 | 178426 | Top 50 - United Kingdom | 0.552 | -5.707 |
| 7 | 93 | 137773 | Top 50 - United Kingdom | 0.504 | -14.958 |
| 8 | 95 | 219724 | Top 50 - United Kingdom | 0.511 | -5.745 |
| 9 | 81 | 177291 | Top 50 - United Kingdom | 0.620 | -0.504 |

| | speechiness | playlist_type |
|---|-------------|---------------|
| 0 | NaN | mixed_pop |
| 1 | 0.1940 | mixed_pop |
| 2 | 0.0322 | mixed_pop |
| 3 | 0.0375 | mixed_pop |
| 4 | 0.0755 | mixed_pop |
| 5 | 0.2410 | mixed_pop |
| 6 | 0.1570 | mixed_pop |
| 7 | 0.0321 | mixed_pop |
| 8 | 0.0578 | mixed_pop |
| 9 | 0.2520 | mixed_pop |

The `.info()` method gives us an overview of our DataFrame, including... - A summary of the index labels - Information the columns - a 'Non-Null' Count. i.e. how many 'cells' in the column have a value in them. - The type (Dtype) of values that column holds. - Integer (int64) - e.g. 5 - Float (float64)- e.g. 5.3 - Boolean (bool) - e.g. True / False - Other (object) - Usually a string, but can also be any python object e.g. lists, dictionaries, classes. - A summary of how much computer memory the data needs.

```
[ ]: songs_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1280 entries, 0 to 1279
Data columns (total 14 columns):
#   Column          Non-Null Count  Dtype
---  -
0   track_id        1280 non-null   object
1   track_name      1280 non-null   object
2   artists         1280 non-null   object
3   genre           1280 non-null   object
4   release_year    1280 non-null   int64
5   release_date    1280 non-null   object
6   explicit        1280 non-null   bool
7   popularity      1280 non-null   int64
8   duration_ms     1280 non-null   int64
9   playlist_name   1280 non-null   object
```

```

10  danceability    1280 non-null    float64
11  loudness        1280 non-null    float64
12  speechiness    1279 non-null    float64
13  playlist_type   1280 non-null    object
dtypes: bool(1), float64(3), int64(3), object(7)
memory usage: 131.4+ KB

```

We can also get some of this information separately. For example the row and column names. Sometimes this is necessary if there are more columns than `.info()` wants to show.

```
[ ]: songs_df.index
```

```
[ ]: RangeIndex(start=0, stop=1280, step=1)
```

```
[ ]: songs_df.columns
```

```
[ ]: Index(['track_id', 'track_name', 'artists', 'genre', 'release_year',
           'release_date', 'explicit', 'popularity', 'duration_ms',
           'playlist_name', 'danceability', 'loudness', 'speechiness',
           'playlist_type'],
          dtype='object')
```

1.2 Accessing columns, rows and cells

Being able to tell Pandas to provide us specific columns, specific rows, or even specific cells can be important when exploring data.

```
[ ]: # We can access a single column by
     # providing the name in square brackets.
     songs_df['track_name']
```

```
[ ]: 0          Strangers
     1    Paint The Town Red
     2          greedy
     3          Prada
     4          Water
     ...
    1275    Chantilly Lace
    1276    Maybellene
    1277    Autumn In New York
    1278    It's Only Make Believe
    1279    Stupid Cupid
     Name: track_name, Length: 1280, dtype: object
```

```
[ ]: # or multiple names in a list if we want a few columns

     songs_df[['track_name', 'artists']]
```

```
[ ]:      track_name      artists
0      Strangers      Kenya Grace
1  Paint The Town Red      Doja Cat
2      greedy      Tate McRae
3      Prada      cassö
4      Water      Tyla
...      ...      ...
1275    Chantilly Lace  The Big Bopper
1276      Maybellene      Chuck Berry
1277    Autumn In New York  Billie Holiday
1278  It's Only Make Believe  Conway Twitty
1279      Stupid Cupid  Connie Francis
```

[1280 rows x 2 columns]

```
[ ]: # We can also assign this to a new variable if we want quick access.
# Note: When you set columns to a variable like this, it is referencing
# the original DataFrame, not copying it.

songs_track_artist = songs_df[['track_name','artists']]
songs_track_artist.head()
```

```
[ ]:      track_name      artists
0      Strangers  Kenya Grace
1  Paint The Town Red      Doja Cat
2      greedy      Tate McRae
3      Prada      cassö
4      Water      Tyla
```

We can access specific rows... - By referring to their row label - .loc - By referring to their row index - .iloc

```
[ ]: songs_df.loc[1]
```

```
[ ]: track_id      56y1j0TK0XSvJzVv9vHQBK
track_name      Paint The Town Red
artists      Doja Cat
genre      dance pop
release_year      2023
release_date      2023-09-20
explicit      True
popularity      87
duration_ms      230480
playlist_name  Top 50 - United Kingdom
danceability      0.864
loudness      -7.683
speechiness      0.194
playlist_type      mixed_pop
```

Name: 1, dtype: object

At the moment row labels and row indexes (the row's position in the data) are the same by default. However, say we messed up the order by sorting the data by year.

```
[ ]: by_year = songs_df.sort_values('release_year')
      by_year
```

```
[ ]:
      track_id      track_name \
1256  7GnMzVWOHLBPcfco4L1GtE      Earth Angel
1269  0x0ffSAP6PkdoDgH0froof  My Funny Valentine - Remastered
1211  4I4aQGNJ2HuflonTb65nxR      That's Amore
1182  648TTtYB0bHOP8HfyOFmkL      Unforgettable
1204  1uRKT2LRANv4baowBWHfDS  (We're Gonna) Rock Around The Clock
...
251   5S9opvHp0ztqiShDAAusum      Terrified
252   3gA9Y5RsXBUKkr1IP6v4g3      Paint My Bedroom Black
253   2gyxAWHebV7xPYVxqoi86f      get him back!
255   0LzidBf7cUsnZnG34OUPSF      Mosquito
0     5mjYQaktjmjcMKcUIcqz4s      Strangers
```

```
      artists      genre  release_year \
1256      The Penguins      doo-wop      1954
1269      Frank Sinatra  adult standards      1954
1211      Dean Martin    adult standards      1954
1182      Nat King Cole  adult standards      1954
1204  Bill Haley & His Comets  rock-and-roll      1955
...
251      Jazmin Bean      pov: indie      2023
252      Holly Humberstone      alt z      2023
253      Olivia Rodrigo      pop      2023
255      PinkPantheress  bedroom pop      2023
0      Kenya Grace  singer-songwriter pop      2023
```

```
      release_date  explicit  popularity  duration_ms      playlist_name \
1256  1954-05-26      False      59      171066      All Out 50s
1269  1954-01-01      False      58      150666      All Out 50s
1211      1954      False      64      190400      All Out 50s
1182  1954-01-01      False      71      191973      All Out 50s
1204  1955-12-19      False      66      129893      All Out 50s
...
251   2023-10-13      True      49      186071      alt/pop
252   2023-10-11      False     59      229713      alt/pop
253   2023-09-08      True     92      211141      alt/pop
255   2023-09-29      False     76      146240      alt/pop
0     2023-09-01      False     97      172964  Top 50 - United Kingdom
```

```
      danceability  loudness  speechiness  playlist_type
```


| | | | | |
|------|-------|---------|--------|-----------------|
| 1256 | 0.487 | -11.121 | 0.0271 | all_out_decades |
| 1269 | 0.257 | -14.267 | 0.0332 | all_out_decades |
| 1211 | 0.471 | -13.600 | 0.0309 | all_out_decades |
| 1182 | 0.349 | -13.507 | 0.0310 | all_out_decades |
| 1204 | 0.811 | -6.317 | 0.1680 | all_out_decades |
| ... | ... | ... | ... | ... |
| 251 | 0.415 | -4.550 | 0.0346 | mixed_pop |
| 252 | 0.717 | -9.560 | 0.0413 | mixed_pop |
| 253 | 0.546 | -5.719 | 0.1810 | mixed_pop |
| 255 | 0.710 | -4.032 | 0.1180 | mixed_pop |
| 0 | 0.628 | -8.307 | NaN | mixed_pop |

[1280 rows x 14 columns]

To access the very first row of this dataset we either need to know the row label...

```
[ ]: by_year.loc[1211]
```

```
[ ]: track_id      4I4aQGNJ2HufloNtB65nxR
track_name      That's Amore
artists        Dean Martin
genre          adult standards
release_year      1954
release_date      1954
explicit         False
popularity        64
duration_ms      190400
playlist_name     All Out 50s
danceability      0.471
loudness         -13.6
speechiness       0.0309
playlist_type     all_out_decades
Name: 1211, dtype: object
```

Or simply ask for the first row by index location like we would a list.

```
[ ]: by_year.iloc[0]
```

```
[ ]: track_id      7GnMzVWOHLBPcfco4L1GtE
track_name      Earth Angel
artists        The Penguins
genre          doo-wop
release_year      1954
release_date      1954-05-26
explicit         False
popularity        59
duration_ms      171066
playlist_name     All Out 50s
```

```
danceability      0.487
loudness          -11.121
speechiness       0.0271
playlist_type     all_out_decades
Name: 1256, dtype: object
```

Or the very last row, again like a list.

```
[ ]: by_year.iloc[-1]
```

```
[ ]: track_id      5mjYQaktjmcMKcUIcqz4s
     track_name    Strangers
     artists       Kenya Grace
     genre          singer-songwriter pop
     release_year   2023
     release_date   2023-09-01
     explicit       False
     popularity     97
     duration_ms    172964
     playlist_name  Top 50 - United Kingdom
     danceability   0.628
     loudness       -8.307
     speechiness    NaN
     playlist_type  mixed_pop
     Name: 0, dtype: object
```

The index labels given to rows stick to them like an ID, however an index position can change.

| index | label | name |
|-------|-------|--------|
| 0 | 0 | Arthur |
| 1 | 1 | Betty |
| 2 | 2 | Carole |

After reversing the order of **name**

| index | label | name |
|-------|-------|--------|
| 0 | 2 | Carole |
| 1 | 1 | Betty |
| 2 | 0 | Arthur |

Finally we can access specific cells, or collections of cells using `.loc` and `.iloc` as well.

```
[ ]: # Row named 4, column artists
     songs_df.loc[2, 'artists']
```

```
[ ]: 'Tate McRae'
```

```
[ ]: # Rows named 2 to 4, column artists and track_name
songs_df.loc[2:4,['artists','track_name']]
```

```
[ ]:      artists track_name
2  Tate McRae    greedy
3      cassö    Prada
4      Tyla    Water
```

```
[ ]: # Rows at position -5 to -1, second column
songs_df.iloc[-5:-1, 1 ]
```

```
[ ]: 1275      Chantilly Lace
1276      Maybellene
1277      Autumn In New York
1278      It's Only Make Believe
Name: track_name, dtype: object
```

Often if you are accessing specific rows, you'll use `.loc` but it is helpful to know the difference from `.iloc` as sometimes you won't know the index names, but you'll know the position.

1.3 Exercises 1

Take a look at section 1 of the exercises sheet. Complete the tasks before moving on.

1.4 Filtering Data

Filtering allows you to select multiple rows based on particular criteria.

A filter is essentially a list of `True` or `False` values that is the same length as our dataset.

If the first item in our list is `True` then that means we want to **keep** the first row, if the second item is `False` that means we **drop** the second row and so on.

```
[ ]: toy_example = pd.read_csv('toy_example.csv')
toy_example
```

```
[ ]:      name  age
0   Arya   20
1    Bob   25
2  Chloe   30
```

```
[ ]: toy_filter = [True, False, True]
toy_example[toy_filter]
```

```
[ ]:      name  age
0   Arya   20
2  Chloe   30
```

Rather than manually make our filters, we use Pandas to generate them based on conditional expressions.

```
[ ]: age_filter = toy_example['age'] > 20
age_filter
```

```
[ ]: 0    False
      1     True
      2     True
      Name: age, dtype: bool
```

```
[ ]: toy_example[age_filter]
```

```
[ ]:      name  age
      1   Bob   25
      2  Chloe   30
```

We can also combine conditions for more complex questions using the & operator which means **and**.

We should wrap our different conditions in parentheses () to make sure we're clear where a condition begins and ends.

```
[ ]: (toy_example['age'] > 20) & (toy_example['age'] < 30)
```

```
[ ]: 0    False
      1     True
      2    False
      Name: age, dtype: bool
```

For readability's sake it is worth either making filters as their own variables, or laying them out on separate lines.

```
[ ]: filter_above_20 = toy_example['age'] > 20
      filter_below_30 = toy_example['age'] < 30
      toy_example[filter_above_20 & filter_below_30]
```

```
[ ]:      name  age
      1   Bob   25
```

```
[ ]: toy_example[
      (toy_example['age'] > 20) &
      (toy_example['age'] < 30)
      ]
```

```
[ ]:      name  age
      1   Bob   25
```

Let's move filter our spotify data to answer a specific question.

Q: What are the most popular explicit songs from the UK Top 50?

Break it down, we'll need three filters - Songs from the UK Top 50 playlist - Songs marked as explicit - Songs above a certain popularity score

Spotify scores songs 0-100, so we'll say 90, which is the top 10%.

```
[ ]: # we check our columns to see what we'll need
songs_df.columns
```

```
[ ]: Index(['track_id', 'track_name', 'artists', 'genre', 'release_year',
         'release_date', 'explicit', 'popularity', 'duration_ms',
         'playlist_name', 'danceability', 'loudness', 'speechiness',
         'playlist_type'],
         dtype='object')
```

Looks like we will be using `playlist_name`, `explicit` and `popularity`.

1.4.1 Filter 1: By playlist

We check to see what the values of the `playlist_name` column look like, what type of values they are and then we can decide how to use them.

```
[ ]: songs_df['playlist_name']
```

```
[ ]: 0      Top 50 - United Kingdom
     1      Top 50 - United Kingdom
     2      Top 50 - United Kingdom
     3      Top 50 - United Kingdom
     4      Top 50 - United Kingdom
     ...
    1275      All Out 50s
    1276      All Out 50s
    1277      All Out 50s
    1278      All Out 50s
    1279      All Out 50s
     Name: playlist_name, Length: 1280, dtype: object
```

```
[ ]: songs_df['playlist_name'].unique()
```

```
[ ]: array(['Top 50 - United Kingdom', 'The Pop List', "Today's Top Hits ",
         'Cheesy Hits!', 'alt/pop', 'Hit Rewind',
         'Every Official UK Number 1 Ever', 'Every UK Number One: 2023',
         'All Out 2000s', 'All Out 2010s', 'All Out 90s', 'All Out 80s',
         'All Out 70s', 'All Out 60s', 'All Out 50s'], dtype=object)
```

```
[ ]: playlist_filter = songs_df['playlist_name'] == 'Top 50 - United Kingdom'
```

1.4.2 Filter 2: By Explicit

We don't know exactly how Spotify categorises this so let's check the column values like before, and then create our filter.

```
[ ]: songs_df['explicit']
```

```
[ ]: 0      False
      1      True
      2      True
      3      True
      4      False
      ...
      1275   False
      1276   False
      1277   False
      1278   False
      1279   False
      Name: explicit, Length: 1280, dtype: bool
```

```
[ ]: explicit_filter = songs_df['explicit'] == True
```

1.4.3 Filter 3: By Popularity

If we check the popularity column we can get a sense of how it works.

```
[ ]: songs_df['popularity']
```

```
[ ]: 0      97
      1      87
      2      31
      3      94
      4      91
      ..
      1275   57
      1276   57
      1277   57
      1278   57
      1279   57
      Name: popularity, Length: 1280, dtype: int64
```

We can check the minimum and maximum values to confirm the range they use, though the API documentation would also tell us this.

The `.describe()` method provides a range of summary information about a column including that... - The **count** of records is 1,280 - The **mean** or average value is 75.49 - The **minimum** value is 0 - The **maximum** value is 100

```
[ ]: # t
      songs_df['popularity'].describe()
```

```
[ ]: count    1280.000000
      mean      75.585938
      std       15.567165
      min        0.000000
      25%       72.000000
```

```

50%          79.000000
75%          85.000000
max          100.000000
Name: popularity, dtype: float64

```

We can also get each of these values independently by using the associated method. For example we can get the mean()

```
[ ]: avg_popularity = songs_df['popularity'].mean()
avg_popularity
```

```
[ ]: 75.5859375
```

```
[ ]: popularity_filter = songs_df['popularity'] > avg_popularity
```

1.4.4 Combining them all together

```
[ ]: uk_popular_explicit = songs_df[playlist_filter & explicit_filter &
    popularity_filter]
uk_popular_explicit
```

```
[ ]:
      track_id      track_name \
1  56y1j0TK0XSvJzVv9vHQBK      Paint The Town Red
3  59NraMJsLaMCVtwXTSia8i      Prada
5  2FDTHlrBguDzQkp7PVj16Q      Sprinter
8  1kuGVB7EU95pJ0bxwfwKS      vampire
10 2YSzYUF3jWqb9YP9VXmpjE      IDGAF (feat. Yeat)
12 4rXLjWdF2ZZpXCVTfWcshS      fukumean
15 602d2gJewoiF1FivuOMMwE      Disconnect
16 3IX0yuEVvDbnqUwMBB3ouC      bad idea right?
17 2gyxAWHebV7xPYVxqoi86f      get him back!
18 7aqfrAY2p9BUSiupwk3svU      First Person Shooter (feat. J. Cole)
22 4RoKNqyZ9622tcAeNPNv5k      City Boys
30 1yeB8MUNeLo9Ek1UEpsyz6      Rich Baby Daddy (feat. Sexyy Red & SZA)
31 3JvKfv6T31z00ini8iNItO      Another Love
33 6WzRpISELf3YglGAh7TXcG      Popular (with Playboi Carti & Madonna) - Music...
38 5mHdCZtVyb4DcJw8799hZp      Escapism.
46 68Dni7IE4VyPkTOH9mRWHr      No Role Modelz
47 3eP13S8D5m2cweMEg3ZDed      Virginia Beach
48 0AYt6NMyyLd0rLuvrOUkMH      Slime You Out (feat. SZA)
```

```

      artists      genre  release_year  release_date \
1      Doja Cat      dance pop      2023      2023-09-20
3      cassö      ***00PS!***      2023      2023-08-11
5      Dave      uk hip hop      2023      2023-06-01
8  Olivia Rodrigo      pop      2023      2023-09-08
10     Drake      canadian hip hop      2023      2023-10-06
12     Gunna      atl hip hop      2023      2023-06-16

```

| | | | | |
|----|----------------|---------------------------|------|------------|
| 15 | Becky Hill | pop dance | 2023 | 2023-07-14 |
| 16 | Olivia Rodrigo | pop | 2023 | 2023-09-08 |
| 17 | Olivia Rodrigo | pop | 2023 | 2023-09-08 |
| 18 | Drake | canadian hip hop | 2023 | 2023-10-06 |
| 22 | Burna Boy | afrobeats | 2023 | 2023-08-24 |
| 30 | Drake | canadian hip hop | 2023 | 2023-10-06 |
| 31 | Tom Odell | chill pop | 2013 | 2013-06-24 |
| 33 | The Weeknd | canadian contemporary r&b | 2023 | 2023-06-02 |
| 38 | RAYE | uk contemporary r&b | 2023 | 2023-02-03 |
| 46 | J. Cole | conscious hip hop | 2014 | 2014-12-09 |
| 47 | Drake | canadian hip hop | 2023 | 2023-10-06 |
| 48 | Drake | canadian hip hop | 2023 | 2023-10-06 |

| | explicit | popularity | duration_ms | playlist_name | danceability \ |
|----|----------|------------|-------------|-------------------------|----------------|
| 1 | True | 87 | 230480 | Top 50 - United Kingdom | 0.864 |
| 3 | True | 94 | 132359 | Top 50 - United Kingdom | 0.638 |
| 5 | True | 94 | 229133 | Top 50 - United Kingdom | 0.916 |
| 8 | True | 95 | 219724 | Top 50 - United Kingdom | 0.511 |
| 10 | True | 89 | 260111 | Top 50 - United Kingdom | 0.663 |
| 12 | True | 95 | 125040 | Top 50 - United Kingdom | 0.847 |
| 15 | True | 80 | 164914 | Top 50 - United Kingdom | 0.504 |
| 16 | True | 94 | 184783 | Top 50 - United Kingdom | 0.627 |
| 17 | True | 92 | 211141 | Top 50 - United Kingdom | 0.546 |
| 18 | True | 89 | 247444 | Top 50 - United Kingdom | 0.470 |
| 22 | True | 84 | 153346 | Top 50 - United Kingdom | 0.808 |
| 30 | True | 84 | 319191 | Top 50 - United Kingdom | 0.645 |
| 31 | True | 92 | 244360 | Top 50 - United Kingdom | 0.445 |
| 33 | True | 93 | 215466 | Top 50 - United Kingdom | 0.855 |
| 38 | True | 86 | 272373 | Top 50 - United Kingdom | 0.538 |
| 46 | True | 90 | 292799 | Top 50 - United Kingdom | 0.690 |
| 47 | True | 86 | 251094 | Top 50 - United Kingdom | 0.402 |
| 48 | True | 82 | 310490 | Top 50 - United Kingdom | 0.483 |

| | loudness | speechiness | playlist_type |
|----|----------|-------------|---------------|
| 1 | -7.683 | 0.1940 | mixed_pop |
| 3 | -5.804 | 0.0375 | mixed_pop |
| 5 | -8.067 | 0.2410 | mixed_pop |
| 8 | -5.745 | 0.0578 | mixed_pop |
| 10 | -8.399 | 0.2710 | mixed_pop |
| 12 | -6.747 | 0.0903 | mixed_pop |
| 15 | -0.653 | 0.0946 | mixed_pop |
| 16 | -3.446 | 0.0955 | mixed_pop |
| 17 | -5.719 | 0.1810 | mixed_pop |
| 18 | -7.779 | 0.3200 | mixed_pop |
| 22 | -6.764 | 0.1720 | mixed_pop |
| 30 | -4.560 | 0.0528 | mixed_pop |
| 31 | -8.532 | 0.0400 | mixed_pop |


```

33    -6.276      0.1890    mixed_pop
38    -5.355      0.1140    mixed_pop
46    -8.492      0.3390    mixed_pop
47    -7.322      0.0471    mixed_pop
48    -9.243      0.0502    mixed_pop

```

```
[ ]: len(uk_popular_explicit)
```

```
[ ]: 18
```

```
[ ]: uk_popular_explicit.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 18 entries, 1 to 48
Data columns (total 14 columns):
#   Column          Non-Null Count  Dtype
---  -
0   track_id        18 non-null    object
1   track_name      18 non-null    object
2   artists         18 non-null    object
3   genre           18 non-null    object
4   release_year    18 non-null    int64
5   release_date    18 non-null    object
6   explicit        18 non-null    bool
7   popularity      18 non-null    int64
8   duration_ms     18 non-null    int64
9   playlist_name   18 non-null    object
10  danceability    18 non-null    float64
11  loudness        18 non-null    float64
12  speechiness     18 non-null    float64
13  playlist_type   18 non-null    object
dtypes: bool(1), float64(3), int64(3), object(7)
memory usage: 2.0+ KB

```

The same result could be achieved all in one go, but exploring the data and understanding each column is an important part of the process.

```
[ ]: songs_df[
    (songs_df['playlist_name'] == 'Top 50 - United Kingdom') &
    (songs_df['explicit'] == True) &
    (songs_df['popularity'] > avg_popularity)
].sort_values('popularity', ascending=False)
```

```

[ ]:          track_id          track_name \
8    1kuGVB7EU95pJ0bxwfwKS          vampire
12   4rXLjWdF2ZZpXCVTfWcshS          fukumean
5    2FDTHlrBguDzQkp7PVj16Q          Sprinter
16   3IX0yuEVvDbnqUwMBB3ouC          bad idea right?

```

| | | |
|----|-------------------------|---------------------------------------------------|
| 3 | 59NraMJSLaMCVtwXTSia8i | Prada |
| 33 | 6WzRpISELf3YglGAh7TXcG | Popular (with Playboi Carti & Madonna) - Music... |
| 31 | 3JvKfv6T31z00ini8iNIIt0 | Another Love |
| 17 | 2gyxAWHebV7xPYVxqoi86f | get him back! |
| 46 | 68Dni7IE4VyPkTOH9mRWHR | No Role Modelz |
| 18 | 7aqfrAY2p9BUSiupwk3svU | First Person Shooter (feat. J. Cole) |
| 10 | 2YSzYUF3jWqb9YP9VXmpjE | IDGAF (feat. Yeat) |
| 1 | 56y1jOTK0XSvJzVv9vHQBK | Paint The Town Red |
| 38 | 5mHdCZtVyb4DcJw8799hZp | Escapism. |
| 47 | 3eP13S8D5m2cweMEg3ZDed | Virginia Beach |
| 30 | 1yeB8MUNeLo9Ek1UEpsyz6 | Rich Baby Daddy (feat. Sexyy Red & SZA) |
| 22 | 4RoKNqyZ9622tcAeNPNv5k | City Boys |
| 48 | 0AYt6NMyyLd0rLuvrOUkMH | Slime You Out (feat. SZA) |
| 15 | 602d2gJewoiF1FivuOMMwE | Disconnect |

| | artists | genre | release_year | release_date | \ |
|----|----------------|---------------------------|--------------|--------------|---|
| 8 | Olivia Rodrigo | pop | 2023 | 2023-09-08 | |
| 12 | Gunna | atl hip hop | 2023 | 2023-06-16 | |
| 5 | Dave | uk hip hop | 2023 | 2023-06-01 | |
| 16 | Olivia Rodrigo | pop | 2023 | 2023-09-08 | |
| 3 | cassö | ***OOPS!*** | 2023 | 2023-08-11 | |
| 33 | The Weeknd | canadian contemporary r&b | 2023 | 2023-06-02 | |
| 31 | Tom Odell | chill pop | 2013 | 2013-06-24 | |
| 17 | Olivia Rodrigo | pop | 2023 | 2023-09-08 | |
| 46 | J. Cole | conscious hip hop | 2014 | 2014-12-09 | |
| 18 | Drake | canadian hip hop | 2023 | 2023-10-06 | |
| 10 | Drake | canadian hip hop | 2023 | 2023-10-06 | |
| 1 | Doja Cat | dance pop | 2023 | 2023-09-20 | |
| 38 | RAYE | uk contemporary r&b | 2023 | 2023-02-03 | |
| 47 | Drake | canadian hip hop | 2023 | 2023-10-06 | |
| 30 | Drake | canadian hip hop | 2023 | 2023-10-06 | |
| 22 | Burna Boy | afrobeats | 2023 | 2023-08-24 | |
| 48 | Drake | canadian hip hop | 2023 | 2023-10-06 | |
| 15 | Becky Hill | pop dance | 2023 | 2023-07-14 | |

| | explicit | popularity | duration_ms | playlist_name | danceability | \ |
|----|----------|------------|-------------|-------------------------|--------------|---|
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| 5 | True | 94 | 229133 | Top 50 - United Kingdom | 0.916 | |
| 16 | True | 94 | 184783 | Top 50 - United Kingdom | 0.627 | |
| 3 | True | 94 | 132359 | Top 50 - United Kingdom | 0.638 | |
| 33 | True | 93 | 215466 | Top 50 - United Kingdom | 0.855 | |
| 31 | True | 92 | 244360 | Top 50 - United Kingdom | 0.445 | |
| 17 | True | 92 | 211141 | Top 50 - United Kingdom | 0.546 | |
| 46 | True | 90 | 292799 | Top 50 - United Kingdom | 0.690 | |
| 18 | True | 89 | 247444 | Top 50 - United Kingdom | 0.470 | |
| 10 | True | 89 | 260111 | Top 50 - United Kingdom | 0.663 | |

| | | | | | |
|----|------|----|--------|-------------------------|-------|
| 1 | True | 87 | 230480 | Top 50 - United Kingdom | 0.864 |
| 38 | True | 86 | 272373 | Top 50 - United Kingdom | 0.538 |
| 47 | True | 86 | 251094 | Top 50 - United Kingdom | 0.402 |
| 30 | True | 84 | 319191 | Top 50 - United Kingdom | 0.645 |
| 22 | True | 84 | 153346 | Top 50 - United Kingdom | 0.808 |
| 48 | True | 82 | 310490 | Top 50 - United Kingdom | 0.483 |
| 15 | True | 80 | 164914 | Top 50 - United Kingdom | 0.504 |

| | loudness | speechiness | playlist_type |
|----|----------|-------------|---------------|
| 8 | -5.745 | 0.0578 | mixed_pop |
| 12 | -6.747 | 0.0903 | mixed_pop |
| 5 | -8.067 | 0.2410 | mixed_pop |
| 16 | -3.446 | 0.0955 | mixed_pop |
| 3 | -5.804 | 0.0375 | mixed_pop |
| 33 | -6.276 | 0.1890 | mixed_pop |
| 31 | -8.532 | 0.0400 | mixed_pop |
| 17 | -5.719 | 0.1810 | mixed_pop |
| 46 | -8.492 | 0.3390 | mixed_pop |
| 18 | -7.779 | 0.3200 | mixed_pop |
| 10 | -8.399 | 0.2710 | mixed_pop |
| 1 | -7.683 | 0.1940 | mixed_pop |
| 38 | -5.355 | 0.1140 | mixed_pop |
| 47 | -7.322 | 0.0471 | mixed_pop |
| 30 | -4.560 | 0.0528 | mixed_pop |
| 22 | -6.764 | 0.1720 | mixed_pop |
| 48 | -9.243 | 0.0502 | mixed_pop |
| 15 | -0.653 | 0.0946 | mixed_pop |

1.5 Exercises 2

Take a look at section 2 of the exercises sheet. Complete the tasks.

If there is time, work through the appropriate chapter of the McLevey textbook OR the recommended DataCamp course.

See Moodle for details.