In [2]: #import sys !{sys.executable} -m pip install PyAthena

Requirement already satisfied: PyAthena in /Users/thamilventhananthonysmariathas/anaconda3/lib/pytho n3.6/site-packages (2.3.0)

Requirement already satisfied: botocore>=1.5.52 in /Users/thamilventhananthonysmariathas/anaconda3/lib/python3.6/site-packages (from PyAthena) (1.21.56)

Requirement already satisfied: boto3>=1.4.4 in /Users/thamilventhananthonysmariathas/anaconda3/lib/python3.6/site-packages (from PyAthena) (1.10.37)

Requirement already satisfied: tenacity>=4.1.0 in /Users/thamilventhananthonysmariathas/anaconda3/lib/python3.6/site-packages (from PyAthena) (8.0.1)

Requirement already satisfied: python-dateutil<3.0.0,>=2.1 in /Users/thamilventhananthonysmariathas/anaconda3/lib/python3.6/site-packages (from botocore>=1.5.52->PyAthena) (2.8.0)

Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /Users/thamilventhananthonysmariathas/anaconda3/lib/python3.6/site-packages (from botocore>=1.5.52->PyAthena) (0.9.4)

Requirement already satisfied: urllib3<1.27,>=1.25.4 in /Users/thamilventhananthonysmariathas/anacon da3/lib/python3.6/site-packages (from botocore>=1.5.52->PyAthena) (1.25.6)

Requirement already satisfied: s3transfer<0.3.0,>=0.2.0 in /Users/thamilventhananthonysmariathas/ana conda3/lib/python3.6/site-packages (from boto3>=1.4.4->PyAthena) (0.2.1)

Requirement already satisfied: six>=1.5 in /Users/thamilventhananthonysmariathas/anaconda3/lib/pytho n3.6/site-packages (from python-dateutil<3.0.0,>=2.1->botocore>=1.5.52->PyAthena) (1.12.0)

In [2]: !aws configure get region

us-west-2

In [3]: from pyathena import connect

import numpy as np
import pandas as pd
import seaborn as sns

import matplotlib.pyplot as plt

%matplotlib inline
%config InlineBackend.figure_format='retina'

```
In [4]: # Set Athena database & table
    database_name = "amazonreviewsdb"
    table_name = "amazon_reviews_parquet"

In [5]: # Set S3 staging directory -- this is a temporary directory used for Athena queries
    s3_staging_dir = "s3://athena-query-2060/queryresults/"

In [10]: conn = connect(region_name='us-west-2', s3_staging_dir=s3_staging_dir)
```

Set Seaborn Parameters

```
In [11]: sns.set style = "seaborn-whitegrid"
         sns.set(
             rc={
                  "font.style": "normal",
                  "axes.facecolor": "white",
                  "grid.color": ".8",
                  "grid.linestyle": "-",
                  "figure.facecolor": "white",
                  "figure.titlesize": 20,
                  "text.color": "black",
                  "xtick.color": "black",
                  "ytick.color": "black",
                  "axes.labelcolor": "black",
                  "axes.grid": True,
                  "axes.labelsize": 10,
                  "xtick.labelsize": 10,
                  "font.size": 10,
                  "ytick.labelsize": 10,
```

Helper Code to Display Values on Bars

1. Which Product Categories are Highest Rated by Average Rating?

```
In [14]: df = pd.read_sql(statement, conn)
    df.head(5)
```

Out[14]:

	product_category	avg_star_rating
0	Gift_Card	4.731363
1	Digital_Music_Purchase	4.636946
2	Music	4.440541
3	Books	4.340540
4	Digital_Ebook_Purchase	4.312491

```
In [15]: # Store number of categories
   num_categories = df.shape[0]
   print(num_categories)

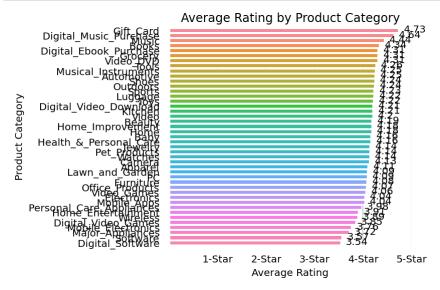
# Store average star ratings
   average_star_ratings = df
```

43

Visualization for a Subset of Product Categories

19/10/2021, 10:40

```
In [18]: # Create plot
         barplot = sns.barplot(y="product category", x="avg star rating", data=df, saturation=1)
         if num categories < 10:</pre>
             sns.set(rc={"figure.figsize": (50.0, 20.0)})
         # Set title and x-axis ticks
         plt.title("Average Rating by Product Category")
         plt.xticks([1, 2, 3, 4, 5], ["1-Star", "2-Star", "3-Star", "4-Star", "5-Star"])
         # Helper code to show actual values afters bars
         show values barplot(barplot, 0.1)
         plt.xlabel("Average Rating")
         plt.ylabel("Product Category")
         # Export plot if needed
         plt.tight layout()
         # plt.savefig('avg ratings per category.png', dpi=300)
         # Show graphic
         plt.show(barplot)
```



2. Which Product Categories Have the Most Reviews?

Out[20]:

	product_category	count_star_rating
0	Books	20726160
1	Digital_Ebook_Purchase	19180765
2	Wireless	9038249
3	Video_DVD	7135819
4	PC	7004337

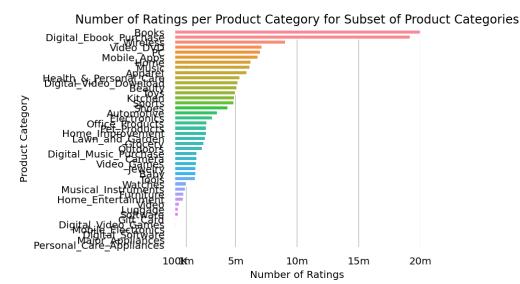
```
In [21]: # Store counts
    count_ratings = df["count_star_rating"]

# Store max ratings
    max_ratings = df["count_star_rating"].max()
    print(max_ratings)
```

20726160

Visualization for a Subset of Product Categories

```
In [22]: # Create Seaborn barplot
         barplot = sns.barplot(y="product_category", x="count_star_rating", data=df, saturation=1)
         if num categories < 10:</pre>
             sns.set(rc={"figure.figsize": (10.0, 5.0)})
         # Set title
         plt.title("Number of Ratings per Product Category for Subset of Product Categories")
         # Set x-axis ticks to match scale
         if max ratings > 200000:
             plt.xticks([100000, 1000000, 5000000, 10000000, 15000000, 20000000], ["100K", "1m", "5m", "10m",
          "15m", "20m"])
             plt.xlim(0, 20000000)
         elif max ratings <= 200000:</pre>
             plt.xticks([50000, 100000, 150000, 200000], ["50K", "100K", "150K", "200K"])
             plt.xlim(0, 200000)
         plt.xlabel("Number of Ratings")
         plt.ylabel("Product Category")
         plt.tight layout()
         # Export plot if needed
         # plt.savefig('ratings per category.png', dpi=300)
         # Show the barplot
         plt.show(barplot)
```



3. When did each product category become available in the Amazon catalog based on the date of the first review?

ORDER BY first review date

```
In [24]: df = pd.read_sql(statement, conn)
    df.head()
```

Out[24]:

	product_category	first_review_date
0	Music	1322
1	Books	9305
2	Video	9445
3	Video_DVD	9685
4	Toys	9866

```
In [25]: # Convert date strings (e.g. 2014-10-18) to datetime
import datetime as datetime

dates = pd.to_datetime(df["first_review_date"])
```

```
In [26]: # See: https://stackoverflow.com/questions/60761410/how-to-graph-events-on-a-timeline

def modify_dataframe(df):
    """ Modify dataframe to include new columns """
    df["year"] = pd.to_datetime(df["first_review_date"], format="%Y-%m-%d").dt.year
    return df

def get_x_y(df):
    """ Get X and Y coordinates; return tuple """
    series = df["year"].value_counts().sort_index()
    # new_series = series.reindex(range(1,21)).fillna(0).astype(int)
    return series.index, series.values
```

```
In [27]: new_df = modify_dataframe(df)
print(new_df)

X, Y = get_x_y(new_df)
```

	product_category	first_review_date	year
0	Music	1322	1970
1	Books	9305	1970
2	Video	9445	1970
3	Video DVD	9685	1970
4	_ Toys	9866	1970
5	Sports	10143	1970
6	Video Games	10171	1970
7	Home	10375	1970
8	Office_Products	10422	1970
9	$\operatorname{\mathtt{Pet_Products}}$	10461	1970
10	Software	10490	1970
11	Home_Entertainment	10514	1970
12	Camera	10550	1970
13	Wireless	10564	1970
14	<pre>Health_&_Personal_Care</pre>	10628	1970
15	Outdoors	10674	1970
16	Electronics	10751	1970
17	PC	10773	1970
18	Baby	10785	1970
19	Digital_Ebook_Purchase	10831	1970
20	Grocery	10839	1970
21	Automotive	10888	1970
22	Shoes	10903	1970
23	Home_Improvement	10903	1970
24	Tools	10904	1970
25	Lawn_and_Garden	10922	1970
26	Musical_Instruments	10938	1970
27	Kitchen	10976	1970
28	Furniture	11033	1970
29	Digital_Music_Purchase	11136	1970
30	Major_Appliances	11195	1970
31	Apparel	11206	1970
32	Digital_Video_Download	11234	1970
33	Personal_Care_Appliances	11259	1970
34	Beauty	11261	1970
35	Watches	11417	1970
36	Jewelry	11636	1970
37	Mobile_Electronics	11678	1970
38	Luggage	11996	1970
39	Gift_Card	12705	1970
40	Digital_Video_Games	13368	1970

41 Digital_Software 13904 1970 42 Mobile_Apps 14917 1970

4. What is the breakdown of ratings (1-5) per product category?

GROUP BY product_category, star_rating

ORDER BY product category ASC, star rating DESC, count reviews

Out[30]:

	product_category	star_rating	count_reviews
0	Apparel	5	3320651
1	Apparel	4	1147254
2	Apparel	3	623483
3	Apparel	2	369608
4	Apparel	1	445464
210	Wireless	5	4845345
211	Wireless	4	1507668
212	Wireless	3	818151
213	Wireless	2	600436
214	Wireless	1	1266649

215 rows × 3 columns

Prepare for Stacked Percentage Horizontal Bar Plot Showing Proportion of Star Ratings per Product Category

```
In [31]: # Create grouped DataFrames by category and by star rating
grouped_category = df.groupby("product_category")
grouped_star = df.groupby("star_rating")

# Create sum of ratings per star rating
df_sum = df.groupby(["star_rating"]).sum()

# Calculate total number of star ratings
total = df_sum["count_reviews"].sum()
print(total)
```

160796570

```
In [32]: # Create dictionary of product categories and array of star rating distribution per category
distribution = {}
count_reviews_per_star = []
i = 0

for category, ratings in grouped_category:
    count_reviews_per_star = []
    for star in ratings["star_rating"]:
        count_reviews_per_star.append(ratings.at[i, "count_reviews"])
        i = i + 1
        distribution[category] = count_reviews_per_star

# Check if distribution has been created succesfully
print(distribution)
```

{'Apparel': [3320651, 1147254, 623483, 369608, 445464], 'Automotive': [2301688, 526898, 240023, 1478 43, 300024], 'Baby': [1078545, 289129, 150753, 101427, 145039], 'Beauty': [3254946, 741443, 398405, 264029, 456898], 'Books': [13662131, 3546319, 1543611, 861867, 1112232], 'Camera': [1085596, 344163, 144596, 92396, 172009], 'Digital Ebook Purchase': [11612150, 4311745, 1701301, 749678, 805891], 'Dig ital Music Purchase': [1467877, 231676, 70507, 28743, 53381], 'Digital Software': [46410, 16693, 830 8, 6890, 23783], 'Digital Video Download': [3122980, 984221, 449260, 251930, 365352], 'Digital Video Games': [80677, 20406, 11629, 7749, 24970], 'Electronics': [1796672, 542181, 240859, 180668, 36055 8], 'Furniture': [447763, 153678, 73574, 43853, 73346], 'Gift Card': [129709, 9859, 3156, 1569, 479 3], 'Grocery': [1662278, 293389, 161497, 105265, 180049], 'Health & Personal Care': [3359016, 78202 5, 400572, 278066, 513204], 'Home': [3897623, 960326, 500884, 324345, 545389], 'Home Entertainment': [395254, 138172, 60898, 46242, 103134], 'Home Improvement': [1660254, 419488, 189490, 124792, 24663 0], 'Jewelry': [1081479, 270579, 159735, 100852, 155108], 'Kitchen': [3129752, 732976, 350188, 24236 1, 427554], 'Lawn and Garden': [1546990, 401137, 194777, 135055, 281302], 'Luggage': [216857, 61488, 27872, 17887, 25028, 'Major Appliances': [49704, 15158, 6718, 5475, 19846, 'Mobile Apps': [374738 3, 1375672, 614333, 314750, 755028], 'Mobile Electronics': [52373, 18112, 9754, 7320, 17600], 'Musi c': [4296015, 1018968, 405541, 202861, 254396], 'Musical Instruments': [582772, 161623, 68231, 4090 6, 67197], 'Office Products': [1586632, 419489, 194113, 138717, 307540], 'Outdoors': [1436837, 41810 3, 179395, 109599, 161662], 'PC': [4166576, 1185891, 520947, 366603, 764320], 'Personal Care Applian ces': [49310, 13782, 7107, 5381, 11106], 'Pet Products': [1645587, 381294, 216639, 151284, 248866], 'Shoes': [2647584, 850452, 405263, 243424, 232752], 'Software': [154057, 58696, 30693, 24672, 7401 7], 'Sports': [3046077, 838841, 381388, 229890, 363858], 'Tools': [1118325, 300861, 126637, 75055, 1 27732], 'Toys': [3160447, 788495, 395143, 233754, 403762], 'Video': [263744, 81794, 39030, 21697, 31 144], 'Video DVD': [4702079, 1138174, 540591, 296845, 458130], 'Video Games': [1040615, 321861, 1554 96, 95999, 194515], 'Watches': [582164, 176290, 80381, 52864, 86343], 'Wireless': [4845345, 1507668, 818151, 600436, 12666491}

In [33]: # Check if distribution keys are set correctly to product categories
 print(distribution.keys())

dict_keys(['Apparel', 'Automotive', 'Baby', 'Beauty', 'Books', 'Camera', 'Digital_Ebook_Purchase',
'Digital_Music_Purchase', 'Digital_Software', 'Digital_Video_Download', 'Digital_Video_Games', 'Elec
tronics', 'Furniture', 'Gift_Card', 'Grocery', 'Health_&_Personal_Care', 'Home', 'Home_Entertainmen
t', 'Home_Improvement', 'Jewelry', 'Kitchen', 'Lawn_and_Garden', 'Luggage', 'Major_Appliances', 'Mob
ile_Apps', 'Mobile_Electronics', 'Music', 'Musical_Instruments', 'Office_Products', 'Outdoors', 'P
C', 'Personal_Care_Appliances', 'Pet_Products', 'Shoes', 'Software', 'Sports', 'Tools', 'Toys', 'Vid
eo', 'Video_DVD', 'Video_Games', 'Watches', 'Wireless'])

In [34]: # Check if star rating distributions are set correctly
print(distribution.items())

dict items([('Apparel', [3320651, 1147254, 623483, 369608, 445464]), ('Automotive', [2301688, 52689 8, 240023, 147843, 300024]), ('Baby', [1078545, 289129, 150753, 101427, 145039]), ('Beauty', [325494 6, 741443, 398405, 264029, 456898]), ('Books', [13662131, 3546319, 1543611, 861867, 1112232]), ('Cam era', [1085596, 344163, 144596, 92396, 172009]), ('Digital_Ebook_Purchase', [11612150, 4311745, 1701 301, 749678, 805891]), ('Digital Music Purchase', [1467877, 231676, 70507, 28743, 53381]), ('Digital _Software', [46410, 16693, 8308, 6890, 23783]), ('Digital_Video_Download', [3122980, 984221, 449260, 251930, 365352]), ('Digital Video Games', [80677, 20406, 11629, 7749, 24970]), ('Electronics', [1796 672, 542181, 240859, 180668, 360558]), ('Furniture', [447763, 153678, 73574, 43853, 73346]), ('Gift Card', [129709, 9859, 3156, 1569, 4793]), ('Grocery', [1662278, 293389, 161497, 105265, 180049]), ('Health & Personal Care', [3359016, 782025, 400572, 278066, 513204]), ('Home', [3897623, 960326, 50 0884, 324345, 545389]), ('Home Entertainment', [395254, 138172, 60898, 46242, 103134]), ('Home Impro vement', [1660254, 419488, 189490, 124792, 246630]), ('Jewelry', [1081479, 270579, 159735, 100852, 1 55108]), ('Kitchen', [3129752, 732976, 350188, 242361, 427554]), ('Lawn and Garden', [1546990, 40113 7, 194777, 135055, 281302]), ('Luggage', [216857, 61488, 27872, 17887, 25028]), ('Major Appliances', [49704, 15158, 6718, 5475, 19846]), ('Mobile Apps', [3747383, 1375672, 614333, 314750, 755028]), ('M obile Electronics', [52373, 18112, 9754, 7320, 17600]), ('Music', [4296015, 1018968, 405541, 202861, 254396]), ('Musical Instruments', [582772, 161623, 68231, 40906, 67197]), ('Office Products', [15866 32, 419489, 194113, 138717, 307540]), ('Outdoors', [1436837, 418103, 179395, 109599, 161662]), ('P C', [4166576, 1185891, 520947, 366603, 764320]), ('Personal Care Appliances', [49310, 13782, 7107, 5 381, 11106]), ('Pet Products', [1645587, 381294, 216639, 151284, 248866]), ('Shoes', [2647584, 85045 2, 405263, 243424, 232752]), ('Software', [154057, 58696, 30693, 24672, 74017]), ('Sports', [304607] 7, 838841, 381388, 229890, 363858]), ('Tools', [1118325, 300861, 126637, 75055, 127732]), ('Toys', [3160447, 788495, 395143, 233754, 403762]), ('Video', [263744, 81794, 39030, 21697, 31144]), ('Video DVD', [4702079, 1138174, 540591, 296845, 458130]), ('Video Games', [1040615, 321861, 155496, 95999, 194515]), ('Watches', [582164, 176290, 80381, 52864, 86343]), ('Wireless', [4845345, 1507668, 81815 1, 600436, 12666491)1)

Out[36]:

	5	4	3	2	1
Gift_Card	87.002804	6.612962	2.116899	1.052413	3.214923
Digital_Music_Purchase	79.251144	12.508261	3.806695	1.551844	2.882057
Music	69.539775	16.494078	6.564509	3.283720	4.117919
Books	65.917329	17.110352	7.447646	4.158354	5.366320
Digital_Ebook_Purchase	60.540599	22.479526	8.869829	3.908489	4.201558
Grocery	69.190145	12.211933	6.722101	4.381518	7.494304
Video_DVD	65.894034	15.950152	7.575739	4.159929	6.420146
Tools	63.955084	17.205723	7.242152	4.292266	7.304774
Musical_Instruments	63.294628	17.553808	7.410541	4.442784	7.298239
Automotive	65.454392	14.983694	6.825669	4.204294	8.531951
Shoes	60.454370	19.419040	9.253689	5.558292	5.314610
Outdoors	62.319548	18.134270	7.780851	4.753608	7.011723
Sports	62.675785	17.259911	7.847403	4.730194	7.486707
Luggage	62.113184	17.611677	7.983227	5.123277	7.168635
Toys	63.442395	15.828144	7.932048	4.692347	8.105065
Digital_Video_Download	60.362101	19.023384	8.683462	4.869395	7.061657
Kitchen	64.097078	15.011292	7.171823	4.963534	8.756273
Video	60.296885	18.699661	8.922999	4.960346	7.120110
Beauty	63.626339	14.493421	7.787856	5.161130	8.931253
Home_Improvement	62.872834	15.885762	7.175874	4.725799	9.339732
Home	62.576561	15.418089	8.041721	5.207378	8.756252
Baby	61.111070	16.382240	8.541764	5.746921	8.218005
Health_&_Personal_Care	62.986868	14.664207	7.511359	5.214178	9.623388
Jewelry	61.178174	15.306380	9.036047	5.705096	8.774303
Pet_Products	62.246309	14.422905	8.194631	5.722499	9.413656

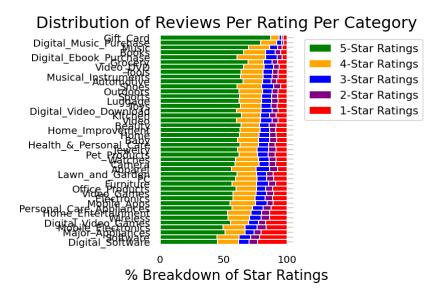
19/10/2021, 10:40 AWS_data_analytics_notebook

	5	4	3	2	1
Watches	59.523415	18.024788	8.218563	5.405085	8.828148
Camera	59.039570	18.717125	7.863778	5.024908	9.354619
Apparel	56.220663	19.423716	10.555951	6.257691	7.541979
Lawn_and_Garden	60.446746	15.673939	7.610674	5.277109	10.991532
PC	59.485659	16.930810	7.437492	5.233943	10.912096
Furniture	56.520460	19.398546	9.287137	5.535499	9.258357
Office_Products	59.952292	15.850762	7.334731	5.241544	11.620671
Video_Games	57.540672	17.797262	8.598131	5.308252	10.755682
Electronics	57.568334	17.372373	7.717520	5.788901	11.552873
Mobile_Apps	55.050560	20.209174	9.024798	4.623804	11.091664
Personal_Care_Appliances	56.883464	15.898761	8.198556	6.207461	12.811757
Home_Entertainment	53.146968	18.578997	8.188517	6.217830	13.867689
Wireless	53.609333	16.680974	9.052096	6.643278	14.014318
Digital_Video_Games	55.474417	14.031396	7.996232	5.328300	17.169654
Mobile_Electronics	49.803631	17.223443	9.275478	6.960888	16.736561
Major_Appliances	51.293588	15.642769	6.932849	5.650096	20.480697
Software	45.028132	17.155801	8.971020	7.211189	21.633858
Digital_Software	45.462560	16.352220	8.138396	6.749344	23.297481

Visualization for a Subset of Product Categories

```
In [40]: categories = df sorted distribution pct.index
         # Plot bars
         if len(categories) > 10:
             plt.figure(figsize=(10,10))
         else:
             plt.figure(figsize=(10,5))
         df sorted distribution pct.plot(kind="barh",
                                          stacked=True,
                                          edgecolor='white',
                                          width=1.0,
                                          color=['green',
                                                  'orange',
                                                 'blue',
                                                  'purple',
                                                 'red'])
         plt.title("Distribution of Reviews Per Rating Per Category",
                   fontsize='16')
         plt.legend(bbox to anchor=(1.04,1),
                    loc="upper left",
                    labels=['5-Star Ratings',
                             '4-Star Ratings',
                             '3-Star Ratings',
                             '2-Star Ratings',
                             '1-Star Ratings'])
         plt.xlabel("% Breakdown of Star Ratings", fontsize='14')
         plt.gca().invert yaxis()
         plt.tight layout()
         plt.show();
```

<Figure size 720x720 with 0 Axes>

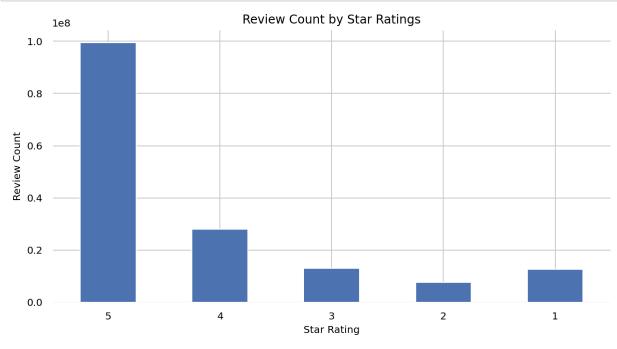


5. How Many Reviews per Star Rating? (5, 4, 3, 2, 1)

ORDER BY star rating DESC, count reviews

Out[45]:

	star_rating	count_reviews
0	5	99530924
1	4	27996469
2	3	12900929
3	2	7700647
4	1	12667601



6. How Did Star Ratings Change Over Time?

Is there a drop-off point for certain product categories throughout the year?

```
In [48]: # SQL statement
statement = """
SELECT year, ROUND(AVG(star_rating),4) AS avg_rating
FROM {}.{}
GROUP BY year
ORDER BY year
""".format(
          database_name, table_name
)
print(statement)
```

SELECT year, ROUND(AVG(star_rating),4) AS avg_rating FROM amazonreviewsdb.amazon_reviews_parquet GROUP BY year
ORDER BY year

```
In [49]: df = pd.read_sql(statement, conn)
    df
```

Out[49]:

	year	avg_rating
0	1973	5.0000
1	1995	4.6204
2	1996	4.6111
3	1997	4.4348
4	1998	4.3610
5	1999	4.2886
6	2000	4.2629
7	2001	4.2037
8	2002	4.1698
9	2003	4.1280
10	2004	4.0695
11	2005	4.0734
12	2006	4.1061
13	2007	4.1733
14	2008	4.1312
15	2009	4.1171
16	2010	4.0760
17	2011	4.0571
18	2012	4.1271
19	2013	4.2077
20	2014	4.2358
21	2015	4.2549

```
In [50]: df["year"] = pd.to_datetime(df["year"], format="%Y").dt.year
```

Visualization for a Subset of Product Categories

```
In [51]: fig = plt.gcf()
fig.set_size_inches(12, 5)

fig.suptitle("Average Star Rating Over Time (Across Subset of Product Categories)")

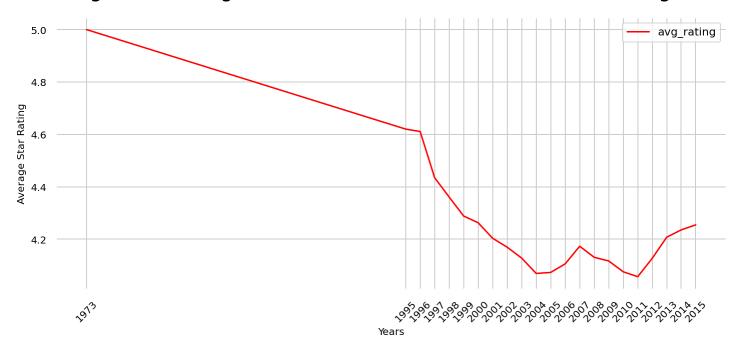
ax = plt.gca()
# ax = plt.gca().set_xticks(df['year'])
ax.locator_params(integer=True)
ax.set_xticks(df["year"].unique())

df.plot(kind="line", x="year", y="avg_rating", color="red", ax=ax)

# plt.xticks(range(1995, 2016, 1))
# plt.yticks(range(0,6,1))
plt.xlabel("Years")
plt.ylabel("Average Star Rating")
plt.ylabel("Average Star Rating")
plt.xticks(rotation=45)

# fig.savefig('average-rating.png', dpi=300)
plt.show()
```

Average Star Rating Over Time (Across Subset of Product Categories)



Average Star Rating Per Product Categories Across Time

```
In [52]: # SQL statement
    statement = """
    SELECT product_category, year, ROUND(AVG(star_rating), 4) AS avg_rating_category
    FROM {}.{}
    GROUP BY product_category, year
    ORDER BY year
    """.format(
         database_name, table_name
    )
    print(statement)
```

SELECT product_category, year, ROUND(AVG(star_rating), 4) AS avg_rating_category FROM amazonreviewsdb.amazon_reviews_parquet GROUP BY product_category, year ORDER BY year

Out[53]:

	product_category	year	avg_rating_category
0	Music	1973	5.0000
1	Video	1995	5.0000
2	Books	1995	4.6114
3	Music	1995	5.0000
4	Music	1996	4.6000
704	Software	2015	3.7407
705	Jewelry	2015	4.1817
706	Health_&_Personal_Care	2015	4.2382
707	Office_Products	2015	4.2103
708	PC	2015	4.1576

709 rows × 3 columns

Visualization

```
In [55]: fig = plt.gcf()
    fig.set_size_inches(12, 5)

fig.suptitle("Average Star Rating Over Time Across Subset Of Categories")

ax = plt.gca()

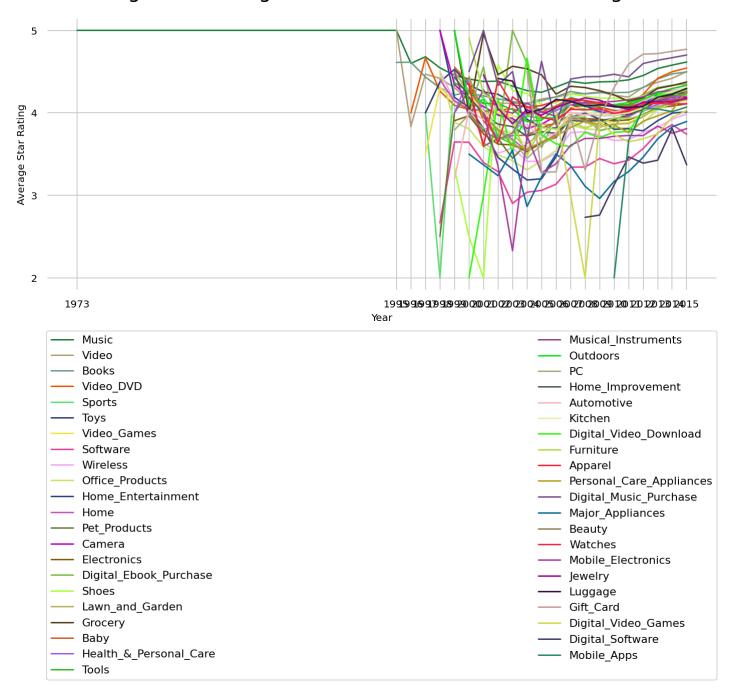
ax.locator_params(integer=True)
    ax.set_xticks(df["year"].unique())

plot_categories(df)

plt.xlabel("Year")
    plt.ylabel("Average Star Rating")
    plt.ylabel("Average Star Rating")
    plt.legend(bbox_to_anchor=(0, -0.15, 1, 0), loc=2, ncol=2, mode="expand", borderaxespad=0)

# fig.savefig('average_rating_category_all_data.png', dpi=300)
    plt.show()
```

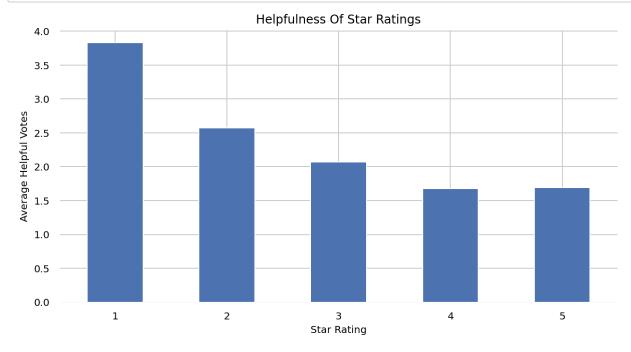
Average Star Rating Over Time Across Subset Of Categories



7. Which Star Ratings (1-5) are Most Helpful?

```
In [56]: # SQL statement
         statement = """
         SELECT star rating,
                   AVG(helpful_votes) AS avg_helpful_votes
         FROM {}.{}
         GROUP BY star_rating
         ORDER BY star_rating ASC
          """.format(
              database name, table name
         print(statement)
         SELECT star_rating,
                   AVG(helpful_votes) AS avg_helpful_votes
         FROM amazonreviewsdb.amazon reviews parquet
         GROUP BY star rating
         ORDER BY star rating ASC
In [57]: df = pd.read_sql(statement, conn)
         df
Out[57]:
             star_rating avg_helpful_votes
          0
                   1
                            3.832321
                    2
                            2.570054
          1
                    3
                            2.068739
          2
                            1.680050
                    5
                            1.691202
```

Visualization for a Subset of Product Categories



8. Which Products have Most Helpful Reviews? How Long are the Most Helpful Reviews?

```
In [59]: # SQL statement
         statement = """
         SELECT product title,
                  helpful_votes,
                  star_rating,
                  LENGTH(review_body) AS review_body_length,
                  SUBSTR(review_body, 1, 100) AS review body substr
         FROM {}.{}
         ORDER BY helpful votes DESC LIMIT 10
         """.format(
             database name, table name
         print(statement)
         SELECT product title,
                  helpful votes,
                  star_rating,
                  LENGTH (review_body) AS review_body_length,
                  SUBSTR(review_body, 1, 100) AS review_body_substr
         FROM amazonreviewsdb.amazon reviews parquet
```

ORDER BY helpful_votes DESC LIMIT 10

In [60]: df = pd.read_sql(statement, conn)
df

Out[60]:

	product_title	helpful_votes	star_rating	review_body_length	review_body_substr
0	Kindle: Amazon's Original Wireless Reading Dev	47524	5	12906	This is less a \\"pros and cons\\" review than
1	BIC Cristal For Her Ball Pen, 1.0mm, Black, 16	41393	5	863	Someone has answered my gentle prayers and FIN
2	The Mountain Kids 100% Cotton Three Wolf Moon	41278	5	1566	This item has wolves on it which makes it intr
3	Kindle Keyboard 3G, Free 3G + Wi-Fi, 6" E Ink	31924	5	23069	UPDATE NOVEMBER 2011: />dbr />dby review is
4	Kindle Fire HD 7", Dolby Audio, Dual-Band Wi-Fi	31417	4	13594	I've been an iPad user since the original came
5	Kindle Fire (Previous Generation - 1st)	28611	4	29778	UPDATE November 2012 - With the [[ASIN:B007T36
6	Fifty Shades of Grey: Book One of the Fifty Sh	27550	2	2849	I really don't like writing bad reviews. I adm
7	Fifty Shades of Grey: Book One of the Fifty Sh	27550	2	2849	I really don't like writing bad reviews. I adm
8	Wheelmate Laptop Steering Wheel Desk	26132	5	572	My husband Brad always warns me not to try and
9	Kindle Wireless Reading Device (6" Display, U	24714	1	10222	I was DELIGHTED to upgrade my Kindle 1 to K2

9. What is the Ratio of Positive (5, 4) to Negative (3, 2, 1) Reviews?

```
In [61]: # SQL statement
         statement = """
         SELECT (CAST(positive review count AS DOUBLE) / CAST(negative review count AS DOUBLE)) AS positive to
          negative sentiment ratio
         FROM (
           SELECT count(*) AS positive review count
           FROM {}.{}
           WHERE star rating >= 4
           SELECT count(*) AS negative review count
           FROM { } . { }
           WHERE star rating < 4
         """.format(
             database name, table name, database name, table name
         print(statement)
         SELECT (CAST(positive_review_count AS DOUBLE) / CAST(negative_review_count AS DOUBLE)) AS positive t
         o_negative_sentiment_ratio
         FROM (
           SELECT count(*) AS positive_review_count
           FROM amazonreviewsdb.amazon reviews parquet
           WHERE star rating >= 4
         ), (
           SELECT count(*) AS negative_review count
           FROM amazonreviewsdb.amazon reviews parquet
           WHERE star rating < 4
In [62]: | df = pd.read sql(statement, conn)
         df
Out[62]:
            positive to negative sentiment ratio
                                  3.8332
```

localhost:8888/nbconvert/html/AnacondaProjects/Freelance_notebooks/AWS_data_analytics_notebook.ipynb?download=false

0

10. Which Customers are Abusing the Review System by Repeatedly Reviewing the Same Product? What Was Their Average Star Rating for Each Product?

```
In [65]: # SQL statement
    statement = """
    SELECT customer_id, product_category, product_title,
    ROUND(AVG(star_rating),4) AS avg_star_rating, COUNT(*) AS review_count
    FROM {}.{}
    GROUP BY customer_id, product_category, product_title
    HAVING COUNT(*) > 1
    ORDER BY review_count DESC
    LIMIT 5
    """.format(
        database_name, table_name
    )
    print(statement)
```

```
SELECT customer_id, product_category, product_title,
ROUND(AVG(star_rating),4) AS avg_star_rating, COUNT(*) AS review_count
FROM amazonreviewsdb.amazon_reviews_parquet
GROUP BY customer_id, product_category, product_title
HAVING COUNT(*) > 1
ORDER BY review_count DESC
LIMIT 5
```

Out[66]:

	customer_id	product_category	product_title	avg_star_rating	review_count
0	38118182	Video_DVD	Pearl Harbor	4.2308	260
1	33132919	Video_DVD	Shania Twain - Up (Live in Chicago)	5.0000	220
2	52895956	Books	Frankenstein	3.0299	201
3	29088361	Music	In The Zone	4.9672	122
4	23974294	Video_DVD	Shania Twain - Up (Live in Chicago)	5.0000	114