```
In [1]: # pip install pyathena
         # pip install PyAthena[Pandas]
 In [2]:
         import sqlalchemy
In [3]:
         from urllib.parse import quote plus
         from sqlalchemy.engine import create engine
         from sqlalchemy.sql.expression import select
         from sqlalchemy.sql.functions import func
         from sqlalchemy.sql.schema import Table, MetaData
         import sqlalchemy
         import athena
 In [4]:
         from pyathena import connect
         import pandas as pd
 In [5]:
         import numpy as np
         import matplotlib.pyplot as plt
         %matplotlib inline
         %config InlineBackend.figure format='retina'
         import seaborn as sns
 In [6]: | database = 'dsoaws'
         table = 'amazon reviews tsv'
         bucket = 'data-science-on-aws22'
         create connection engine
         engine = create engine("awsathena+rest://AKIAQUEPHTPDTCLPBSHY:rcpmIAV4pUgVxOmL7F8PLDdhn5
 In [7]:
                                 "default?s3 staging dir=s3://data-science-on-aws22/athena/staging
         # conn str = "awsathena+rest://{aws access key id}:{aws secret access key}@athena.us-eas
In [8]:
                      "{schema name}?s3 staging dir={s3 staging dir}"
         # engine = create engine(conn str.format(
               aws access key id=quote plus("AKIAQUEPHTPDTCLPBSHY"),
         #
               aws secret access key=quote plus("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"),
               region name="us-east-1",
         #
               schema name="default",
               s3 staging dir=quote plus("s3://{0}/path/to/").format(bucket)))
               session token = 'kiane'
         the sql statement
         sql_statement="""
In [9]:
         SELECT DISTINCT product category from {0}.{1}
         ORDER BY product category
         """.format(database, table)
        pd.read sql(sql statement, con=engine)
In [10]:
Out[10]:
                 product_category
          0
                         Apparel
                      Automotive
```

2	Baby
3	Beauty
4	Books
5	Camera
6	Digital_Ebook_Purchase
7	Digital_Music_Purchase
8	Digital_Software
9	Digital_Video_Download
10	Digital_Video_Games
11	Electronics
12	Furniture
13	Gift Card
14	Grocery
15	Health & Personal Care
16	Home
17	Home Entertainment
18	Home Improvement
19	Jewelry
20	Kitchen
21	Lawn and Garden
22	Luggage
23	Major Appliances
24	Mobile_Apps
25	Mobile_Electronics
26	Music
27	Musical Instruments
28	Office Products
29	Outdoors
30	PC
31	Personal_Care_Appliances
32	Pet Products
33	Shoes
34	Software
35	Sports
36	Tools
37	Toys
38	Video
39	Video DVD

40	Video Games
41	Watches
42	Wireless

Which product categories are the highest rated by average rating?

```
In [11]: sql2 = """SELECT product_category, AVG(star_rating) AS avg_star_rating
        FROM {0}.{1}
        GROUP BY product category
        ORDER BY avg star rating DESC
         """.format(database, table)
```

In [12]: pd.read_sql(sql2, con=engine)

Out[12]:	product_ca

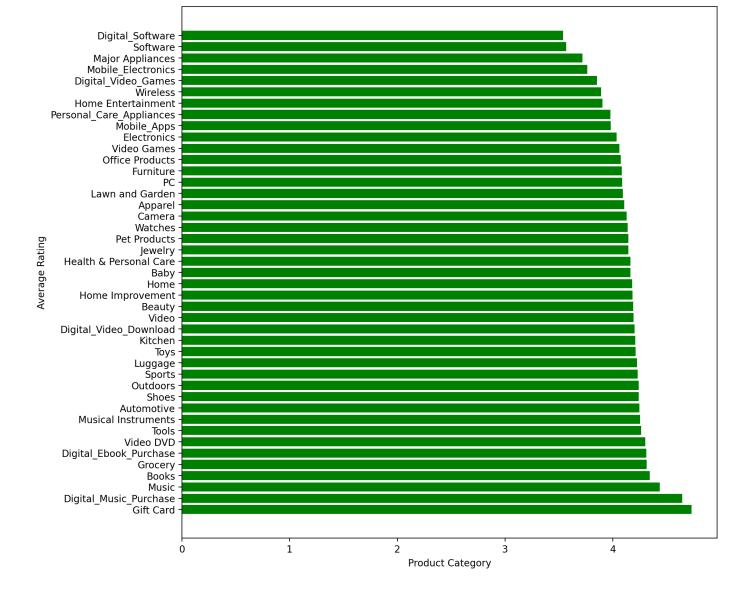
	product_category	avg_star_rating
0	Gift Card	4.731363
1	Digital_Music_Purchase	4.642891
2	Music	4.436624
3	Books	4.341658
4	Grocery	4.312219
5	Digital_Ebook_Purchase	4.308775
6	Video DVD	4.302017
7	Tools	4.261769
8	Musical Instruments	4.251103
9	Automotive	4.246302
10	Shoes	4.241260
11	Outdoors	4.240019
12	Sports	4.229365
13	Luggage	4.223391
14	Toys	4.211735
15	Kitchen	4.207424
16	Digital_Video_Download	4.201208
17	Video	4.191511
18	Beauty	4.187224
19	Home Improvement	4.182270
20	Home	4.178399
21	Baby	4.162683
22	Health & Personal Care	4.161833
23	Jewelry	4.144090
24	Pet Products	4.143653

25	Watches	4.138283
26	Camera	4.127015
27	Apparel	4.105229
28	Lawn and Garden	4.093177
29	PC	4.086444
30	Furniture	4.083949
31	Office Products	4.072539
32	Video Games	4.059893
33	Electronics	4.035507
34	Mobile_Apps	3.981594
35	Personal_Care_Appliances	3.977402
36	Home Entertainment	3.902123
37	Wireless	3.891779
38	Digital_Video_Games	3.853126
39	Mobile_Electronics	3.763163
40	Major Appliances	3.716185
41	Software	3.567035
42	Digital_Software	3.539330

```
In [13]: result = pd.read_sql(sql2, con=engine)
```

Set the size of plot canvas

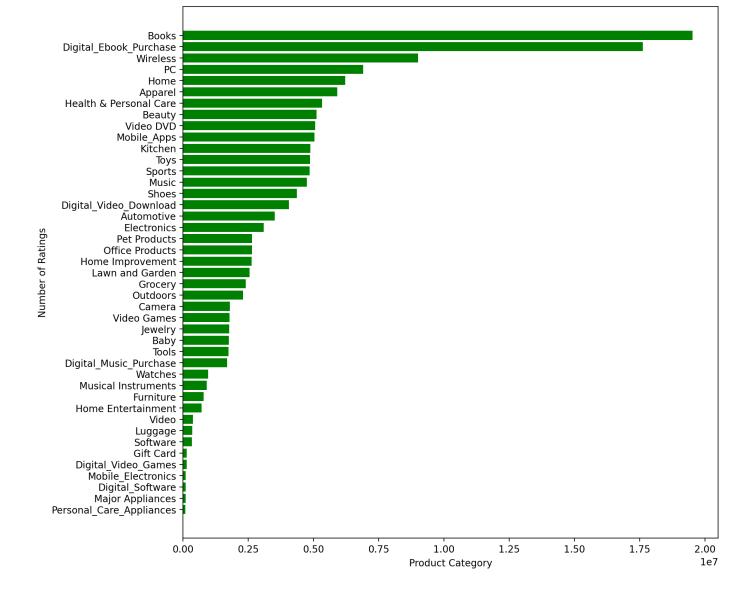
```
In [14]: plt.rcParams['figure.figsize'] = [10, 10]
In [15]: plt.barh(result['product_category'], result['avg_star_rating'], color ='green')
    plt.xlabel("Product Category")
    plt.ylabel("Average Rating")
    plt.show()
```



Which product categories have the most reviews?

```
In [16]: sql3 = """SELECT product_category, COUNT(star_rating) AS count_star_rating
    FROM {0}.{1}
    GROUP BY product_category
    ORDER BY count_star_rating
    """.format(database,table)
    result2 = pd.read_sql(sql3, con=engine)

In [17]: plt.barh(result2['product_category'],result2['count_star_rating'], color ='green')
    plt.xlabel("Product Category")
    plt.ylabel("Number of Ratings")
    plt.show()
```



When did each product category become available in the Amazon catalog

I need to check first the column schema

```
In [18]: sql_test = """SELECT *
FROM {0}.{1}
LIMIT 3
    """.format(database, table)

result_all = pd.read_sql(sql_test, con=engine)
result_all
```

Out[18]:		marketplace	customer_id	review_id	product_id	product_parent	product_title	product_category	sta
	0	US	12076615	RQ58W7SMO911M	0385730586	122662979	Sisterhood of the Traveling Pants (Book 1)	Books	
	1	US	12703090	RF6IUKMGL8SF	0811828964	56191234	The Bad Girl's Guide to Getting What You Want	Books	
	2	US	12257412	R1DOSHH6AI622S	1844161560	253182049	Eisenhorn (A	Books	

Then run the query after investigation.

```
In [21]: sql4 = """SELECT product_category, MIN(EXTRACT(YEAR FROM CAST(review_date AS DATE))) AS
FROM {0}.{1}
GROUP BY product_category
""".format(database,table)

result4 = pd.read_sql(sql4, con=engine)
result4
```

Out[21]: product_category release_year

0 Wireless 1998 1 Personal_Care_Appliances 2000 2 PC 1999 3 Home Improvement 1999 4 Mobile_Apps 2010 5 Gift Card 2004 6 Tools 1999 7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1997 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music		product_category	release_year
2 PC 1999 3 Home Improvement 1999 4 Mobile_Apps 2010 5 Gift Card 2004 6 Tools 1999 7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997 <th>0</th> <th>Wireless</th> <th>1998</th>	0	Wireless	1998
3 Home Improvement 1999 4 Mobile_Apps 2010 5 Gift Card 2004 6 Tools 1999 7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	1	Personal_Care_Appliances	2000
4 Mobile_Apps 2010 5 Gift Card 2004 6 Tools 1999 7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	2	PC	1999
5 Gift Card 2004 6 Tools 1999 7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	3	Home Improvement	1999
6 Tools 1999 7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	4	Mobile_Apps	2010
7 Beauty 2000 8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	5	Gift Card	2004
8 Automotive 1999 9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	6	Tools	1999
9 Major Appliances 2000 10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	7	Beauty	2000
10 Books 1995 11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	8	Automotive	1999
11 Office Products 1998 12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	9	Major Appliances	2000
12 Musical Instruments 1999 13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	10	Books	1995
13 Digital_Music_Purchase 2000 14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	11	Office Products	1998
14 Digital_Ebook_Purchase 1999 15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	12	Musical Instruments	1999
15 Electronics 1999 16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	13	Digital_Music_Purchase	2000
16 Sports 1997 17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	14	Digital_Ebook_Purchase	1999
17 Home 1998 18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	15	Electronics	1999
18 Mobile_Electronics 2001 19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	16	Sports	1997
19 Baby 1999 20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	17	Home	1998
20 Digital_Software 2008 21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	18	Mobile_Electronics	2001
21 Jewelry 2001 22 Music 1995 23 Digital_Video_Games 2006 24 Health & Personal Care 1999 25 Video Games 1997	19	Baby	1999
 Music 1995 Digital_Video_Games 2006 Health & Personal Care 1999 Video Games 1997 	20	Digital_Software	2008
23Digital_Video_Games200624Health & Personal Care199925Video Games1997	21	Jewelry	2001
24 Health & Personal Care 199925 Video Games 1997	22	Music	1995
Video Games 1997	23	Digital_Video_Games	2006
	24	Health & Personal Care	1999
26 Furniture 2000	25	Video Games	1997
	26	Furniture	2000

27	Video	1995
28	Luggage	2002
29	Shoes	1999
30	Home Entertainment	1998
31	Outdoors	1999
32	Apparel	2000
33	Camera	1998
34	Pet Products	1998
35	Lawn and Garden	1999
36	Digital_Video_Download	2000
37	Video DVD	1996
38	Software	1998
39	Kitchen	2000
40	Watches	2001
41	Toys	1997
42	Grocery	1999

What is the breakdown of star ratings (1–5) per product category?

```
In [23]: sql5 = """SELECT product_category, star_rating, COUNT(*) AS count_reviews
    FROM {0}.{1}
    GROUP BY product_category, star_rating
    ORDER BY product_category ASC, star_rating DESC, count_reviews
    """.format(database, table)

result5 = pd.read_sql(sql5, con=engine)
    result5
```

Out[23]:		product_category	star_rating	count_reviews
	0	Apparel	5	3320566
	1	Apparel	4	1147237
	2	Apparel	3	623471
	3	Apparel	2	369601
	4	Apparel	1	445458
	•••			
	210	Wireless	5	4824783
	211	Wireless	4	1501327
	212	Wireless	3	815205
	213	Wireless	2	598330
	214	Wireless	1	1262376

```
In [28]: sql5_2 = """SELECT star_rating, COUNT(*) AS count_reviews,
FROM {0}.{1}
GROUP BY star_rating
ORDER BY star_rating DESC
""".format(database,table)

result5_2 = pd.read_sql(sql5_2, con=engine)
result5_2
```

Out[28]:		star_rating	count_reviews
	0	5	93200812
	1	4	26223470
	2	3	12133927
	3	2	7304430
	4	1	12099639