# PROJECT REPORT PART 1

CHINMAY SUNIL KARANDIKAR
BIG DATA MIS 6346.502
THE UNIVERSITY OF TEXAS AT DALLAS

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# Setting up the project and loading data

```
create database amazon_review;
drop table amazon_review.amazon_reviews_parquet;
CREATE EXTERNAL TABLE amazon_review.amazon_reviews_parquet(
`marketplace` string,
`customer_id` string,
`review_id` string,
`product_id` string,
`product_parent` string,
`product_title` string,
`star_rating` int,
'helpful votes' int,
`total_votes` int,
`vine` string,
`verified_purchase` string,
`review_headline` string,
`review_body` string,
`review_date` DATE,
'year' int)
PARTITIONED BY (
`product_category` string)
--ROW FORMAT DELIMITED
--STORED AS PARQUET
ROW FORMAT SERDE
'org.apache.hadoop.hive.ql.io.parquet.serde.ParquetHiveSerDe'
STORED AS INPUTFORMAT
'org.apache.hadoop.hive.ql.io.parquet.MapredParquetInputFormat'
OUTPUTFORMAT
```

```
'org.apache.hadoop.hive.gl.io.parquet.MapredParquetOutputFormat'
LOCATION
'hdfs:///hive/amazon-reviews-pds/parquet/'
TBLPROPERTIES (
'transient_lastDdlTime'='1583454851');
Msck repair table amazon_review.amazon_reviews_parquet;
create view temporary
as
select * from amazon_review.amazon_reviews_parquet where review_id in (select review_id from
(select customer id, product id, review id, count(*)
from amazon_review.amazon_reviews_parquet
group by customer id, product id, review id
having count(*)=1) as t) and product category in
('Wireless','Automotive','Music','Digital_Music_Purchase','Sports','Toys','Digital_Video_Games','Video_G
ames');
Creating table to filter reviews that are reviewed multiple times by same customer for same product.
```

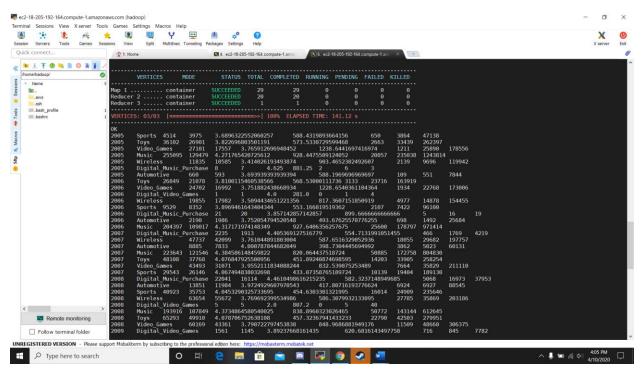
```
create table amazon_review.filtered_reviews
AS
select z.* from(
select *,row_number() over(partition by customer_id,product_id) as row1 from temporary)z where
row1=1;
```

# Question 1

Explore the dataset and provide basic exploratory analysis over time and per product category

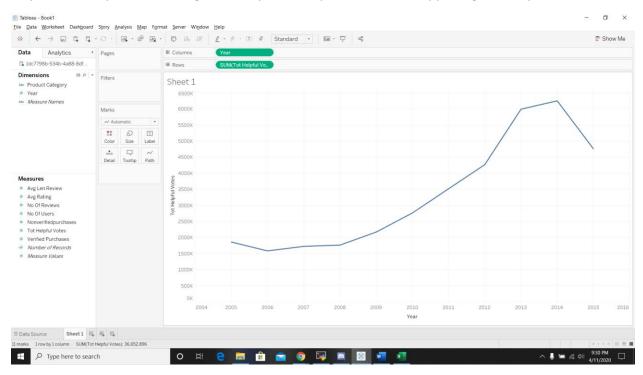
#### Query

Select year, product\_category, count(review\_id) as NoOfReviews, count(Distinct(customer\_id)) as NoOfUsers, avg(star\_rating) as AvgRating ,avg(length(review\_body)) as AvgLenReview, sum(case when verified\_purchase='Y' then 1 else 0 end) as VerifiedPurchases, sum(case when verified\_purchase='N' then 1 else 0 end) as Nonverifiedpurchases, sum(helpful\_votes) as TotHelpfulVotes from amazon\_review.filtered\_reviews where year>=2005 group by year,product\_category order by year;



#### Vizualization

I have taken a line graph for the total helpful votes' vs year. We can clearly see that the number of total helpful votes kept on increasing until the year 2014 put since then dropped significantly.



# Question 2

Provide detailed analysis of Music/Digital\_Music\_Purchase and Digital Video Games/Video Games over time.

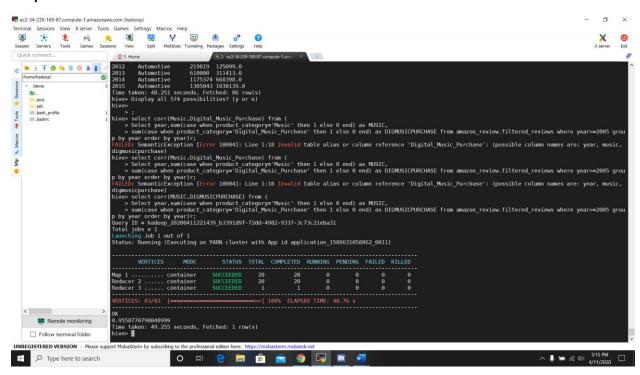
1. Do you see correlation (maybe negative) between the categories over time?

#### Α.

#### Correlation for Music/Digital\_Music\_Purchase

#### Query

select corr(MUSIC,DIGMUSICPURCHASE) from (
Select year,sum(case when product\_category='Music' then 1 else 0 end) as MUSIC,
sum(case when product\_category='Digital\_Music\_Purchase' then 1 else 0 end) as DIGMUSICPURCHASE
from amazon\_review.filtered\_reviews where year>=2005 group by year order by year)r;



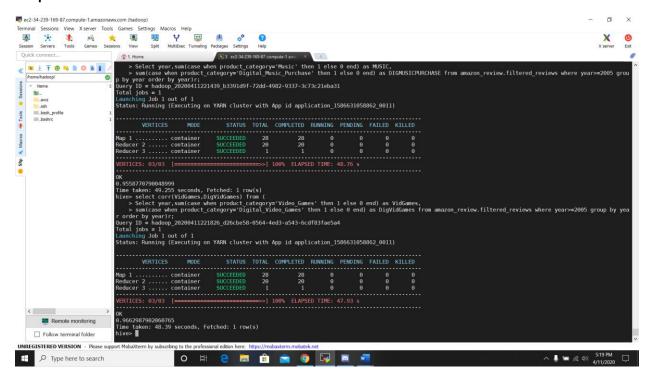
Correlation between Digital\_Video\_Games/Video\_Games based on count of reviews.

#### Query

select corr(VidGames, DigVidGames) from (

Select year,sum(case when product\_category='Video\_Games' then 1 else 0 end) as VidGames, sum(case when product\_category='Digital\_Video\_Games' then 1 else 0 end) as DigVidGames from amazon\_review.filtered\_reviews where year>=2005 group by year order by year)r;

#### Output



#### 2. Are there same users reviewing in both categories?

A. Providing detailed analysis over time for the Music/Digital\_Music\_Purchase category.

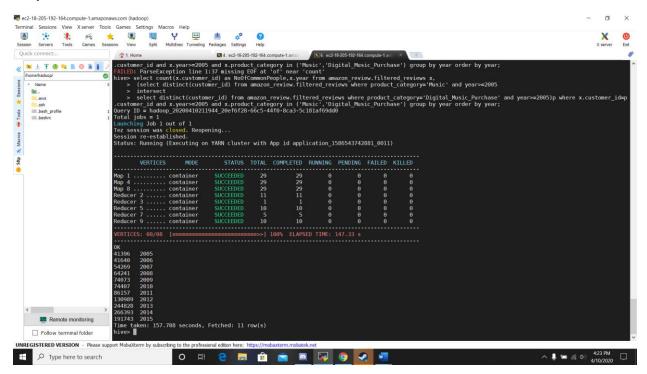
#### Query

select count(x.customer\_id) as NoOfCommonPeople,x.year from amazon\_review.filtered\_reviews x, (select distinct(customer\_id) from amazon\_review.filtered\_reviews where product\_category='Music' and year>=2005

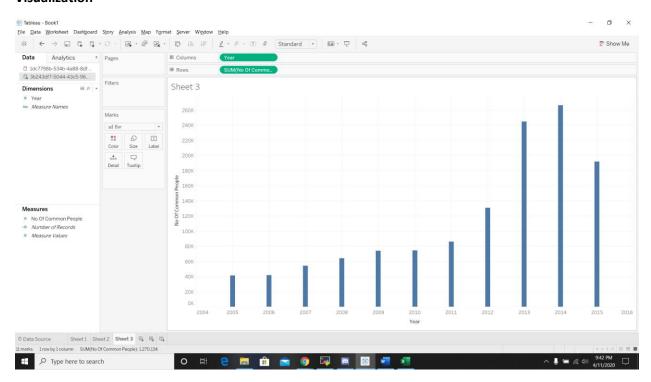
intersect

select distinct(customer\_id) from amazon\_review.filtered\_reviews where product\_category='Digital\_Music\_Purchase' and year>=2005)p where x.customer\_id=p.customer\_id and x.year>=2005 and x.product\_category in ('Music', 'Digital\_Music\_Purchase') group by year order by year;

#### Output



#### Visualization

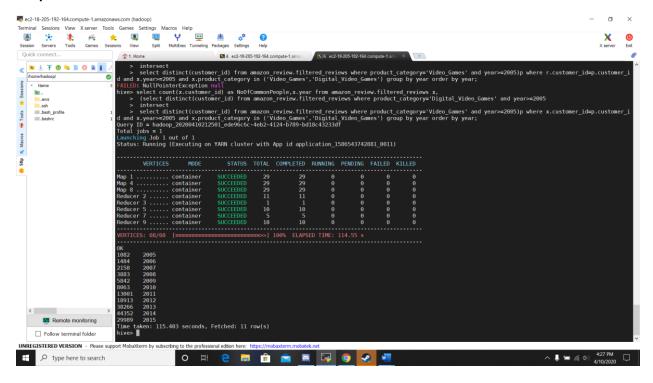


The number of common people dropped after the year 2014 significantly.

Providing detailed analysis over time for the Digital\_Video\_Games/Video\_Games category.

#### Query

select count(x.customer\_id) as NoOfCommonPeople,x.year from amazon\_review.filtered\_reviews x, (select distinct(customer\_id) from amazon\_review.filtered\_reviews where product\_category='Digital\_Video\_Games' and year>=2005 intersect select distinct(customer\_id) from amazon\_review.filtered\_reviews where product\_category='Video\_Games' and year>=2005)p where x.customer\_id=p.customer\_id and x.year>=2005 and x.product\_category in ('Video\_Games','Digital\_Video\_Games') group by year order by year;



#### 3. Can you identify similar items in both categories? Do they get same rating?

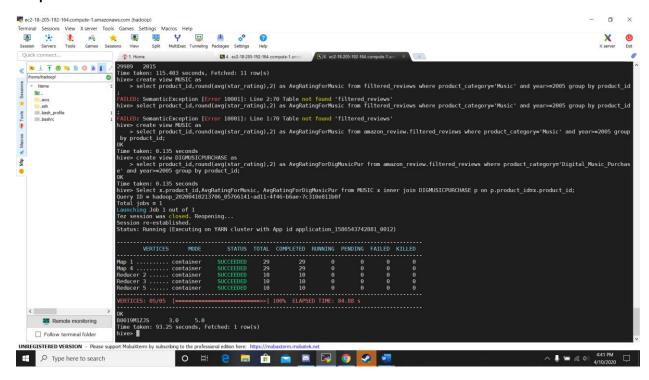
create view MUSIC as

select product\_id,round(avg(star\_rating),2) as AvgRatingForMusic from amazon\_review.filtered\_reviews where product\_category='Music' and year>=2005 group by product\_id;

create view DIGMUSICPURCHASE as select product\_id,round(avg(star\_rating),2) as AvgRatingForDigMusicPur from amazon\_review.filtered\_reviews where product\_category='Digital\_Music\_Purchase' and year>=2005 group by product\_id;

#### Query

Select x.product\_id,AvgRatingForMusic, AvgRatingForDigMusicPur from MUSIC x inner join DIGMUSICPURCHASE p on p.product\_id=x.product\_id;



#### 4. You should cover additional questions and not limit yourself to the above questions

List of customers who have given reviews for products in both Music and Digital\_Music\_Purchase category and their ratings in both categories.

#### Query-

#### create view music as

select customer\_id,product\_category,round(avg(star\_rating),2) as AvgRatingForMusic from amazon\_review.filtered\_reviews where product\_category='Music' and year>=2005 group by customer\_id,product\_category;

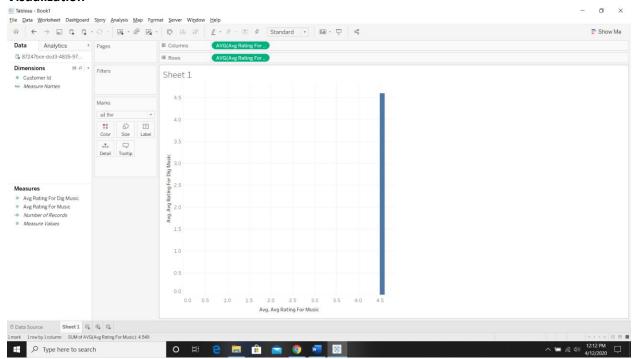
#### create view DigMusic as

select customer\_id,product\_category,round(avg(star\_rating),2) as AvgRatingForDigMusic from amazon\_review.filtered\_reviews where product\_category='Digital\_Music\_Purchase' and year>=2005 group by customer\_id,product\_category;

select r.customer\_id, AvgRatingForMusic, AvgRatingForDigMusic from DigMusic r inner join music u on u.customer\_id=r.customer\_id;

# 

#### Visualization



People are fond of both digital music and music categories equally.

# Question 3

# You should demonstrate your ability to use Hive advanced functions:

1. Window functions: moving average, rank, aggregation functions using relevant ordering and partitioning

Calculating three year Moving average based on number of reviews per product category over time.

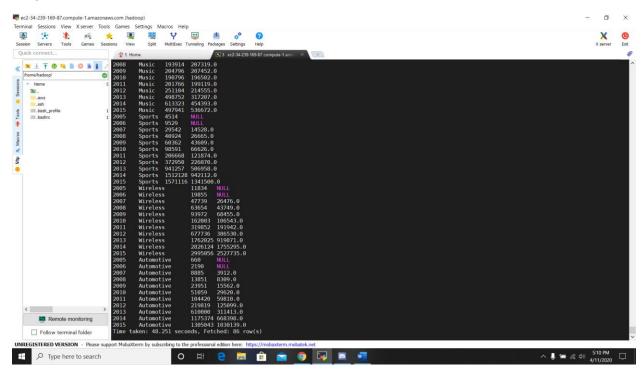
#### Query

select year,product\_category,NoOfReviews,(case when row\_number() over (Partition by product\_category order by year) > 2

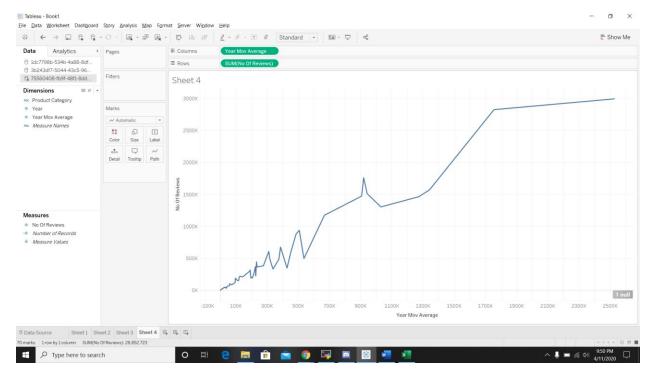
then round(AVG(NoOfReviews) OVER (PARTITION BY product\_category order by year ROWS 2 PRECEDING))

end) as 3YearMovAverage from

(Select year,product\_category,count(review\_id) as NoOfReviews,count(Distinct(customer\_id)) as NumberOfUser,avg(star\_rating) as average\_review\_stars,avg(length(review\_body)) as AvgLenOfReview from amazon\_review.filtered\_reviews group by year,product\_category order by product\_category,year) as x where year>=2005;



#### Visualization



The number of reviews is increasing over the years.

#### Ranking top 10 products in each category based on average length of reviews.

#### Query

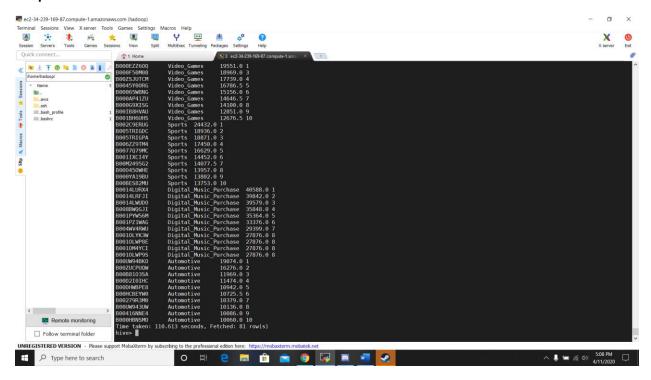
select product\_id,product\_category, AvgLenOfReview,ranking from(

select product\_id,product\_category, AvgLenOfReview,rank() over (Partition by product\_category order by AvgLenOfReview desc) as ranking from

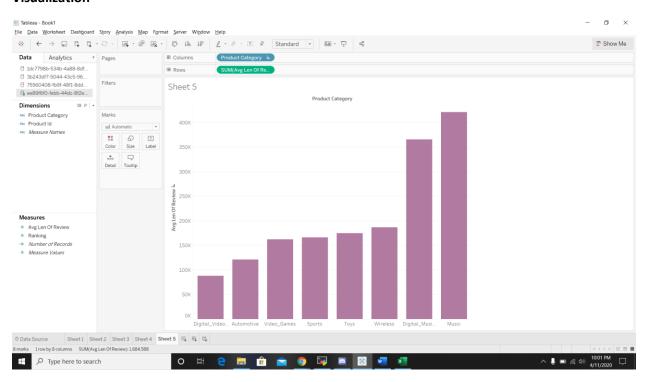
(Select product\_id,product\_category,count(Distinct(customer\_id)) as NoOfUsers,avg(star\_rating) as AvgReviewRating,avg(length(review\_body)) as AvgLenOfReview

from amazon\_review.filtered\_reviews group by product\_category,product\_id)as x)as z where ranking<=10;

#### Output



#### Visualization

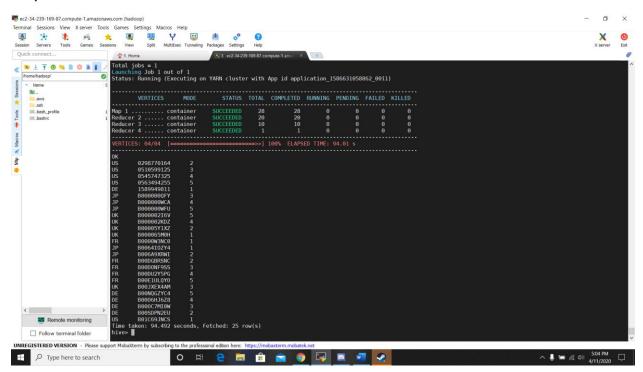


The category music has received the lengthiest reviews.

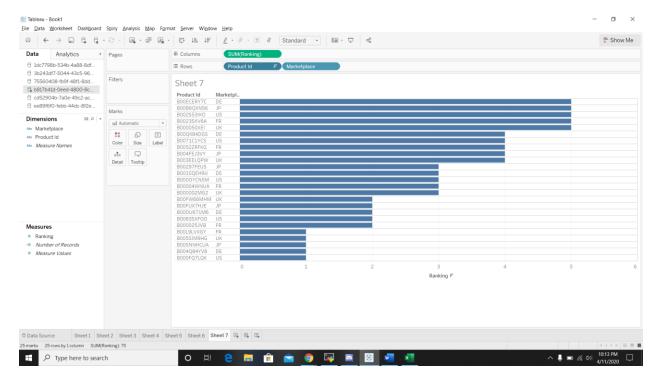
Using aggregate function average to find out top 5 products in each marketplace based on average star rating

#### Query

```
SELECT v.marketplace, v.product id,
    v.Ranking
FROM
  (SELECT z.product id,
    z.marketplace, z. AvgRating,
    Row_number()
    OVER (partition by z.marketplace
    ORDER BY z.AvgRating desc) AS Ranking
  FROM
    (SELECT product_id,
    marketplace,
    avg(star_rating) AS AvgRating
    FROM amazon_review.filtered_reviews
    WHERE year>= 2005
    GROUP BY product_id,marketplace)as z)as v
    WHERE v.Ranking <=5 order by v.product_id;
```



#### Visualization



2. Analytical Aggregate functions: percentile, min, max, average, standard deviation, correlation

#### Percentile

#### Products having highest Percentile of star ratings given by customers:

#### Query

SELECT a.product\_id, a.product\_category,a.Ranking,round(a.Ranking,2) as Percentile from (SELECT b.product\_id,b.product\_category,PERCENT\_RANK()OVER (partition by b.product\_category ORDER BY b.AvgRating desc) AS Ranking

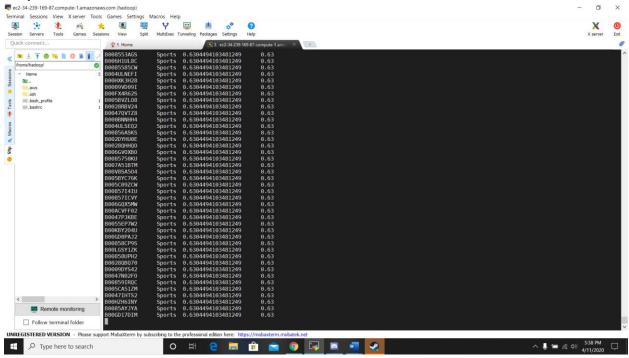
#### FROM

(SELECT product\_id,product\_category,avg(star\_rating) AS AvgRating FROM amazon\_review.filtered\_reviews

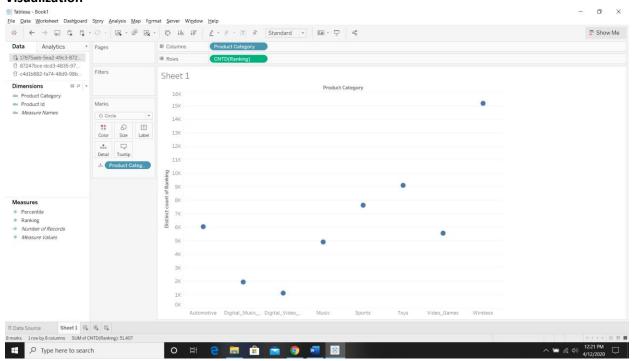
WHERE year>= 2005

GROUP BY product id, product category) as b) as a order by a. Ranking desc;

#### **Output**



#### Visualization



Among all the categories, the product category wireless has the maximum number of reviews and users.

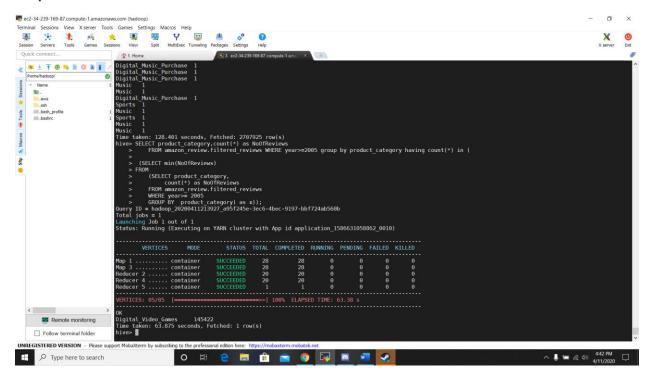
#### Min

#### Query

```
Product category which has got minimum number of reviews.
```

```
SELECT product_category,count(*) as NoOfReviews
FROM amazon_review.filtered_reviews WHERE year>=2005 group by product_category having count(*) in (
```

```
(SELECT min(NoOfReviews)
FROM
  (SELECT product_category,
        count(*) as NoOfReviews
FROM amazon_review.filtered_reviews
WHERE year>= 2005
GROUP BY product category) as x));
```



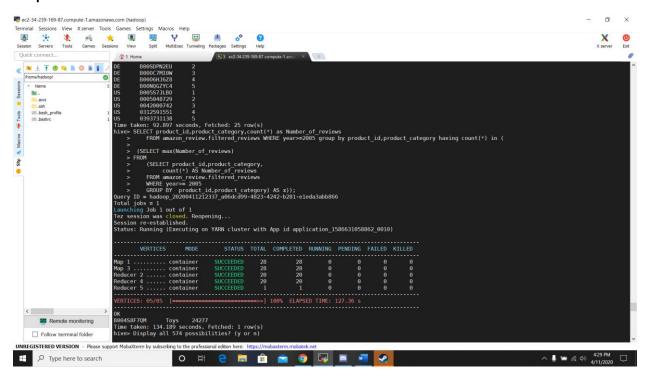
#### Max

Top product which has got maximum number of reviews.

#### Query

```
SELECT product_id,product_category,count(*) as NoOfReviews
   FROM amazon_review.filtered_reviews WHERE year>=2005 group by product_id,product_category
having count(*) in (

(SELECT max(NoOfReviews))
FROM
   (SELECT product_id,product_category,
        count(*) as NoOfReviews
   FROM amazon_review.filtered_reviews
   WHERE year>= 2005
   GROUP BY product_id,product_category) as x));
```



#### Standard deviation

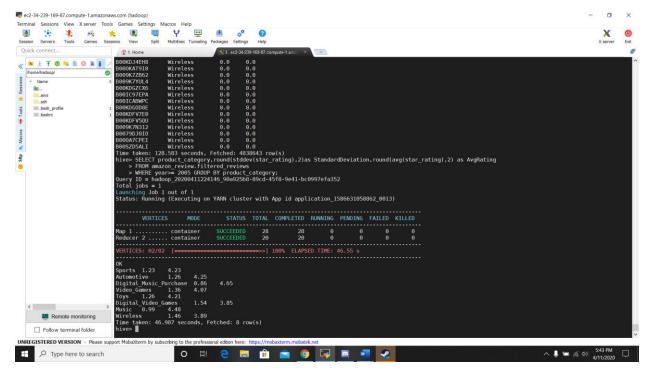
#### Calculating Standard Deviation to analyze normal distribution of star rating of product categories.

SELECT product\_category,round(stddev(star\_rating),2)as StandardDeviation,round(avg(star\_rating),2) as AvgRating

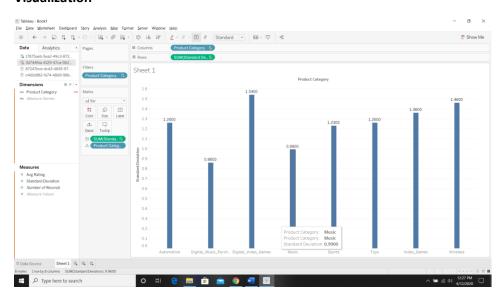
FROM amazon\_review.filtered\_reviews

WHERE year>= 2005 GROUP BY product\_category;

#### **Output**



#### Visualization



The standard deviation for digital video games category is the highest which means the points of data are spread-out from the mean significantly.

# References

https://www.w3schools.com/

https://www.pluralsight.com/courses/aws-athena-get-

started?aid=701j0000001heIoAAI&promo=&oid=7014Q0000022aAOQAY&utm\_source=non\_branded&utm\_medium=digital\_paid\_search\_google&utm\_campaign=US\_Dynamic&utm\_content=&gclid=Cj0KCQiw-

 $\underline{\mathsf{MrOBRDyARIsAKEFbefZsQ7ZU8ISJP85zZBCMAmla0xhAxMLSjTxR4MUagHvCcMCBk9A3ugaApqpEALw\_wc}\underline{\mathsf{B}}$ 

https://www.practicefusion.com/ehr-training/

https://www.kaggle.com/learn/advanced-sql

http://www.sqlcourse2.com/