

▼ Importing Libraries

```
!pip install pyspark
from pyspark.sql import SparkSession
from pyspark.sql.types import *
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/
Collecting pyspark
  Downloading pyspark-3.3.2.tar.gz (281.4 MB)
    ━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━ 281.4/281.4 MB 5.1 MB/s eta 0:00:00
  Preparing metadata (setup.py) ... done
Collecting py4j==0.10.9.5
  Downloading py4j-0.10.9.5-py2.py3-none-any.whl (199 kB)
    ━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━━ 199.7/199.7 kB 13.7 MB/s eta 0:00:00
Building wheels for collected packages: pyspark
  Building wheel for pyspark (setup.py) ... done
  Created wheel for pyspark: filename=pyspark-3.3.2-py2.py3-none-any.whl size=281824
  Stored in directory: /root/.cache/pip/wheels/6c/e3/9b/0525ce8a69478916513509d43693
Successfully built pyspark
Installing collected packages: py4j, pyspark
  Attempting uninstall: py4j
    Found existing installation: py4j 0.10.9.7
    Uninstalling py4j-0.10.9.7:
      Successfully uninstalled py4j-0.10.9.7
  Successfully installed py4j-0.10.9.5 pyspark-3.3.2
```

```
spark = SparkSession.builder.appName('AmazonReviews').getOrCreate()
```

```
import pyspark
from pyspark.context import SparkContext

from pyspark import SparkConf
sc = SparkContext.getOrCreate(SparkConf().setMaster("local[4]"))
```

▼ Connecting The Collab With Drive

```
from google.colab import drive
drive.mount('/content/drive')
```

```
Mounted at /content/drive
```

```
import pandas as pd
import sklearn
```

▼ Loading The Dataset

```
df = spark.read.csv('/content/drive/MyDrive/Amazon_Unlocked_Mobile.csv', header=True, info
```

▼ Displaying 1st 5 Data

```
# display the dataset
df.show(5)
```

```
+-----+-----+-----+-----+-----+
|      Product Name|Brand Name| Price|Rating|      Reviews|Review Votes|
+-----+-----+-----+-----+-----+
|""CLEAR CLEAN ES...| Samsung|199.99| 5|I feel so LUCKY t...| 1|
|""CLEAR CLEAN ES...| Samsung|199.99| 4|nice phone, nice ...| 0|
|""CLEAR CLEAN ES...| Samsung|199.99| 5|      Very pleased| 0|
|""CLEAR CLEAN ES...| Samsung|199.99| 4|It works good but...| 0|
|""CLEAR CLEAN ES...| Samsung|199.99| 4|Great phone to re...| 0|
+-----+-----+-----+-----+-----+
only showing top 5 rows
```

▼ Checking Null Values

```
from pyspark.sql.functions import isnan, when, count, col
```

```
# Count the number of null values in each column
```

```
df.select([count(when(isnan(c) | col(c).isNull(), c)).alias(c) for c in df.columns]).show(
```

```
+-----+-----+-----+-----+-----+
|Product Name|Brand Name|Price|Rating|Reviews|Review Votes|
+-----+-----+-----+-----+-----+
|      0|      64376| 6| 379| 78|      11531|
+-----+-----+-----+-----+-----+
```

we have 0 null values in product name column

we have 64376 null values in brand name column

we have 6 null values in price column

we have 379 null values in rating column

we have 78 null values in reviews column

we have 11531 null values in review votes column

▼ Replacing Null Value

```
# Fill null values with a default value
```

```
df = df.fillna({'Price': 0.0, 'Brand Name': 'Unknown', 'Reviews': 0, 'Rating': 0, 'Review
```

```
# Verify that there are no more null values
```

```
df.select([count(when(isnan(c) | col(c).isNull(), c)).alias(c) for c in df.columns]).show()
```

```
+-----+-----+-----+-----+-----+-----+
|Product Name|Brand Name|Price|Rating|Reviews|Review Votes|
+-----+-----+-----+-----+-----+-----+
|          0|          0|    0|    0|    0|          0|
+-----+-----+-----+-----+-----+-----+
```

we have replaced the null value with default value in each column

as we have 0 null value in product name column we are not going to do anything

we replaced the null values in Price column with 0.0

we replaced the null values in Brand Name column with Unknown

we replaced the null values in Reviews column with 0

we replaced the null value in Rating column with 0

we relaced the null value in Review Votes column with 0

▼ NO OF RECORDS

```
# total value
```

```
df.count()
```

```
413848
```

we have 4,13,848 records

▼ NO OF COLUMNS

```
# total column
```

```
len(df.columns)
```

```
6
```

we have 6 columns

▼ COLUMN NAMES

```
# column name
```

```
df.columns
```

```
['Product Name', 'Brand Name', 'Price', 'Rating', 'Reviews', 'Review Votes']
```

```
# no of record grouped by product name
```

```
df.groupby("Product Name").count().show()
```

```
+-----+-----+
|      Product Name|count|
+-----+-----+
|Apple iPhone 4 A1...|  330|
|Apple iPhone 6s 1...|  163|
|"BLU Studio M HD ...|   47|
|BlueCosmo Iridium...|    1|
|"Cellphones Unloc...|   40|
|CNPGD® All-in-1 S...|  261|
|Flip Phone Unlock...|    2|
|H2O Nano SIM Card...|    1|
|LG H955 Unlocked ...|   44|
|LG LS670 OPTIMUS ...|    1|
|"LG Nexus 5 D820 ...|   62|
|4G-Unlocked Huawe...|    7|
|Apple iPhone 5C Y...|   10|
|Apple iPhone 5s U...|   10|
|Apple iPhone 6 Pl...|  176|
|Apple iPhone SE 6...|   12|
|ASUS ZenFone 2 Un...|   10|
|Blackberry 9530 S...|  484|
|BlackBerry Torch ...|   14|
|BLU Win JR Smartp...|   42|
```

```
+-----+
only showing top 20 rows
```

330 people has bought apple iphone

1 person has bought LG LS670 OPTIMUS

no of record grouped by brand name

```
df.groupby("Brand Name").count().show()
```

```
+-----+
|          Brand Name|count|
+-----+
|          DOOGEE|    97|
|           H2O|     1|
|        Getnord|     5|
|          Kata|    40|
|          P710|    30|
| Android 4.1 - In...|    36|
|          Nokia|16086|
|          4GB"|     1|
|        LandRum|     6|
|        Ulefone|   141|
|        px phone|    84|
|          JIAKE|    32|
|          13MP|   123|
|        Jelly Bean|    42|
|8gb - Internation...|   138|
|        AeroAntenna|     1|
|        Android 4.4 KK| 1390|
|          Doro|    21|
|        Maxwest|    15|
|          htc|   203|
+-----+
only showing top 20 rows
```

16086 people has bought nokia brand

1 person has bought AeroAntenna , H2O brand

phone with the price >= 2000

```
price_lesser_than_1000= df.filter(df['Price'] > 2000).show()
```

```
+-----+-----+-----+-----+-----+-----+
|          Product Name|Brand Name|Price|Rating|          Reviews|Review Votes|
+-----+-----+-----+-----+-----+-----+
|BlueCosmo Iridium...|  Iridium| 2598|    5|These folks are g...|          0|
|"Huawei Ascend P7...|   Huawei| 2066|    1|This phonesoftware...|          0|
```

"Huawei Ascend P7...	Huawei	2066	5	great product	0
"Huawei Ascend P7...	Huawei	2066	5	All very good, ex...	0
"Huawei Ascend P7...	Huawei	2066	5	Super productTwo ...	0
"Huawei Ascend P7...	Huawei	2066	1	item were not as ...	3
"Huawei Ascend P7...	Huawei	2066	5	Great phone, grea...	1
Lenovo S8 S898T 5...	Lenovo	2224	3	good phone. I am ...	0
Lenovo S8 S898T 5...	Lenovo	2224	4	Very good product...	0
Lenovo S8 S898T 5...	Lenovo	2224	1	the phone does no...	0
Lenovo S8 S898T 5...	Lenovo	2224	5	excelente producto	0
Lenovo S8 S898T 5...	Lenovo	2224	1	The worst phone e...	0
Lenovo S8 S898T 5...	Lenovo	2224	5	excelent	0
Lenovo S8 S898T 5...	Lenovo	2224	5	Very good	0
Lenovo S8 S898T 5...	Lenovo	2224	5	Excelentej	0
Lenovo S8 S898T 5...	Lenovo	2224	5	Good	0
Lenovo S8 S898T 5...	Lenovo	2224	5	i like	0
Lenovo S8 S898T 5...	Lenovo	2224	5	lenovo recommend ...	0
Lenovo S8 S898T 5...	Lenovo	2224	5	Excelente	0
Lenovo S8 S898T 5...	Lenovo	2224	2	Do not work	0

+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

maximum every phone has rating 5

so it's best to buy phone at the rate greater than 2000 price

phone with highest and lowest price

from pyspark.sql.functions import min , max

df.select(max('Price'),min('Price')).show()

max(Price)	min(Price)
verykool	1 GHZ Dual Core

verykool mobile phone has the highest price

1 GHZ Dual Core mobile phone has the lowest price

phone with 5 rating

rat_five= df.filter(df['Rating'] == 5).show()

Product Name	Brand Name	Price	Rating	Reviews	Review Votes
""CLEAR CLEAN ES...	Samsung	199.99	5	I feel so LUCKY t...	1
""CLEAR CLEAN ES...	Samsung	199.99	5	Very pleased	0

"""CLEAR CLEAN ES...	Samsung	199.99	5	I originally was ...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	This is a great p...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	These guys are th...	2
"""CLEAR CLEAN ES...	Samsung	199.99	5	Ordered this phon...	1
"""CLEAR CLEAN ES...	Samsung	199.99	5	I was able to get...	6
"""CLEAR CLEAN ES...	Samsung	199.99	5	I brought this ph...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	the phone was gre...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	Phone works great...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	as described, fas...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	Perfect in every ...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	Just got this pho...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	The phone was gre...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	This phone came i...	0
"""CLEAR CLEAN ES...	Samsung	199.99	5	Met all of my exp...	0
((Unlocked))Black...	Unknown	269.1	5	Great. Arrived qu...	0
((Unlocked))Black...	Unknown	269.1	5	Avianna LLC is an...	0
((Unlocked))Black...	Unknown	269.1	5	Exactly what I wa...	1
((Unlocked))Black...	Unknown	269.1	5	Got it faster tha...	0

only showing top 20 rows

samsung product has the highest rating - 5

so it's best to buy samsung product

phone with 4 rating

```
rat_five= df.filter(df['Rating'] == 4).show()
```

Product Name	Brand Name	Price	Rating	Reviews	Review Votes
"""CLEAR CLEAN ES...	Samsung	199.99	4	nice phone, nice ...	0
"""CLEAR CLEAN ES...	Samsung	199.99	4	It works good but...	0
"""CLEAR CLEAN ES...	Samsung	199.99	4	Great phone to re...	0
"""CLEAR CLEAN ES...	Samsung	199.99	4	I love the phone...	1
"""CLEAR CLEAN ES...	Samsung	199.99	4	The battery was o...	0
"""CLEAR CLEAN ES...	Samsung	199.99	4	pros-beautiful sc...	0
"""CLEAR CLEAN ES...	Samsung	199.99	4	Phone good just a...	0
"""CLEAR CLEAN ES...	Samsung	199.99	4	Phone's speaker l...	0
((Unlocked))Black...	Unknown	269.1	4	I liked	0
((Unlocked))Black...	Unknown	269.1	4	Phone works great...	0
"(LANDVO) 5.0"" C...	HTM	69.99	4	All around good p...	0
"(LANDVO) 5.0"" C...	HTM	69.99	4	I have no problem...	0
"(LANDVO) 5.0"" C...	HTM	69.99	4	MUY BUENO	0
"(LANDVO) 5.0"" C...	HTM	69.99	4	Its great for the...	1
"[XMAS DEAL] [New...	Jethro	79.99	4	Nice phone. Easy ...	6
[XMAS DEAL] Jethr...	Jethro	59.99	4	This is a great, ...	9
[XMAS DEAL] Jethr...	Jethro	59.99	4	easy to use. My m...	1
[XMAS DEAL] Jethr...	Jethro	59.99	4	"I bought it for ...	I assume
[XMAS DEAL] Jethr...	Jethro	59.99	4	simple to use. Do...	0
[XMAS DEAL] Jethr...	Jethro	59.99	4	Good phone for my...	1

only showing top 20 rows

phone with 3 rating

```
rat_five= df.filter(df['Rating'] == 3).show()
```

Product Name	Brand Name	Price	Rating	Reviews	Review V
""CLEAR CLEAN ES...	Samsung	199.99	3	It's battery life...	
""CLEAR CLEAN ES...	Samsung	199.99	3	My fiance had thi...	
""CLEAR CLEAN ES...	Samsung	199.99	3	unfortunately Spr...	
""CLEAR CLEAN ES...	Samsung	199.99	3	the reasons for t...	
((Unlocked))Black...	Unknown	269.1	3	Ad advertised as ...	
""(LANDVO) 5.0"" C...	HTM	69.99	3	Valid for Movilne...	
""(LANDVO) 5.0"" C...	HTM	69.99	3	The phone works g...	
""[XMAS DEAL] [New...	Jethro	79.99	3	the charger did n...	
""[XMAS DEAL] [New...	Jethro	79.99	3	No internet	
""[XMAS DEAL] [New...	Jethro	79.99	3	The only reason I...	
""[XMAS DEAL] [New...	Jethro	79.99	3	good phone for my...	
""[XMAS DEAL] [New...	Jethro	79.99	3	Not as sensitive ...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	This may be an is...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	"Word to the wise...	like Sprint. It
[XMAS DEAL] Jethr...	Jethro	59.99	3	Phone number come...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	Bought this for m...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	returned would no...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	My 79 year old mo...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	It worked alright...	
[XMAS DEAL] Jethr...	Jethro	59.99	3	"Word to the wise...	like Sprint. It

only showing top 20 rows

phone with 2 rating

```
rat_five= df.filter(df['Rating'] == 2).show()
```

Product Name	Brand Name	Price	Rating	Reviews	Review Votes
""CLEAR CLEAN ES...	Samsung	199.99	2	The charging port...	0
""CLEAR CLEAN ES...	Samsung	199.99	2	Phone looks good ...	0
""CLEAR CLEAN ES...	Samsung	199.99	2	Had this phone be...	0
""CLEAR CLEAN ES...	Samsung	199.99	2	One of the phones...	0
((Unlocked))Black...	Unknown	269.1	2	when i got phone ...	0
""(LANDVO) 5.0"" C...	HTM	69.99	2	Delivery was fast...	0
""(LANDVO) 5.0"" C...	HTM	69.99	2	When I first got ...	0
""[XMAS DEAL] [New...	Jethro	79.99	2	I like the FM rad...	0
""[XMAS DEAL] [New...	Jethro	79.99	2	does not work wel...	1
""[XMAS DEAL] [New...	Jethro	79.99	2	Sounds like you a...	0
""[XMAS DEAL] [New...	Jethro	79.99	2	the phone is unab...	1
[XMAS DEAL] Jethr...	Jethro	59.99	2	sound quality poo...	0
[XMAS DEAL] Jethr...	Jethro	59.99	2	The speakers are ...	0
[XMAS DEAL] Jethr...	Jethro	59.99	2	My Granny couldn'...	1

[XMAS DEAL] Jethr...	Jethro	59.99	2	Not good sound, n...	0
[XMAS DEAL] Jethr...	Jethro	59.99	2	The SIM card from...	2
[XMAS DEAL] Jethr...	Jethro	59.99	2	I bought this pho...	0
[XMAS DEAL] Jethr...	Jethro	59.99	2	We bought the pho...	0
[XMAS DEAL] Jethr...	Jethro	59.99	2	As arrival of thi...	1
[XMAS DEAL] Jethr...	Jethro	59.99	2	Pros:* The cradle...	0

+-----+-----+-----+-----+-----+-----+

only showing top 20 rows

phone with 1 rating

```
rat_five= df.filter(df['Rating'] == 1).show()
```

Product Name	Brand Name	Price	Rating	Reviews	Review Votes
""CLEAR CLEAN ES...	Samsung	199.99	1	I already had a p...	1
""CLEAR CLEAN ES...	Samsung	199.99	1	I'm really disapp...	1
""CLEAR CLEAN ES...	Samsung	199.99	1	I purchased this ...	19
""CLEAR CLEAN ES...	Samsung	199.99	1	was not in good c...	0
""CLEAR CLEAN ES...	Samsung	199.99	1	Just... not good...	0
"(LANDVO) 5.0"" C...	HTM	69.99	1	Worked OK for aw...	0
"[XMAS DEAL] [New...	Jethro	79.99	1	More complicated ...	2
"[XMAS DEAL] [New...	Jethro	79.99	1	phone reception p...	1
"[XMAS DEAL] [New...	Jethro	79.99	1	I was contacting ...	1
"[XMAS DEAL] [New...	Jethro	79.99	1	Bought this phone...	1
"[XMAS DEAL] [New...	Jethro	79.99	1	I searched for un...	1
"[XMAS DEAL] [New...	Jethro	79.99	1	I am very unhappy...	0
"[XMAS DEAL] [New...	Jethro	79.99	1	Shortly after ret...	0
"[XMAS DEAL] [New...	Jethro	79.99	1	Stopped working a...	0
"[XMAS DEAL] [New...	Jethro	79.99	1	Defective product...	2
[XMAS DEAL] Jethr...	Jethro	59.99	1	I bought this pho...	0
[XMAS DEAL] Jethr...	Jethro	59.99	1	Ordered phone but...	0
[XMAS DEAL] Jethr...	Jethro	59.99	1	Returning product...	0
[XMAS DEAL] Jethr...	Jethro	59.99	1	Locks accidentall...	0
[XMAS DEAL] Jethr...	Jethro	59.99	1	Very bad same wee...	0

+-----+-----+-----+-----+-----+-----+

only showing top 20 rows

no of record grouped by reviews

```
df.groupBy("Reviews").count().show()
```

Reviews	count
It was really bad...	2
I took this phone...	1
On point and fair...	2
After only three ...	1
Good phone..i jus...	1
I bought the phon...	3
Great condition a...	4

```
|the iphone 3gs is...| 3|
|Exactly as I expe...| 2|
|After having a fe...| 1|
|seller was great ...| 1|
|Do NOT get this p...| 1|
|The phone is very...| 1|
|Got to my house e...| 1|
|did not function ...| 3|
|I am happy with m...| 2|
|Having a great ti...| 2|
|My iphone 4 arriv...| 2|
|good items, the c...| 2|
|It's an iPhone, s...| 1|
```

```
+-----+
```

only showing top 20 rows

```
from pyspark.sql.functions import when
from pyspark.sql.functions import lit
```

```
df.withColumn("sentiment", \
  when((df.Rating > 3), lit("positive")) \
    .when((df.Rating < 3), lit("negative")) \
    .otherwise(lit("neutral"))) \
).show()
```

```
+-----+-----+-----+-----+-----+-----+-----+
|      Product Name|Brand Name| Price|Rating|      Reviews|Review Votes|sen
+-----+-----+-----+-----+-----+-----+-----+
|""CLEAR CLEAN ES...| Samsung|199.99| 5|I feel so LUCKY t...| 1| po
|""CLEAR CLEAN ES...| Samsung|199.99| 4|nice phone, nice ...| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 5|      Very pleased| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 4|It works good but...| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 4|Great phone to re...| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 1|I already had a p...| 1| ne
|""CLEAR CLEAN ES...| Samsung|199.99| 2|The charging port...| 0| ne
|""CLEAR CLEAN ES...| Samsung|199.99| 2|Phone looks good ...| 0| ne
|""CLEAR CLEAN ES...| Samsung|199.99| 5|I originally was ...| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 3|It's battery life...| 0| n
|""CLEAR CLEAN ES...| Samsung|199.99| 3|My fiance had thi...| 0| n
|""CLEAR CLEAN ES...| Samsung|199.99| 5|This is a great p...| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 5|These guys are th...| 2| po
|""CLEAR CLEAN ES...| Samsung|199.99| 1|I'm really disapp...| 1| ne
|""CLEAR CLEAN ES...| Samsung|199.99| 5|Ordered this phon...| 1| po
|""CLEAR CLEAN ES...| Samsung|199.99| 2|Had this phone be...| 0| ne
|""CLEAR CLEAN ES...| Samsung|199.99| 5|I was able to get...| 6| po
|""CLEAR CLEAN ES...| Samsung|199.99| 5|I brought this ph...| 0| po
|""CLEAR CLEAN ES...| Samsung|199.99| 4|I love the phone....| 1| po
|""CLEAR CLEAN ES...| Samsung|199.99| 3|unfortunately Spr...| 0| n
```

```
+-----+
```

only showing top 20 rows

```
df.filter(df['Reviews'] == "great phone").show()
```

Product Name	Brand Name	Price	Rating	Reviews	Review Votes
Apple iPhone 4 - ...	Unknown	NA	5	great phone	0
Apple iPhone 4 16...	Apple	208.79	5	great phone	0
Apple iPhone 4 16...	Apple	208.79	5	great phone	0
Apple iPhone 4 32...	Apple	99.99	5	great phone	0
Apple iPhone 4S 3...	Apple	209.48	5	great phone	0
Apple iPhone 4S 6...	Apple	114	5	great phone	0
Apple iPhone 4S 6...	Apple	114	5	great phone	0
Apple iPhone 4s a...	Unknown	159.99	5	great phone	0
Apple iPhone 5 Un...	Apple	265	5	great phone	0
Apple iPhone 5 Un...	Apple	309	5	great phone	0
Apple iPhone 5 Un...	Apple	309	5	great phone	0
Apple iPhone 5 Un...	Apple	314.95	5	great phone	0
Apple iPhone 5C 1...	Unknown	149.99	5	great phone	0
Apple iPhone 5s 1...	Apple	149.99	5	great phone	0
Apple iPhone 5s 3...	Apple	125	5	great phone	0
Apple iPhone 5s 3...	Apple	209	5	great phone	0
Apple iPhone 5s 3...	Apple	209	5	great phone	0
Apple iPhone 5s 3...	Apple	49	5	great phone	0
Apple iPhone 5s 6...	Apple	239.95	5	great phone	0
Apple iPhone 5s F...	Apple	272.99	5	great phone	0

only showing top 20 rows

```
df.filter(df['Reviews'] == "not good").show()
```

Product Name	Brand Name	Price	Rating	Reviews	Review Votes
4 Inch Touch Scre...	Unknown	23.9	1	not good	0
4 Inch Touch Scre...	Unknown	23.9	1	not good	0
Apple iPhone 5 Un...	Apple	309	1	not good	0
Apple iPhone 5 Un...	Apple	314.95	1	not good	0
Apple iPhone 5 Un...	Apple	314.95	1	not good	0
Apple iPhone 6 Pl...	Unknown	699.95	2	not good	1
Apple iPhone 6 Pl...	Apple	615	2	not good	1
Apple iPhone 6 Pl...	Apple	605	2	not good	1
Apple iPhone 6 Pl...	Apple	519	2	not good	1
Apple iPhone 6 Pl...	Unknown	490	2	not good	1
BLU Studio 5.0 HD...	BLU	107.98	1	not good	0
BLU Studio 5.0 HD...	BLU	119.99	1	not good	0
LG Neon GT365 Pre...	LG	69.99	1	not good	0
Motorola Droid RA...	Motorola	68.34	1	not good	0
Nokia 6350 Gray A...	Nokia	269.1	1	not good	0
Nokia C2-01.5 Unl...	Nokia	98	1	not good	0
"Samsung Galaxy G...	Samsung	434.99	2	not good	0
Samsung Galaxy S6...	Samsung	529	2	not good	0
Samsung Galaxy S6...	Samsung	529	2	not good	0
Samsung Galaxy S6...	Samsung	449	2	not good	0

only showing top 20 rows

```
from pyspark.ml.feature import Tokenizer, StopWordsRemover, HashingTF, IDF, StringIndexer
from pyspark.ml.classification import NaiveBayes, LogisticRegression, RandomForestClassifi
from pyspark.ml.evaluation import MulticlassClassificationEvaluator
from pyspark.ml.pipeline import Pipeline
from pyspark.mllib.evaluation import MulticlassMetrics
```

```
#Naive Bayes classifier
tokenizer = Tokenizer(inputCol='Review Votes', outputCol='words')
remover = StopWordsRemover(inputCol=tokenizer.getOutputCol(), outputCol='filtered')
hashingTF = HashingTF(inputCol=remover.getOutputCol(), outputCol='rawFeatures', numFeature
idf = IDF(inputCol=hashingTF.getOutputCol(), outputCol='features')
labelIndexer = StringIndexer(inputCol='Rating', outputCol='label', handleInvalid='keep')

nb = NaiveBayes()

pipeline = Pipeline(stages=[tokenizer, remover, hashingTF, idf, labelIndexer, nb])

(trainingData, testData) = df.randomSplit([0.7, 0.3], seed=123)

model = pipeline.fit(trainingData)

predictions = model.transform(testData)

evaluator = MulticlassClassificationEvaluator(labelCol='label', predictionCol='prediction'
accuracy = evaluator.evaluate(predictions)

print(f'Accuracy: {accuracy}')
```

Accuracy: 0.5200615047738653

```
# Calculate evaluation metrics
metrics = MulticlassMetrics(predictions.select('prediction', 'label').rdd)
prec = metrics.weightedPrecision
rec = metrics.weightedRecall
f1 = metrics.weightedFMeasure()

# Print evaluation metrics
print(f'Precision: {prec}')
print(f'Recall: {rec}')
print(f'F1 Score: {f1}')
```

Precision: 0.4379434522510633
Recall: 0.5200615047738654
F1 Score: 0.39322363287054907

```
#Logistic regression classifier
lr = LogisticRegression(maxIter=10, regParam=0.01)

pipeline = Pipeline(stages=[tokenizer, remover, hashingTF, idf, labelIndexer, lr])

(trainingData, testData) = df.randomSplit([0.7, 0.3], seed=123)

model = pipeline.fit(trainingData)

predictions = model.transform(testData)

evaluator = MulticlassClassificationEvaluator(labelCol='label', predictionCol='prediction'
accuracy = evaluator.evaluate(predictions)

print(f'Accuracy: {accuracy}')
```

Calculate evaluation metrics

```
metrics = MulticlassMetrics(predictions.select('prediction', 'label').rdd)
prec = metrics.weightedPrecision
rec = metrics.weightedRecall
f1 = metrics.weightedFMeasure()
```

Print evaluation metrics

```
print(f'Precision: {prec}')
print(f'Recall: {rec}')
print(f'F1 Score: {f1}')
```

Accuracy: 0.5396560884895909
Precision: 0.48746599312436023
Recall: 0.5396560884895908
F1 Score: 0.3901890112525952

#Random forest classifier

```
rf = RandomForestClassifier(numTrees=10)

pipeline = Pipeline(stages=[tokenizer, remover, hashingTF, idf, labelIndexer, rf])

(trainingData, testData) = df.randomSplit([0.7, 0.3], seed=123)

model = pipeline.fit(trainingData)

predictions = model.transform(testData)

evaluator = MulticlassClassificationEvaluator(labelCol='label', predictionCol='prediction'
accuracy = evaluator.evaluate(predictions)

print(f'Random forest classifier Accuracy: {accuracy}')
```

Calculate evaluation metrics

```
metrics = MulticlassMetrics(predictions.select('prediction', 'label').rdd)
prec = metrics.weightedPrecision
rec = metrics.weightedRecall
f1 = metrics.weightedFMeasure()
```

```
# Print evaluation metrics
print(f'Precision: {prec}')
print(f'Recall: {rec}')
print(f'F1 Score: {f1}')
```



Random forest classifier Accuracy: 0.5223961100645639
Precision: 0.4173603011272721
Recall: 0.5223961100645639
F1 Score: 0.3586813433651473

```
#Decision Tree classifier
from pyspark.ml.classification import DecisionTreeClassifier

dt = DecisionTreeClassifier(labelCol='label', featuresCol='features')

pipeline = Pipeline(stages=[tokenizer, remover, hashingTF, idf, labelIndexer, dt])

(trainingData, testData) = df.randomSplit([0.7, 0.3], seed=123)

model = pipeline.fit(trainingData)

predictions = model.transform(testData)

evaluator = MulticlassClassificationEvaluator(labelCol='label', predictionCol='prediction')
accuracy = evaluator.evaluate(predictions)

print(f'Accuracy: {accuracy}')
```

```
# Calculate evaluation metrics
metrics = MulticlassMetrics(predictions.select('prediction', 'label').rdd)
prec = metrics.weightedPrecision
rec = metrics.weightedRecall
f1 = metrics.weightedFMeasure()

# Print evaluation metrics
print(f'Precision: {prec}')
print(f'Recall: {rec}')
print(f'F1 Score: {f1}')
```

Accuracy: 0.5290698610507334
Precision: 0.3314578038398621
Recall: 0.5290698610507334
F1 Score: 0.3674930330395204

```
# K-means clustering
from pyspark.ml.clustering import KMeans

kmeans = KMeans().setK(2).setSeed(1)
pipeline = Pipeline(stages=[tokenizer, remover, hashingTF, idf, labelIndexer, kmeans])

(trainingData, testData) = df.randomSplit([0.7, 0.3], seed=123)

model = pipeline.fit(trainingData)

predictions = model.transform(testData).withColumn("prediction", col("prediction").cast("c"))

evaluator = MulticlassClassificationEvaluator(labelCol='label', predictionCol='prediction')
accuracy = evaluator.evaluate(predictions)

print(f'Accuracy: {accuracy}')
```



```
# Calculate evaluation metrics
metrics = MulticlassMetrics(predictions.select('prediction', 'label').rdd)
prec = metrics.weightedPrecision
rec = metrics.weightedRecall
f1 = metrics.weightedFMeasure()
```



```
# Print evaluation metrics
print(f'Precision: {prec}')
```

```
print(f'Recall: {rec}')
```

```
print(f'F1 Score: {f1}')
```



```
Accuracy: 0.5218889371910673
Precision: 0.27237244815661843
Recall: 0.5218889371910673
F1 Score: 0.35793800406490034
```

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