

10	Duck Hunt	NES	1984	Shooter	Nintendo	26.93	0.63	
11	Nintendogs	DS	2005	Simulation	Nintendo	9.07	11	
12	Mario Kart DS	DS	2005	Racing	Nintendo	9.81	7.57	
13	Pokemon Gold/Poke...	GB	1999	Role-Playing	Nintendo	9	6.18	
14	Wii Fit	Wii	2007	Sports	Nintendo	8.94	8.03	
15	Wii Fit Plus	Wii	2009	Sports	Nintendo	9.09	8.59	
16	Kinect Adventures!	X360	2010	Misc	Microsoft Game St...	14.97	4.94	
17	Grand Theft Auto V	PS3	2013	Action	Take-Two Interactive	7.01	9.27	
18	Grand Theft Auto:...	PS2	2004	Action	Take-Two Interactive	9.43	0.4	
19	Super Mario World	SNES	1990	Platform	Nintendo	12.78	3.75	
20	Brain Age: Train ...	DS	2005	Misc	Nintendo	4.75	9.26	

only showing top 20 rows

```
# Checking the schema
```

```
# By default it will read all column as string , so we use inferSchema = true to handle this case
```

```
df_pyspark.printSchema()
```

```
root
|-- Rank: integer (nullable = true)
|-- Name: string (nullable = true)
|-- Platform: string (nullable = true)
|-- Year: string (nullable = true)
|-- Genre: string (nullable = true)
|-- Publisher: string (nullable = true)
|-- NA_Sales: double (nullable = true)
|-- EU_Sales: double (nullable = true)
|-- JP_Sales: double (nullable = true)
|-- Other_Sales: double (nullable = true)
```

```
# Another way to read the csv file or dataset
```

```
df_pyspark = spark.read.csv('vgsales2.csv', header = True , inferSchema = True)
```

```
df_pyspark.show()
```

Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales
Wii Sports	Wii	2006	Sports	Nintendo	41.49	29.02	3.77	8.46
ario Bros.	NES	1985	Platform	Nintendo	29.08	3.58	6.81	0.77
o Kart Wii	Wii	2008	Racing	Nintendo	15.85	12.88	3.79	3.31
rts Resort	Wii	2009	Sports	Nintendo	15.75	11.01	3.28	2.96
d/Pokem...	GB	1996	Role-Playing	Nintendo	11.27	8.89	10.22	1.0
Tetris	GB	1989	Puzzle	Nintendo	23.2	2.26	4.22	0.58
Mario B...	DS	2006	Platform	Nintendo	11.38	9.23	6.5	2.9
Wii Play	Wii	2006	Misc	Nintendo	14.03	9.2	2.93	2.85
Mario B...	Wii	2009	Platform	Nintendo	14.59	7.06	4.7	2.26
Duck Hunt	NES	1984	Shooter	Nintendo	26.93	0.63	0.28	0.47
Nintendogs	DS	2005	Simulation	Nintendo	9.07	11.0	1.93	2.75
io Kart DS	DS	2005	Racing	Nintendo	9.81	7.57	4.13	1.92
ld/Poke...	GB	1999	Role-Playing	Nintendo	9.0	6.18	7.2	0.71
Wii Fit	Wii	2007	Sports	Nintendo	8.94	8.03	3.6	2.15
i Fit Plus	Wii	2009	Sports	Nintendo	9.09	8.59	2.53	1.79
dventures!	X360	2010	Misc	Microsoft Game St...	14.97	4.94	0.24	1.67
eft Auto V	PS3	2013	Action	Take-Two Interactive	7.01	9.27	0.97	4.14
t Auto:...	PS2	2004	Action	Take-Two Interactive	9.43	0.4	0.41	10.57
ario World	SNES	1990	Platform	Nintendo	12.78	3.75	3.54	0.55
Train ...	DS	2005	Misc	Nintendo	4.75	9.26	4.16	2.05

20 rows

```
# checking the schema
```

```
df_pyspark.printSchema()
```

```
root
|-- Rank: integer (nullable = true)
|-- Name: string (nullable = true)
|-- Platform: string (nullable = true)
|-- Year: string (nullable = true)
```

```
|-- Genre: string (nullable = true)
|-- Publisher: string (nullable = true)
|-- NA_Sales: double (nullable = true)
|-- EU_Sales: double (nullable = true)
|-- JP_Sales: double (nullable = true)
|-- Other_Sales: double (nullable = true)
```

```
type( df_pyspark)
```

```
# it will give dataframe as output which is a type of datastructure
```

```
pyspark.sql.dataframe.DataFrame
```

```
def __init__(jdf: JavaObject, sql_ctx: Union['SQLContext',
'SparkSession'])
```

```
    Supports Spark connect.
```

```
Examples
```

```
-----
A :class:`DataFrame` is equivalent to a relational table in Spark SQL,
and can be created using various functions in :class:`SparkSession`:
```

```
# see the columns
```

```
df_pyspark.columns
```

```
['Rank',
 'Name',
 'Platform',
 'Year',
 'Genre',
 'Publisher',
 'NA_Sales',
 'EU_Sales',
 'JP_Sales',
 'Other_Sales']
```

```
# see the head element of the dataset as a form of list
```

```
df_pyspark.head(3 )
```

```
[Row(Rank=1, Name='Wii Sports', Platform='Wii', Year='2006', Genre='Sports',
Publisher='Nintendo', NA_Sales=41.49, EU_Sales=29.02, JP_Sales=3.77, Other_Sales=8.46),
 Row(Rank=2, Name='Super Mario Bros.', Platform='NES', Year='1985', Genre='Platform',
Publisher='Nintendo', NA_Sales=29.08, EU_Sales=3.58, JP_Sales=6.81, Other_Sales=0.77),
 Row(Rank=3, Name='Mario Kart Wii', Platform='Wii', Year='2008', Genre='Racing',
Publisher='Nintendo', NA_Sales=15.85, EU_Sales=12.88, JP_Sales=3.79, Other_Sales=3.31)]
```

```
# Selecting a column
```

```
df_pyspark.select( 'Name')
```

```
# this will return the type of dataframe
```

```
DataFrame[Name: string]
```

```
df_pyspark.select( 'Name').show()
```

```
# this will return the entire data present in that column
```

```
+-----+
|           Name|
+-----+
|    Wii Sports|
| Super Mario Bros.|
|   Mario Kart Wii|
| Wii Sports Resort|
| Pokemon Red/Pokem...|
|           Tetris|
| New Super Mario B...|
|           Wii Play|
| New Super Mario B...|
|           Duck Hunt|
|           Nintendogs|
```

```

|      Mario Kart DS|
|Pokemon Gold/Poke...|
|      Wii Fit|
|      Wii Fit Plus|
|Kinect Adventures!|
|Grand Theft Auto V|
|Grand Theft Auto:...|
|      Super Mario World|
|Brain Age: Train ...|
+-----+
only showing top 20 rows

```

Now selecting more than 1 columns

```
df_pyspark.select( ['Name' , 'publisher'])
```

```
DataFrame[Name: string, publisher: string]
```

```
df_pyspark.select( ['Name' , 'publisher']).show()
```

```

+-----+-----+
|      Name|publisher|
+-----+-----+
|      Wii Sports|Nintendo|
|Super Mario Bros.|Nintendo|
|Mario Kart Wii|Nintendo|
|Wii Sports Resort|Nintendo|
|Pokemon Red/Pokem...|Nintendo|
|Tetris|Nintendo|
|New Super Mario B...|Nintendo|
|Wii Play|Nintendo|
|New Super Mario B...|Nintendo|
|Duck Hunt|Nintendo|
|Nintendogs|Nintendo|
|Mario Kart DS|Nintendo|
|Pokemon Gold/Poke...|Nintendo|
|Wii Fit|Nintendo|
|Wii Fit Plus|Nintendo|
|Kinect Adventures!|Microsoft Game St...|
|Grand Theft Auto V|Take-Two Interactive|
|Grand Theft Auto:...|Take-Two Interactive|
|Super Mario World|Nintendo|
|Brain Age: Train ...|Nintendo|
+-----+-----+
only showing top 20 rows

```

```
df_pyspark['Name']
```

the return type of this is column

```
Column<'Name'>
```

Now we will see how to check the datatypes

```
df_pyspark.dtypes
```

```

[('Rank', 'int'),
 ('Name', 'string'),
 ('Platform', 'string'),
 ('Year', 'string'),
 ('Genre', 'string'),
 ('Publisher', 'string'),
 ('NA_Sales', 'double'),
 ('EU_Sales', 'double'),
 ('JP_Sales', 'double'),
 ('Other_Sales', 'double')]

```

Now checking the describe option similar to pandas

```
df_pyspark.describe()
```

```
DataFrame[summary: string, Rank: string, Name: string, Platform: string, Year: string, Genre: string, Publisher: string, NA_Sales: string, EU_Sales: string, JP_Sales: string, Other_Sales: string]
```

```
df_pyspark.describe().show()
# NULL values are shown because of mean and max and st dev of string will be NULL
```

summary	Rank	Name	Platform	Year	Genre	Publisher
count	16598	16598	16598	16598	16598	16598
mean	8300.605253645017	1942.0	2600.0	2006.4064433147546	NULL	NULL
stddev	4791.8539328964	NULL	0.0	5.828981114713253	NULL	NULL
min	1	.hack//G.U. Vol.1...	2600	1980	Action	10TACLE Studio
max	16600	iShin Chan Flipa ...	XOne	N/A	Strategy	responsDES...

```
# Adding column in dataframe
df_pyspark = df_pyspark.withColumn('Global Sales',df_pyspark['NA_Sales'] + df_pyspark['EU_Sales'] + d
# df_pyspark = df_pyspark.withColumn('Global Sales', col('NA_Sales') + col('EU_Sales') + col('JP_Sale
df_pyspark = df_pyspark.withColumn('Global + 10 ',df_pyspark['Global Sales']+ 10)
```

```
df_pyspark.show()
```

	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global Sales	Global + 10
	Nintendo	41.49	29.02	3.77	8.46	82.74000000000001	92.74000000000001
	Nintendo	29.08	3.58	6.81	0.77	40.24	50.24
	Nintendo	15.85	12.88	3.79	3.31	35.830000000000005	45.830000000000005
	Nintendo	15.75	11.01	3.28	2.96	33.0	43.0
	Nintendo	11.27	8.89	10.22	1.0	31.380000000000003	41.38
	Nintendo	23.2	2.26	4.22	0.58	30.259999999999998	40.26
	Nintendo	11.38	9.23	6.5	2.9	30.009999999999998	40.01
	Nintendo	14.03	9.2	2.93	2.85	29.009999999999998	39.01
	Nintendo	14.59	7.06	4.7	2.26	28.61	38.61
	Nintendo	26.93	0.63	0.28	0.47	28.31	38.31
	Nintendo	9.07	11.0	1.93	2.75	24.75	34.75
	Nintendo	9.81	7.57	4.13	1.92	23.43	33.43
	Nintendo	9.0	6.18	7.2	0.71	23.09	33.09
	Nintendo	8.94	8.03	3.6	2.15	22.72	32.72
	Nintendo	9.09	8.59	2.53	1.79	22.0	32.0
Microsoft Game St...		14.97	4.94	0.24	1.67	21.82	31.82
Take-Two Interactive		7.01	9.27	0.97	4.14	21.39	31.39
Take-Two Interactive		9.43	0.4	0.41	10.57	20.810000000000002	30.810000000000002
	Nintendo	12.78	3.75	3.54	0.55	20.62	30.62
	Nintendo	4.75	9.26	4.16	2.05	20.220000000000002	30.220000000000002

```
# Drop the columns
df_pyspark = df_pyspark.drop('Global + 10 ')
df_pyspark.show()
```

Rank	Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sa
1	Wii Sports	Wii	2006	Sports	Nintendo	41.49	29.02	:
2	Super Mario Bros.	NES	1985	Platform	Nintendo	29.08	3.58	(
3	Mario Kart Wii	Wii	2008	Racing	Nintendo	15.85	12.88	:
4	Wii Sports Resort	Wii	2009	Sports	Nintendo	15.75	11.01	:
5	Pokemon Red/Pokem...	GB	1996	Role-Playing	Nintendo	11.27	8.89	10
6	Tetris	GB	1989	Puzzle	Nintendo	23.2	2.26	,
7	New Super Mario B...	DS	2006	Platform	Nintendo	11.38	9.23	:
8	Wii Play	Wii	2006	Misc	Nintendo	14.03	9.2	:
9	New Super Mario B...	Wii	2009	Platform	Nintendo	14.59	7.06	:
10	Duck Hunt	NES	1984	Shooter	Nintendo	26.93	0.63	(
11	Nintendogs	DS	2005	Simulation	Nintendo	9.07	11.0	:

12	Mario Kart DS	DS	2005	Racing	Nintendo	9.81	7.57	4
13	Pokemon Gold/Poke...	GB	1999	Role-Playing	Nintendo	9.0	6.18	4
14	Wii Fit	Wii	2007	Sports	Nintendo	8.94	8.03	4
15	Wii Fit Plus	Wii	2009	Sports	Nintendo	9.09	8.59	4
16	Kinect Adventures!	X360	2010	Misc	Microsoft Game St...	14.97	4.94	4
17	Grand Theft Auto V	PS3	2013	Action	Take-Two Interactive	7.01	9.27	4
18	Grand Theft Auto:...	PS2	2004	Action	Take-Two Interactive	9.43	0.4	4
19	Super Mario World	SNES	1990	Platform	Nintendo	12.78	3.75	4
20	Brain Age: Train ...	DS	2005	Misc	Nintendo	4.75	9.26	4

only showing top 20 rows

```
# Rename the column
df_pyspark.withColumnRenamed('Name', 'New Name').show()
```

Rank	New Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_S
1	Wii Sports	Wii	2006	Sports	Nintendo	41.49	29.02	4
2	Super Mario Bros.	NES	1985	Platform	Nintendo	29.08	3.58	4
3	Mario Kart Wii	Wii	2008	Racing	Nintendo	15.85	12.88	4
4	Wii Sports Resort	Wii	2009	Sports	Nintendo	15.75	11.01	4
5	Pokemon Red/Pokem...	GB	1996	Role-Playing	Nintendo	11.27	8.89	10
6	Tetris	GB	1989	Puzzle	Nintendo	23.2	2.26	4
7	New Super Mario B...	DS	2006	Platform	Nintendo	11.38	9.23	4
8	Wii Play	Wii	2006	Misc	Nintendo	14.03	9.2	4
9	New Super Mario B...	Wii	2009	Platform	Nintendo	14.59	7.06	4
10	Duck Hunt	NES	1984	Shooter	Nintendo	26.93	0.63	4
11	Nintendogs	DS	2005	Simulation	Nintendo	9.07	11.0	4
12	Mario Kart DS	DS	2005	Racing	Nintendo	9.81	7.57	4
13	Pokemon Gold/Poke...	GB	1999	Role-Playing	Nintendo	9.0	6.18	4
14	Wii Fit	Wii	2007	Sports	Nintendo	8.94	8.03	4
15	Wii Fit Plus	Wii	2009	Sports	Nintendo	9.09	8.59	4
16	Kinect Adventures!	X360	2010	Misc	Microsoft Game St...	14.97	4.94	4
17	Grand Theft Auto V	PS3	2013	Action	Take-Two Interactive	7.01	9.27	4
18	Grand Theft Auto:...	PS2	2004	Action	Take-Two Interactive	9.43	0.4	4
19	Super Mario World	SNES	1990	Platform	Nintendo	12.78	3.75	4
20	Brain Age: Train ...	DS	2005	Misc	Nintendo	4.75	9.26	4

only showing top 20 rows

```
from pyspark.sql import SparkSession

# Initialize SparkSession
spark = SparkSession.builder.appName("Download DataFrame as CSV").getOrCreate()

# Assuming df_pyspark is your PySpark DataFrame

# Convert PySpark DataFrame to Pandas DataFrame
df_pandas = df_pyspark.toPandas()

# Save Pandas DataFrame as CSV file
df_pandas.to_csv("output.csv", index=False)

# Download the CSV file in Colab
from google.colab import files
files.download("output.csv")
```

✓ Handling Missing values

- Drop Columns
- Dropping Rows
- Various parameters in Dropping functionalities

- Handling missing values by mean, median and mode

```
from pyspark.sql import SparkSession
```

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

```
df_pyspark = spark.read.csv('test2.csv', header = True , inferSchema = True )
```

```
df_pyspark.show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```
# drop the columns
```

```
df_pyspark.drop('Name').show()
```

age	Experience	Salary
31	10	30000
30	8	25000
29	4	20000
24	3	20000
21	1	15000
23	2	18000
NULL	NULL	40000
34	10	38000
36	NULL	NULL

```
df_pyspark.show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```
# Deleting the row values
```

```
df_pyspark.na.drop().show()
```

```
# By using this without any parameter, null rows will be deleted containing it
```

```
# In this case last 3 rows will be deleted
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000

```
# if any = how
df_pyspark.na.drop( how = 'all').show()
# In this case the rows which are having all the null values in it will be dropped
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```
# if how = any
df_pyspark.na.drop( how = 'any').show()
# Drop the rows having any null value in it
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000

```
#Using Threshold value
df_pyspark.na.drop( how = 'any' , thresh = 2).show()
# Drop the rows in which atleast 2 non null values are present for a row
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000

```
# Subset
df_pyspark.na.drop( how = 'all', subset = ['Experience']).show()
# In this case the rows in which null values are present in Experience column got deleted
```


Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
NULL	34	10	38000

Filling the missing values

```
df_pyspark.na.fill('Missing Values', 'age').show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```
df_pyspark.na.fill('Missing Values').show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
Missing Values	34	10	38000
Missing Values	36	NULL	NULL

```
# df_pyspark.withColumn("Name", when(col("Name") == "NULL", None).otherwise(col("Text")))
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-19-5e158572894d> in <cell line: 1>()
----> 1 df_pyspark.withColumn("Name", when(col("Name") == "NULL",
None).otherwise(col("Text")))
```

NameError: name 'when' is not defined

Now Handling the NULL values with the Mean

```
df_pyspark.show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000

Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```

from pyspark.ml.feature import Imputer
imputer = Imputer(
    inputCols = ['age' , 'Experience' , 'Salary'],
    outputCols = ["{}_imputed".format(c) for c in ['age', 'Experience','Salary']]

).setStrategy("mean")

# Add imputation Col to df
imputer.fit(df_pyspark).transform(df_pyspark).show()

```

Name	age	Experience	Salary	age_imputed	Experience_imputed	Salary_imputed
Krish	31	10	30000	31	10	30000
Sudhanshu	30	8	25000	30	8	25000
Sunny	29	4	20000	29	4	20000
Paul	24	3	20000	24	3	20000
Harsha	21	1	15000	21	1	15000
Shubham	23	2	18000	23	2	18000
Mahesh	NULL	NULL	40000	28	5	40000
NULL	34	10	38000	34	10	38000
NULL	36	NULL	NULL	36	5	25750

```

# Now Handling the NULL values with the Mode
df_pyspark.show()

```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```

from pyspark.ml.feature import Imputer
imputer = Imputer(
    inputCols = ['age' , 'Experience' , 'Salary'],
    outputCols = ["{}_imputed".format(c) for c in ['age', 'Experience','Salary']]

).setStrategy("mode")

# Add imputation Col to df
imputer.fit(df_pyspark).transform(df_pyspark).show()

```

Name	age	Experience	Salary	age_imputed	Experience_imputed	Salary_imputed
Krish	31	10	30000	31	10	30000
Sudhanshu	30	8	25000	30	8	25000
Sunny	29	4	20000	29	4	20000
Paul	24	3	20000	24	3	20000

Harsha	21	1	15000	21	1	15000
Shubham	23	2	18000	23	2	18000
Mahesh	NULL	NULL	40000	21	10	40000
NULL	34	10	38000	34	10	38000
NULL	36	NULL	NULL	36	10	20000

```
# Now Handling the NULL values with the Median
df_pyspark.show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000
Mahesh	NULL	NULL	40000
NULL	34	10	38000
NULL	36	NULL	NULL

```
from pyspark.ml.feature import Imputer
imputer = Imputer(
    inputCols = ['age' , 'Experience' , 'Salary'],
    outputCols = ["{}_imputed".format(c) for c in ['age', 'Experience','Salary']]
).setStrategy("median")
```

```
# Add imputation Col to df
imputer.fit(df_pyspark).transform(df_pyspark).show()
```

Name	age	Experience	Salary	age_imputed	Experience_imputed	Salary_imputed
Krish	31	10	30000	31	10	30000
Sudhanshu	30	8	25000	30	8	25000
Sunny	29	4	20000	29	4	20000
Paul	24	3	20000	24	3	20000
Harsha	21	1	15000	21	1	15000
Shubham	23	2	18000	23	2	18000
Mahesh	NULL	NULL	40000	29	4	40000
NULL	34	10	38000	34	10	38000
NULL	36	NULL	NULL	36	4	20000

- Filter Operation
- &,|, ==
- ~

```
from pyspark.sql import SparkSession
```

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

```
df_pyspark = spark.read.csv('test1.csv', header = True , inferSchema = True )
```

```
df_pyspark.show()
```

Name	age	Experience	Salary
------	-----	------------	--------

	Krish	31		10 30000
	Sudhanshu	30		8 25000
	Sunny	29		4 20000
	Paul	24		3 20000
	Harsha	21		1 15000
	Shubham	23		2 18000

Filter Operations

Salary of the people less than or equal 20000

```
df_pyspark.filter("Salary<=20000").show()
```

	Name	age	Experience	Salary
	Sunny	29		4 20000
	Paul	24		3 20000
	Harsha	21		1 15000
	Shubham	23		2 18000

Aliter

```
df_pyspark.filter(df_pyspark['Salary'] <= 20000).show()
```

	Name	age	Experience	Salary
	Sunny	29		4 20000
	Paul	24		3 20000
	Harsha	21		1 15000
	Shubham	23		2 18000

```
df_pyspark.filter("Salary<=20000").select(['Name','age']).show()
```

	Name	age	
	Sunny	29	
	Paul	24	
	Harsha	21	
	Shubham	23	

Use of multiple condition with &

```
df_pyspark.filter((df_pyspark['Salary'] <= 20000) &
                  (df_pyspark['Salary'] >= 18000)).show()
```

	Name	age	Experience	Salary
	Sunny	29		4 20000
	Paul	24		3 20000
	Shubham	23		2 18000

Use of multiple condition with |

```
df_pyspark.filter((df_pyspark['Salary'] <= 20000) |
                  (df_pyspark['Salary'] >= 18000)).show()
```

	Name	age	Experience	Salary
--	------	-----	------------	--------

Krish	31	10	30000
Sudhanshu	30	8	25000
Sunny	29	4	20000
Paul	24	3	20000
Harsha	21	1	15000
Shubham	23	2	18000

Use of Not(~) condition

```
df_pyspark.filter(~(df_pyspark['Salary'] <= 20000)).show()
```

Name	age	Experience	Salary
Krish	31	10	30000
Sudhanshu	30	8	25000

- Group by
- And Aggregate functions

```
from pyspark.sql import SparkSession
```

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

```
df_pyspark = spark.read.csv('test3.csv', header = True , inferSchema = True )
```

```
df_pyspark.show()
```

Name	Departments	salary
Krish	Data Science	10000
Krish	IOT	5000
Mahesh	Big Data	4000
Krish	Big Data	4000
Mahesh	Data Science	3000
Sudhanshu	Data Science	20000
Sudhanshu	IOT	10000
Sudhanshu	Big Data	5000
Sunny	Data Science	10000
Sunny	Big Data	2000

```
df_pyspark.printSchema()
```

```
root
 |-- Name: string (nullable = true)
 |-- Departments: string (nullable = true)
 |-- salary: integer (nullable = true)
```

Group by operation

gives the total salary by Name as group

```
df_pyspark.groupBy('Name').sum().show()
```

Name	sum(salary)
Sudhanshu	35000
Sunny	12000
Krish	19000
Mahesh	7000

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

```
# who is getting maximum salary
df_pyspark.groupBy('Name').max().show()
```

```
+-----+
|      Name|max(salary)|
+-----+
|Sudhanshu|      20000|
|   Sunny|      10000|
|   Krish|      10000|
|  Mahesh|       4000|
+-----+
```

```
# who is getting minimum salary
df_pyspark.groupBy('Name').min().show()
```

```
+-----+
|      Name|min(salary)|
+-----+
|Sudhanshu|       5000|
|   Sunny|       2000|
|   Krish|       4000|
|  Mahesh|       3000|
+-----+
```

```
# group by department to find the maximum salary
df_pyspark.groupBy('Departments').sum().show()
```

```
+-----+-----+
|Departments|sum(salary)|
+-----+-----+
|      IOT|      15000|
|Big Data|      15000|
|Data Science|    43000|
+-----+-----+
```

```
# to find the Department wise mean
df_pyspark.groupBy('Departments').mean().show()
```

```
+-----+-----+
|Departments|avg(salary)|
+-----+-----+
|      IOT|    7500.0|
|Big Data|    3750.0|
|Data Science|  10750.0|
+-----+-----+
```

```
# Number of employees using count
df_pyspark.groupBy('Departments').count().show()
```

```
+-----+-----+
|Departments|count|
+-----+-----+
|      IOT|     2|
|Big Data|     4|
|Data Science|     4|
+-----+-----+
```

```
df_pyspark.agg({'Salary':'sum'}).show()
```

+-----+	
sum(Salary)	
+-----+	
	73000
+-----+	