



PYSPARK COURSE

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```
# Load Pyspark Pkgs
import pyspark
```

```
# Methods/Attrib
dir(pyspark)
```

```
↳ ['Accumulator',
   'AccumulatorParam',
   'Any',
   'BarrierTaskContext',
   'BarrierTaskInfo',
   'BasicProfiler',
   'Broadcast',
   'CPickleSerializer',
   'Callable',
   'HiveContext',
   'InheritableThread',
   'MarshalSerializer',
   'Optional',
   'Profiler',
   'RDD',
   'RDDBarrier',
   'Row',
   'SQLContext',
   'SparkConf',
   'SparkContext',
   'SparkFiles',
   'SparkJobInfo',
   'SparkStageInfo',
   'StatusTracker',
   'StorageLevel',
   'TaskContext',
   'TypeVar',
   'Union',
   '_F',
   '_NoValue',
   '__all__',
   '__builtins__',
   '__cached__',
   '__doc__',
   '__file__',
   '__loader__',
   '__name__',
   '__package__',
   '__path__',
   '__spec__',
   '__version__',
   '__globals__',
   'accumulators',
   'broadcast',
   'cast',
   'cloudpickle',
```

```
'conf',
'context',
'copy_func',
'errors',
'files',
'filterwarnings',
'find_spark_home',
'inheritable_thread_target',
'java_gateway',
'join',
'keyword_only',
'profiler',
```

✳ Working with DataFrames in PySpark

- Read DataSet(CSV)
- Create DataFrame

Tips

- SparkSession
- SparkContext :sc
- SqlContext

```
# Create A SparkSession
from pyspark.sql import SparkSession
spark = SparkSession.builder.appName("PySparkTut").getOrCreate()
```

```
!ls
```

```
📁 sample_data
```

```
!wget https://raw.githubusercontent.com/Jcharis/common_ml_datasets_explorer_app/master/da
```

```
📁 --2024-11-12 07:07:43-- https://raw.githubusercontent.com/Jcharis/common_ml_datasets
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 185.199.108.133, 1
Connecting to raw.githubusercontent.com (raw.githubusercontent.com)|185.199.108.133|:
HTTP request sent, awaiting response... 200 OK
Length: 2772143 (2.6M) [text/plain]
Saving to: 'diamonds.csv'
```

```
diamonds.csv      100%[=====>]    2.64M  --.-KB/s    in 0.08s
```

```
2024-11-12 07:07:43 (34.7 MB/s) - 'diamonds.csv' saved [2772143/2772143]
```

```
!ls
```

```
📁 diamonds.csv  sample_data
```

```
# Read A DataSet without header
df = spark.read.csv('diamonds.csv')
```

```
# Preview dataset
df.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|_c0|_c1|_c2|_c3|_c4|_c5|_c6|_c7|_c8|_c9|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|cut|color|clarity|depth|table|price|x|y|z|
| 0.23|Ideal|E|SI2|61.5|55|326|3.95|3.98|2.43|
| 0.21|Premium|E|SI1|59.8|61|326|3.89|3.84|2.31|
| 0.23|Good|E|VS1|56.9|65|327|4.05|4.07|2.31|
| 0.29|Premium|I|VS2|62.4|58|334|4.2|4.23|2.63|
| 0.31|Good|J|SI2|63.3|58|335|4.34|4.35|2.75|
| 0.24|Very Good|J|VVS2|62.8|57|336|3.94|3.96|2.48|
| 0.24|Very Good|I|VVS1|62.3|57|336|3.95|3.98|2.47|
| 0.26|Very Good|H|SI1|61.9|55|337|4.07|4.11|2.53|
| 0.22|Fair|E|VS2|65.1|61|337|3.87|3.78|2.49|
| 0.23|Very Good|H|VS1|59.4|61|338|4|4.05|2.39|
| 0.3|Good|J|SI1|64|55|339|4.25|4.28|2.73|
| 0.23|Ideal|J|VS1|62.8|56|340|3.93|3.9|2.46|
| 0.22|Premium|F|SI1|60.4|61|342|3.88|3.84|2.33|
| 0.31|Ideal|J|SI2|62.2|54|344|4.35|4.37|2.71|
| 0.2|Premium|E|SI2|60.2|62|345|3.79|3.75|2.27|
| 0.32|Premium|E|I1|60.9|58|345|4.38|4.42|2.68|
| 0.3|Ideal|I|SI2|62|54|348|4.31|4.34|2.68|
| 0.3|Good|J|SI1|63.4|54|351|4.23|4.29|2.7|
| 0.3|Good|J|SI1|63.8|56|351|4.23|4.26|2.71|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows
```

```
# Read A DataSet with header/column names
df = spark.read.csv('diamonds.csv',header=True)
```

```
df.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|cut|color|clarity|depth|table|price|x|y|z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.23|Ideal|E|SI2|61.5|55|326|3.95|3.98|2.43|
| 0.21|Premium|E|SI1|59.8|61|326|3.89|3.84|2.31|
| 0.23|Good|E|VS1|56.9|65|327|4.05|4.07|2.31|
| 0.29|Premium|I|VS2|62.4|58|334|4.2|4.23|2.63|
| 0.31|Good|J|SI2|63.3|58|335|4.34|4.35|2.75|
| 0.24|Very Good|J|VVS2|62.8|57|336|3.94|3.96|2.48|
| 0.24|Very Good|I|VVS1|62.3|57|336|3.95|3.98|2.47|
| 0.26|Very Good|H|SI1|61.9|55|337|4.07|4.11|2.53|
| 0.22|Fair|E|VS2|65.1|61|337|3.87|3.78|2.49|
```

0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71
0.3	Very Good	J	SI1	62.7	59	351	4.21	4.27	2.66

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

only showing top 20 rows

```
# Columns
df.columns
```

```
['carat', 'cut', 'color', 'clarity', 'depth', 'table', 'price', 'x', 'y', 'z']
```

```
# Shape (rows + columns)
(df.count(), len(df.columns))
```

```
(53940, 10)
```

```
# Number of columns
len(df.columns)
```

```
10
```

```
# Number of rows
df.count()
```

```
53940
```

```
# Descriptive Analysis
df.describe().show()
```

summary	carat	cut	color	clarity	depth	t
count	53940	53940	53940	53940	53940	5
mean	0.7979397478679852	NULL	NULL	NULL	61.74940489432624	57.4571839080
stddev	0.4740112444054196	NULL	NULL	NULL	1.4326213188336525	2.234490562821
min	0.2	Fair	D	I1	43	
max	5.01	Very Good	J	VVS2	79	

```
# Pick a column & Get summary/describe a selected column
df.describe('carat').show()
```

```
+-----+-----+
|summary|      carat|
+-----+-----+
|  count|      53940|
|   mean|0.7979397478679852|
|  stddev|0.4740112444054196|
|    min|         0.2|
|    max|         5.01|
+-----+-----+
```

```
# Preview the First Row
df.first()
```

```
Row(carat='0.23', cut='Ideal', color='E', clarity='SI2', depth='61.5', table='55',
price='326', x='3.95', y='3.98', z='2.43')
```

```
# Preview the first 10 rows
# Like a list
df.head(10)
```

```
[Row(carat='0.23', cut='Ideal', color='E', clarity='SI2', depth='61.5', table='55',
price='326', x='3.95', y='3.98', z='2.43'),
 Row(carat='0.21', cut='Premium', color='E', clarity='SI1', depth='59.8',
table='61', price='326', x='3.89', y='3.84', z='2.31'),
 Row(carat='0.23', cut='Good', color='E', clarity='VS1', depth='56.9', table='65',
price='327', x='4.05', y='4.07', z='2.31'),
 Row(carat='0.29', cut='Premium', color='I', clarity='VS2', depth='62.4',
table='58', price='334', x='4.2', y='4.23', z='2.63'),
 Row(carat='0.31', cut='Good', color='J', clarity='SI2', depth='63.3', table='58',
price='335', x='4.34', y='4.35', z='2.75'),
 Row(carat='0.24', cut='Very Good', color='J', clarity='VVS2', depth='62.8',
table='57', price='336', x='3.94', y='3.96', z='2.48'),
 Row(carat='0.24', cut='Very Good', color='I', clarity='VVS1', depth='62.3',
table='57', price='336', x='3.95', y='3.98', z='2.47'),
 Row(carat='0.26', cut='Very Good', color='H', clarity='SI1', depth='61.9',
table='55', price='337', x='4.07', y='4.11', z='2.53'),
 Row(carat='0.22', cut='Fair', color='E', clarity='VS2', depth='65.1', table='61',
price='337', x='3.87', y='3.78', z='2.49'),
 Row(carat='0.23', cut='Very Good', color='H', clarity='VS1', depth='59.4',
table='61', price='338', x='4', y='4.05', z='2.39')]
```

```
# Method 2: Useful Action with show()
# Show first 10 datapoints
df.show(10)
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|    cut|color|clarity|depth|table|price|    x|    y|    z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
```

0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43
0.21	Premium	E	SI1	59.8	61	326	3.89	3.84	2.31
0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.31
0.29	Premium	I	VS2	62.4	58	334	4.2	4.23	2.63
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48
0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47
0.26	Very Good	H	SI1	61.9	55	337	4.07	4.11	2.53
0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49
0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39

+-----+
only showing top 10 rows

```
# Get Last Rows
df.tail(5)
```

```
[Row(carat='0.72', cut='Ideal', color='D', clarity='SI1', depth='60.8', table='57',
price='2757', x='5.75', y='5.76', z='3.5'),
 Row(carat='0.72', cut='Good', color='D', clarity='SI1', depth='63.1', table='55',
price='2757', x='5.69', y='5.75', z='3.61'),
 Row(carat='0.7', cut='Very Good', color='D', clarity='SI1', depth='62.8',
table='60', price='2757', x='5.66', y='5.68', z='3.56'),
 Row(carat='0.86', cut='Premium', color='H', clarity='SI2', depth='61', table='58',
price='2757', x='6.15', y='6.12', z='3.74'),
 Row(carat='0.75', cut='Ideal', color='D', clarity='SI2', depth='62.2', table='55',
price='2757', x='5.83', y='5.87', z='3.64')]
```

✱ Selection of columns

- .select

Note

- Dot & Bracket Notation only gives the column name not the entire column
 - ['colA']*
 - .colA*

```
# List all Columns
df.columns
```

```
['carat', 'cut', 'color', 'clarity', 'depth', 'table', 'price', 'x', 'y', 'z']
```

```
# Select A Column
df.select('carat').show()
```

```
+-----+
|carat|
```

```

+-----+
| 0.23|
| 0.21|
| 0.23|
| 0.29|
| 0.31|
| 0.24|
| 0.24|
| 0.26|
| 0.22|
| 0.23|
| 0.3|
| 0.23|
| 0.22|
| 0.31|
| 0.2|
| 0.32|
| 0.3|
| 0.3|
| 0.3|
| 0.3|

```

```

+-----+

```

only showing top 20 rows

```

# Select A Column irrespective of column word case
# will work irrespective of the case of the column once it is found within the dataset
df.select('CARAT').show()

```

```

+-----+
|CARAT|
+-----+
| 0.23|
| 0.21|
| 0.23|
| 0.29|
| 0.31|
| 0.24|
| 0.24|
| 0.26|
| 0.22|
| 0.23|
| 0.3|
| 0.23|
| 0.22|
| 0.31|
| 0.2|
| 0.32|
| 0.3|
| 0.3|
| 0.3|
| 0.3|

```

```

+-----+

```

only showing top 20 rows


```
# This is not as we would expect in pandas
# For Bracket Notation : pick column name not the entire column
df['carat']
```

```
Column<'carat'>
```

```
# This is not as we would expect in pandas
# For Dot Notation : pick column name not the entire column
df.carat
```

```
Column<'carat'>
```

```
# Select Multiple Columns
df.select('carat','cut').show(5)
```

```
+-----+-----+
|carat|    cut|
+-----+-----+
| 0.23|  Ideal|
| 0.21|Premium|
| 0.23|   Good|
| 0.29|Premium|
| 0.31|   Good|
+-----+-----+
only showing top 5 rows
```

✧ Column Filtering and Applying Conditions

- .filter
- .where

```
# Filter of Columns
# Apply A Condition
df.show(10)
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|    cut|color|clarity|depth|table|price|    x|    y|    z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.23|  Ideal|   E|   SI2| 61.5|   55|  326| 3.95| 3.98| 2.43|
| 0.21|Premium|   E|   SI1| 59.8|   61|  326| 3.89| 3.84| 2.31|
| 0.23|   Good|   E|   VS1| 56.9|   65|  327| 4.05| 4.07| 2.31|
| 0.29|Premium|   I|   VS2| 62.4|   58|  334| 4.2 | 4.23| 2.63|
| 0.31|   Good|   J|   SI2| 63.3|   58|  335| 4.34| 4.35| 2.75|
| 0.24|Very Good|   J|   VS2| 62.8|   57|  336| 3.94| 3.96| 2.48|
| 0.24|Very Good|   I|   VS1| 62.3|   57|  336| 3.95| 3.98| 2.47|
```

```

| 0.26|Very Good|H|SI1|61.9|55|337|4.07|4.11|2.53|
| 0.22|Fair|E|VS2|65.1|61|337|3.87|3.78|2.49|
| 0.23|Very Good|H|VS1|59.4|61|338|4|4.05|2.39|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 10 rows

```

```

# Method 1:using filter
df.filter(df['cut'] == "Good").show()

```

```

+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|cut|color|clarity|depth|table|price| x| y| z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.23|Good|E|VS1|56.9|65|327|4.05|4.07|2.31|
| 0.31|Good|J|SI2|63.3|58|335|4.34|4.35|2.75|
| 0.3|Good|J|SI1|64|55|339|4.25|4.28|2.73|
| 0.3|Good|J|SI1|63.4|54|351|4.23|4.29|2.7|
| 0.3|Good|J|SI1|63.8|56|351|4.23|4.26|2.71|
| 0.3|Good|I|SI2|63.3|56|351|4.26|4.3|2.71|
| 0.23|Good|F|VS1|58.2|59|402|4.06|4.08|2.37|
| 0.23|Good|E|VS1|64.1|59|402|3.83|3.85|2.46|
| 0.31|Good|H|SI1|64|54|402|4.29|4.31|2.75|
| 0.26|Good|D|VS2|65.2|56|403|3.99|4.02|2.61|
| 0.26|Good|D|VS1|58.4|63|403|4.19|4.24|2.46|
| 0.32|Good|H|SI2|63.1|56|403|4.34|4.37|2.75|
| 0.32|Good|H|SI2|63.8|56|403|4.36|4.38|2.79|
| 0.3|Good|I|SI1|63.2|55|405|4.25|4.29|2.7|
| 0.3|Good|H|SI1|63.7|57|554|4.28|4.26|2.72|
| 0.26|Good|E|VVS1|57.9|60|554|4.22|4.25|2.45|
| 0.7|Good|E|VS2|57.5|58|2759|5.85|5.9|3.38|
| 0.7|Good|F|VS1|59.4|62|2759|5.71|5.76|3.4|
| 0.7|Good|H|VVS2|62.1|64|2767|5.62|5.65|3.5|
| 0.71|Good|E|VS2|59.2|61|2772|5.8|5.88|3.46|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

```

```

# Method 1:using filter
df.filter(df.carat >= 0.7).show()

```

```

+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|cut|color|clarity|depth|table|price| x| y| z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.7|Ideal|E|SI1|62.5|57|2757|5.7|5.72|3.57|
| 0.86|Fair|E|SI2|55.1|69|2757|6.45|6.33|3.52|
| 0.7|Ideal|G|VS2|61.6|56|2757|5.7|5.67|3.5|
| 0.71|Very Good|E|VS2|62.4|57|2759|5.68|5.73|3.56|
| 0.78|Very Good|G|SI2|63.8|56|2759|5.81|5.85|3.72|
| 0.7|Good|E|VS2|57.5|58|2759|5.85|5.9|3.38|
| 0.7|Good|F|VS1|59.4|62|2759|5.71|5.76|3.4|
| 0.96|Fair|F|SI2|66.3|62|2759|6.27|5.95|4.07|
| 0.73|Very Good|E|SI1|61.6|59|2760|5.77|5.78|3.56|
| 0.8|Premium|H|SI1|61.5|58|2760|5.97|5.93|3.66|

```

```

0.75|Very Good| D| SI1| 63.2| 56| 2760| 5.8|5.75|3.65|
| 0.75| Premium| E| SI1| 59.9| 54| 2760| 6|5.96|3.58|
| 0.74| Ideal| G| SI1| 61.6| 55| 2760| 5.8|5.85|3.59|
| 0.75| Premium| G| VS2| 61.7| 58| 2760|5.85|5.79|3.59|
| 0.8| Ideal| I| VS1| 62.9| 56| 2760|5.94|5.87|3.72|
| 0.75| Ideal| G| SI1| 62.2| 55| 2760|5.87| 5.8|3.63|
| 0.8| Premium| G| SI1| 63| 59| 2760| 5.9|5.81|3.69|
| 0.74| Ideal| I| VVS2| 62.3| 55| 2761|5.77|5.81|3.61|
| 0.81| Ideal| F| SI2| 58.8| 57| 2761|6.14|6.11| 3.6|
| 0.8| Ideal| F| SI2| 61.4| 57| 2761|5.96| 6|3.67|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

```

```

# Method 2: where
df.where(df['cut'] == 'Good').show()

```

```

+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat| cut|color|clarity|depth|table|price| x| y| z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.23|Good| E| VS1| 56.9| 65| 327|4.05|4.07|2.31|
| 0.31|Good| J| SI2| 63.3| 58| 335|4.34|4.35|2.75|
| 0.3|Good| J| SI1| 64| 55| 339|4.25|4.28|2.73|
| 0.3|Good| J| SI1| 63.4| 54| 351|4.23|4.29| 2.7|
| 0.3|Good| J| SI1| 63.8| 56| 351|4.23|4.26|2.71|
| 0.3|Good| I| SI2| 63.3| 56| 351|4.26| 4.3|2.71|
| 0.23|Good| F| VS1| 58.2| 59| 402|4.06|4.08|2.37|
| 0.23|Good| E| VS1| 64.1| 59| 402|3.83|3.85|2.46|
| 0.31|Good| H| SI1| 64| 54| 402|4.29|4.31|2.75|
| 0.26|Good| D| VS2| 65.2| 56| 403|3.99|4.02|2.61|
| 0.26|Good| D| VS1| 58.4| 63| 403|4.19|4.24|2.46|
| 0.32|Good| H| SI2| 63.1| 56| 403|4.34|4.37|2.75|
| 0.32|Good| H| SI2| 63.8| 56| 403|4.36|4.38|2.79|
| 0.3|Good| I| SI1| 63.2| 55| 405|4.25|4.29| 2.7|
| 0.3|Good| H| SI1| 63.7| 57| 554|4.28|4.26|2.72|
| 0.26|Good| E| VVS1| 57.9| 60| 554|4.22|4.25|2.45|
| 0.7|Good| E| VS2| 57.5| 58| 2759|5.85| 5.9|3.38|
| 0.7|Good| F| VS1| 59.4| 62| 2759|5.71|5.76| 3.4|
| 0.7|Good| H| VVS2| 62.1| 64| 2767|5.62|5.65| 3.5|
| 0.71|Good| E| VS2| 59.2| 61| 2772| 5.8|5.88|3.46|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

```

```

# Method 2: where
# select certain columns
df.where(df['cut'] == 'Good').select('price','cut').show()

```

```

+-----+-----+
|price| cut|
+-----+-----+
| 327|Good|
| 335|Good|

```

```

339|Good|
| 351|Good|
| 351|Good|
| 351|Good|
| 402|Good|
| 402|Good|
| 402|Good|
| 403|Good|
| 403|Good|
| 403|Good|
| 403|Good|
| 405|Good|
| 554|Good|
| 554|Good|
| 2759|Good|
| 2759|Good|
| 2767|Good|
| 2772|Good|

```

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

only showing top 20 rows

```

# Unique Values
# df['cut'].unique()
df.select("cut").distinct().show()

```

```

+-----+
|      cut|
+-----+
|  Premium|
|   Ideal|
|    Good|
|    Fair|
|Very Good|
+-----+

```

✧ How to Add Columns & Delete/Drop Columns

- .withColumn()
- .drop()

```

# Add Columns
df.withColumn("carat10x",df['carat'] * 10).show()

```

```

+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|      cut|color|clarity|depth|table|price|  x|  y|  z|      carat10x|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.23|   Ideal|  E|   SI2| 61.5|  55|  326|3.95|3.98|2.43|2.3000000000000003|
| 0.21| Premium|  E|   SI1| 59.8|  61|  326|3.89|3.84|2.31|                2.1|
| 0.23|    Good|  E|   VS1| 56.9|  65|  327|4.05|4.07|2.31|2.3000000000000003|

```

0.29	Premium	I	VS2	62.4	58	334	4.2	4.23	2.63	2.9
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75	3.1
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48	2.4
0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47	2.4
0.26	Very Good	H	SI1	61.9	55	337	4.07	4.11	2.53	2.6
0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49	2.2
0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39	2.3000000000000003
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73	3.0
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46	2.3000000000000003
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33	2.2
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71	3.1
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27	2.0
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68	3.2
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68	3.0
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7	3.0
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71	3.0
0.3	Very Good	J	SI1	62.7	59	351	4.21	4.27	2.66	3.0

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

```
df.show()
```

carat	cut	color	clarity	depth	table	price	x	y	z
0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43
0.21	Premium	E	SI1	59.8	61	326	3.89	3.84	2.31
0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.31
0.29	Premium	I	VS2	62.4	58	334	4.2	4.23	2.63
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48
0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47
0.26	Very Good	H	SI1	61.9	55	337	4.07	4.11	2.53
0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49
0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71
0.3	Very Good	J	SI1	62.7	59	351	4.21	4.27	2.66

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

```
df2 = df.withColumn("carat10x",df['carat'] * 10)
```

```
df2.show()
```

carat	cut	color	clarity	depth	table	price	x	y	z	carat10x
0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43	2.3000000000000003
0.21	Premium	E	SI1	59.8	61	326	3.89	3.84	2.31	2.1
0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.31	2.3000000000000003
0.29	Premium	I	VS2	62.4	58	334	4.2	4.23	2.63	2.9
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75	3.1
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48	2.4
0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47	2.4
0.26	Very Good	H	SI1	61.9	55	337	4.07	4.11	2.53	2.6
0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49	2.2
0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39	2.3000000000000003
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73	3.0
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46	2.3000000000000003
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33	2.2
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71	3.1
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27	2.0
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68	3.2
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68	3.0
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7	3.0
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71	3.0
0.3	Very Good	J	SI1	62.7	59	351	4.21	4.27	2.66	3.0

only showing top 20 rows

```
# Delete/Drop A Column
df2.drop('carat10x').show()
```

carat	cut	color	clarity	depth	table	price	x	y	z
0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43
0.21	Premium	E	SI1	59.8	61	326	3.89	3.84	2.31
0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.31
0.29	Premium	I	VS2	62.4	58	334	4.2	4.23	2.63
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48
0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47
0.26	Very Good	H	SI1	61.9	55	337	4.07	4.11	2.53
0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49
0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71

```
| 0.3|Very Good| J| SI1| 62.7| 59| 351|4.21|4.27|2.66
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows
```

✧ GroupBy

- value counts
- aggregate

```
# Value Counts
# df['cut'].value_counts()
# df.groupby('cut')[].size()
df.groupby('cut').count().show()
```

```
+-----+-----+
|      cut|count|
+-----+-----+
|  Premium|13791|
|   Ideal|21551|
|    Good| 4906|
|    Fair| 1610|
|Very Good|12082|
+-----+-----+
```

```
# More Groupby
df.groupby('price').mean().show()
```

```
+-----+
|price|
+-----+
| 2904|
| 3210|
| 3414|
| 3606|
| 3959|
| 4032|
| 4821|
| 4937|
| 5325|
| 6194|
| 6240|
| 6613|
| 6731|
| 7273|
| 7711|
| 7762|
| 9009|
| 9030|
| 9586|
```

```
|10096|
+-----+
only showing top 20 rows
```

```
# Sum of A groupby
df.groupBy('price').sum().show()
```

```
+-----+
|price|
+-----+
| 2904|
| 3210|
| 3414|
| 3606|
| 3959|
| 4032|
| 4821|
| 4937|
| 5325|
| 6194|
| 6240|
| 6613|
| 6731|
| 7273|
| 7711|
| 7762|
| 9009|
| 9030|
| 9586|
|10096|
+-----+
only showing top 20 rows
```

```
# # Aggregation
# df.groupBy('carat').agg('col':'sum')
```

```
df.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|carat|      cut|color|clarity|depth|table|price|    x|    y|    z|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0.23|   Ideal|   E|   SI2| 61.5|  55|  326|3.95|3.98|2.43|
| 0.21|  Premium|   E|   SI1| 59.8|  61|  326|3.89|3.84|2.31|
| 0.23|    Good|   E|   VS1| 56.9|  65|  327|4.05|4.07|2.31|
| 0.29|  Premium|   I|   VS2| 62.4|  58|  334| 4.2|4.23|2.63|
| 0.31|    Good|   J|   SI2| 63.3|  58|  335|4.34|4.35|2.75|
| 0.24| Very Good|   J|  VVS2| 62.8|  57|  336|3.94|3.96|2.48|
| 0.24| Very Good|   I|  VVS1| 62.3|  57|  336|3.95|3.98|2.47|
| 0.26| Very Good|   H|   SI1| 61.9|  55|  337|4.07|4.11|2.53|
| 0.22|    Fair|   E|   VS2| 65.1|  61|  337|3.87|3.78|2.49|
```


0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71
0.3	Very Good	J	SI1	62.7	59	351	4.21	4.27	2.66

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

```
df.columns
```

```
['carat', 'cut', 'color', 'clarity', 'depth', 'table', 'price', 'x', 'y', 'z']
```

```
# Rearrange Columns
```

```
df.select('carat', 'color', 'clarity', 'depth', 'table', 'price', 'x', 'y', 'z', 'cut').sh
```

carat	color	clarity	depth	table	price	x	y	z	cut
0.23	E	SI2	61.5	55	326	3.95	3.98	2.43	Ideal
0.21	E	SI1	59.8	61	326	3.89	3.84	2.31	Premium
0.23	E	VS1	56.9	65	327	4.05	4.07	2.31	Good
0.29	I	VS2	62.4	58	334	4.2	4.23	2.63	Premium
0.31	J	SI2	63.3	58	335	4.34	4.35	2.75	Good
0.24	J	VVS2	62.8	57	336	3.94	3.96	2.48	Very Good
0.24	I	VVS1	62.3	57	336	3.95	3.98	2.47	Very Good
0.26	H	SI1	61.9	55	337	4.07	4.11	2.53	Very Good
0.22	E	VS2	65.1	61	337	3.87	3.78	2.49	Fair
0.23	H	VS1	59.4	61	338	4	4.05	2.39	Very Good
0.3	J	SI1	64	55	339	4.25	4.28	2.73	Good
0.23	J	VS1	62.8	56	340	3.93	3.9	2.46	Ideal
0.22	F	SI1	60.4	61	342	3.88	3.84	2.33	Premium
0.31	J	SI2	62.2	54	344	4.35	4.37	2.71	Ideal
0.2	E	SI2	60.2	62	345	3.79	3.75	2.27	Premium
0.32	E	I1	60.9	58	345	4.38	4.42	2.68	Premium
0.3	I	SI2	62	54	348	4.31	4.34	2.68	Ideal
0.3	J	SI1	63.4	54	351	4.23	4.29	2.7	Good
0.3	J	SI1	63.8	56	351	4.23	4.26	2.71	Good
0.3	J	SI1	62.7	59	351	4.21	4.27	2.66	Very Good

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

```
# Assign DF to a New DF
```

```
new_df = df.select('carat', 'color', 'clarity', 'depth', 'table', 'price', 'x', 'y', 'z',
```

```
new_df
```

```
DataFrame[carat: string, color: string, clarity: string, depth: string, table:
string, price: string, x: string, y: string, z: string, cut: string]
```

```
# Check Datatype
```

```
new_df.dtypes
```

```
[('carat', 'string'),
 ('color', 'string'),
 ('clarity', 'string'),
 ('depth', 'string'),
 ('table', 'string'),
 ('price', 'string'),
 ('x', 'string'),
 ('y', 'string'),
 ('z', 'string'),
 ('cut', 'string')]
```

```
# Check For the Schema
```

```
df.printSchema()
```

```
root
|-- carat: string (nullable = true)
|-- cut: string (nullable = true)
|-- color: string (nullable = true)
|-- clarity: string (nullable = true)
|-- depth: string (nullable = true)
|-- table: string (nullable = true)
|-- price: string (nullable = true)
|-- x: string (nullable = true)
|-- y: string (nullable = true)
|-- z: string (nullable = true)
```

```
# Check type of DF
```

```
type(df)
```

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

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

</usr/local/lib/python3.10/dist-packages/pyspark/sql/dataframe.py> _

A distributed collection of data grouped into named columns.

```
.. versionadded:: 1.3.0
```

```
.. versionchanged:: 3.4.0
```

* Saving DataFrames as CSV,parquet etc

```
# Save
new_df.write.format('csv').option('header','true').save("diamond_clean.csv")
```

```
!ls
```

```
    diamond_clean.csv  diamonds.csv  sample_data
```

```
# Save as parquet
new_df.write.format('parquet').save("diamond_clean.parquet")
```

✧ Making SQL Queries

- parse in the spark.SparkContext
- sqlContext

```
from pyspark.sql import SQLContext
```

```
dir(spark)
```

```
['Builder',
 '_annotations_',
 '_class_',
 '_delattr_',
 '_dict_',
 '_dir_',
 '_doc_',
 '_enter_',
 '_eq_',
 '_exit_',
 '_format_',
 '_ge_',
 '_getattribute_',
 '_gt_',
 '_hash_',
 '_init_',
 '_init_subclass_',
 '_le_',
 '_lt_',
 '_module_',
 '_ne_',
 '_new_',
 '_reduce_',
 '_reduce_ex_',
 '_repr_',
 '_setattr_',
 '_sizeof_',
```

```

'__str__',
'__subclasshook__',
'__weakref__',
'_activeSession',
'_convert_from_pandas',
'_createFromLocal',
'_createFromRDD',
'_create_dataframe',
'_create_from_pandas_with_arrow',
'_create_shell_session',
'_getActiveSessionOrCreate',
'_get_numpy_record_dtype',
'_inferSchema',
'_inferSchemaFromList',
'_instantiatedSession',
'_jconf',
'_jsc',
'_jsparkSession',
'_jvm',
'_repr_html_',
'_sc',
'active',
'addArtifact',
'addArtifacts',
'addTag',
'builder',
'catalog',
'clearTags',
'client',
'conf',
'copyFromLocalToFs',

```

```

# Create A Spark Context From the Spark Session
sc = spark.sparkContext

```

```

# Parse into the SQLContext
sqlContext = SQLContext(sc)

```

```

/usr/local/lib/python3.10/dist-packages/pyspark/sql/context.py:113: FutureWarning: De
warnings.warn(

```

```

# Register Current DataFrame As Temporal Table
df.registerTempTable("DiamondsTable")

```

```

/usr/local/lib/python3.10/dist-packages/pyspark/sql/dataframe.py:329: FutureWarning:
warnings.warn("Deprecated in 2.0, use createOrReplaceTempView instead.", FutureWarn

```

```

# Making Quries
sqlContext.sql('SELECT * FROM DiamondsTable').show()

```

carat	cut	color	clarity	depth	table	price	x	y	z
0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43
0.21	Premium	E	SI1	59.8	61	326	3.89	3.84	2.31
0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.31
0.29	Premium	I	VS2	62.4	58	334	4.2	4.23	2.63
0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75
0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48
0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47
0.26	Very Good	H	SI1	61.9	55	337	4.07	4.11	2.53
0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49
0.23	Very Good	H	VS1	59.4	61	338	4	4.05	2.39
0.3	Good	J	SI1	64	55	339	4.25	4.28	2.73
0.23	Ideal	J	VS1	62.8	56	340	3.93	3.9	2.46
0.22	Premium	F	SI1	60.4	61	342	3.88	3.84	2.33
0.31	Ideal	J	SI2	62.2	54	344	4.35	4.37	2.71
0.2	Premium	E	SI2	60.2	62	345	3.79	3.75	2.27
0.32	Premium	E	I1	60.9	58	345	4.38	4.42	2.68
0.3	Ideal	I	SI2	62	54	348	4.31	4.34	2.68
0.3	Good	J	SI1	63.4	54	351	4.23	4.29	2.7
0.3	Good	J	SI1	63.8	56	351	4.23	4.26	2.71
0.3	Very Good	J	SI1	62.7	59	351	4.21	4.27	2.66

only showing top 20 rows

```
# Can also use it to work with DataFrames
dir(sqlContext)
```

```
['_annotations_',
 '_class_',
 '_delattr_',
 '_dict_',
 '_dir_',
 '_doc_',
 '_eq_',
 '_format_',
 '_ge_',
 '_getattr_',
 '_gt_',
 '_hash_',
 '_init_',
 '_init_subclass_',
 '_le_',
 '_lt_',
 '_module_',
 '_ne_',
 '_new_',
 '_reduce_',
 '_reduce_ex_',
 '_repr_',
 '_setattr_',
 '_sizeof_',
 '_str_',
```

```
'_subclasshook_',  
'_weakref_',  
'_get_or_create',  
'_inferSchema',  
'_instantiatedContext',  
'_jsc',  
'_jdbcContext',  
'_jvm',  
'_sc',  
'_ssql_ctx',  
'cacheTable',  
'clearCache',  
'createDataFrame',  
'createExternalTable',  
'dropTempTable',  
'getConf',  
'getOrCreate',  
'newSession',  
'range',  
'read',  
'readStream',  
'registerDataFrameAsTable',  
'registerFunction',  
'registerJavaFunction',  
'setConf',  
'sparkSession',  
'sql',  
'streams',  
'table',  
'tableNames',  
'tables',  
'udf',  
'udtf',
```

