

Module 4 Cheat Sheet: DataFrames and Spark SQL

| Package/Method | Description | Code Example |
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| appName() | A name for your job to display on the cluster web UI. | <pre>from pyspark.sql import SparkSession spark = SparkSession.builder.appName("MyApp").getOrCreate()</pre> |
| createDataFrame() | Used to load the data into a Spark DataFrame. | <pre>from pyspark.sql import SparkSession spark = SparkSession.builder.appName("MyApp").getOrCreate() data = [("Jhon", 30), ("Peter", 25), ("Bob", 35)] columns = ["name", "age"]</pre> <p>Creating a DataFrame</p> <pre>df = spark.createDataFrame(data, columns)</pre> |
| createTempView() | Create a temporary view that can later be used to query the data. The only required parameter is the name of the view. | <pre>df.createOrReplaceTempView("cust_tbl")</pre> |
| fillna() | Used to replace NULL/None values on all or selected multiple DataFrame columns with either zero (0), empty string, space, or any constant literal values. | <p>Replace NULL/None values in a DataFrame</p> <pre>filled_df = df.fillna(0)</pre> <p>Replace with zero</p> |
| filter() | Returns an iterator where the items are filtered through a function to test if the item is accepted or not. | <pre>filtered_df = df.filter(df['age'] > 30)</pre> |
| getOrCreate() | Get or instantiate a SparkContext and register it as a singleton object. | <pre>spark = SparkSession.builder.getOrCreate()</pre> |

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| groupby() | Used to collect the identical data into groups on DataFrame and perform count, sum, avg, min, max functions on the grouped data. | Grouping data and performing aggregation <pre>grouped_df = df.groupBy("age").agg({"age": "count"})</pre> |
| head() | Returns the first n rows for the object based on position. | Returning the first 5 rows <pre>first_5_rows = df.head(5)</pre> |
| import | Used to make code from one module accessible in another. Python imports are crucial for a successful code structure. You may reuse code and keep your projects manageable by using imports effectively, which can increase your productivity. | <pre>from pyspark.sql import SparkSession</pre> |
| pd.read_csv() | Required to access data from the CSV file from Pandas that retrieves data in the form of the data frame. | <pre>import pandas as pd</pre> Reading data from a CSV file into a DataFrame <pre>df_from_csv = pd.read_csv("data.csv")</pre> |
| pip | To ensure that requests will function, the pip program searches for the package in the Python Package Index (PyPI), resolves any dependencies, and installs everything in your current Python environment. | <pre>pip list</pre> |
| pip install | The pip install <package> command looks for the latest version of the package and installs it. | <pre>pip install pyspark</pre> |

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| printSchema() | Used to print or display the schema of the DataFrame or data set in tree format along with the column name and data type. If you have a DataFrame or data set with a nested structure, it displays the schema in a nested tree format. | <code>df.printSchema()</code> |
| rename() | Used to change the row indexes and the column labels. | <pre>import pandas as pd</pre> <p>Create a sample DataFrame</p> <pre>data = {'A': [1, 2, 3], 'B': [4, 5, 6]} df = pd.DataFrame(data)</pre> <p>Rename columns</p> <pre>df = df.rename(columns={'A': 'X', 'B': 'Y'})</pre> <p>The columns 'A' and 'B' are now renamed to 'X' and 'Y'</p> <pre>print(df)</pre> |
| select() | Used to select one or multiple columns, nested columns, column by index, all columns from the list, by regular expression from a DataFrame. select() is a transformation function in Spark and returns a new DataFrame with the selected columns. | <code>selected_df = df.select('name', 'age')</code> |
| show() | Spark DataFrame show() is used to display the contents of the DataFrame in a table row and column format. By default, it shows only twenty rows, and the column values are truncated at twenty characters. | <code>df.show()</code> |
| sort() | Used to sort DataFrame by ascending or descending order based on single or multiple | <p>Sorting DataFrame by a column in ascending order</p> <pre>sorted_df = df.sort("age")</pre> |

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| | columns. | <p>Sorting DataFrame by multiple columns in descending order</p> <pre>sorted_df_desc = df.sort(["age", "name"], ascending=[False, True])</pre> |
| SparkContext() | <p>It is an entry point to Spark and is defined in org.apache.spark package since version 1.x and used to programmatically create Spark RDD, accumulators, and broadcast variables on the cluster.</p> | <pre>from pyspark import SparkContext</pre> <p>Creating a SparkContext</p> <pre>sc = SparkContext("local", "MyApp")</pre> |
| SparkSession | <p>It is an entry point to Spark, and creating a SparkSession instance would be the first statement you would write to the program with RDD, DataFrame, and dataset</p> | <pre>from pyspark.sql import SparkSession</pre> <p>Creating a SparkSession</p> <pre>spark = SparkSession.builder.appName("MyApp").getOrCreate()</pre> |
| spark.read.json() | <p>Spark SQL can automatically infer the schema of a JSON data set and load it as a DataFrame. The read.json() function loads data from a directory of JSON files where each line of the files is a JSON object. Note that the file offered as a JSON file is not a typical JSON file.</p> | <pre>json_df = spark.read.json("customer.json")</pre> |
| spark.sql() | <p>To issue any SQL query, use the sql() method on the SparkSession instance. All spark.sql queries executed in this manner return a</p> | <pre>result = spark.sql("SELECT name, age FROM cust_tbl WHERE age > 30") result.show()</pre> |

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| | DataFrame on which you may perform further Spark operations if required. | |
| spark.udf.register() | In PySpark DataFrame, it is used to register a user-defined function (UDF) with Spark, making it accessible for use in Spark SQL queries. This allows you to apply custom logic or operations to DataFrame columns using SQL expressions. | <p>Registering a UDF (User-defined Function)</p> <pre>from pyspark.sql.functions import udf from pyspark.sql.types import StringType def my_udf(value): return value.upper() spark.udf.register("my_udf", my_udf, StringType())</pre> |
| where() | Used to filter the rows from DataFrame based on the given condition. Both filter() and where() functions are used for the same purpose. | <p>Filtering rows based on a condition</p> <pre>filtered_df = df.where(df['age'] > 30)</pre> |
| withColumn() | Transformation function of DataFrame used to change the value, convert the data type of an existing column, create a new column, and many more. | <p>Adding a new column and performing transformations</p> <pre>from pyspark.sql.functions import col new_df = df.withColumn("age_squared", col("age") ** 2)</pre> |
| withColumnRenamed() | Returns a new DataFrame by renaming an existing column. | <p>Renaming an existing column</p> <pre>renamed_df = df.withColumnRenamed("age", "years_old")</pre> |



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