100 PYSPARK FUNCTIONS

1. Create an empty DataFrame

You can create an empty DataFrame using spark.createDataFrame with no data.

```
from pyspark.sql import SparkSession
spark = SparkSession.builder.master("local").appName("Empty DataFrame").getOrCreate()

# Empty DataFrame with no data
empty_df = spark.createDataFrame([], "id INT, name STRING")
empty df.show()
```

2. Convert RDD to DataFrame

To convert an RDD to DataFrame, you need to define the schema.

```
rdd = spark.sparkContext.parallelize([(1, 'Alice'), (2, 'Bob')])
columns = ["id", "name"]
df_from_rdd = rdd.toDF(columns)
df_from_rdd.show()
```

3. Convert DataFrame to Pandas

You can convert a PySpark DataFrame to a Pandas DataFrame using toPandas().

```
pandas_df = df_from_rdd.toPandas()
print(pandas_df)
```

4. show()

The show() method displays the first n rows of a DataFrame.

```
df_from_rdd.show(5) # Display first 5 rows
```

5. StructType & StructField

These classes are used to define the schema for DataFrames.

```
from pyspark.sql.types import StructType, StructField, IntegerType, StringType

schema = StructType([
    StructField("id", IntegerType(), True),
    StructField("name", StringType(), True)
])

df_with_schema = spark.createDataFrame([(1, "Alice"), (2, "Bob")], schema)

df_with_schema.show()
```

6. Column Class

The Column class represents a column in a DataFrame and is used for performing operations.

```
from pyspark.sql import functions as F
df_with_column = df_from_rdd.withColumn("upper_name", F.upper(df_from_rdd['name']))
df_with_column.show()
```

7. select()

The select() method is used to select specific columns.

```
df_from_rdd.select("id").show() # Select only the 'id' column
```

8. collect()

collect() returns all the rows as a list of Row objects.

```
rows = df_from_rdd.collect()
print(rows)
```

9. withColumn()

This method is used to add or modify a column.

```
df_with_new_col = df_from_rdd.withColumn("id_squared", df_from_rdd["id"] * df_from_rdd["id"])
df_with_new_col.show()
```

10. withColumnRenamed()

Renames an existing column in the DataFrame.

```
df_renamed = df_from_rdd.withColumnRenamed("name", "full_name")
df_renamed.show()
```

11. where() & filter()

Both methods are used to filter rows based on conditions.

```
df_filtered = df_from_rdd.where(df_from_rdd['id'] > 1)
df_filtered.show()

# Alternatively, you can use filter()
df_filtered2 = df_from_rdd.filter(df_from_rdd['id'] > 1)
df_filtered2.show()
```

12. drop() & dropDuplicates()

Used to drop a column or remove duplicate rows.

```
# Dropping a column

df_dropped = df_from_rdd.drop("name")

df_dropped.show()

# Removing duplicates

df_no_duplicates = df_from_rdd.dropDuplicates()

df no_duplicates.show()
```

13. orderBy() and sort()

These methods are used for sorting data in DataFrame.

```
df_sorted = df_from_rdd.orderBy("id", ascending=False)
df_sorted.show()

# Equivalent to orderBy()
df_sorted2 = df_from_rdd.sort("id")
df_sorted2.show()
```

14. groupBy()

Used for group-by operations.

```
df_grouped = df_from_rdd.groupBy("id").count()
df_grouped.show()
```

15. join()

Used for joining DataFrames.

```
df2 = spark.createDataFrame([(1, 'Math'), (2, 'Science')], ["id", "subject"])
df_joined = df_from_rdd.join(df2, on="id")
df_joined.show()
```

16. union() & unionAll()

Both methods combine DataFrames, but unionAll is deprecated in favor of union().

```
df3 = spark.createDataFrame([(3, "Charlie")], ["id", "name"])
df_union = df_from_rdd.union(df3)
df_union.show()
```

17. unionByName()

Union DataFrames by column name.

```
df_union_by_name = df_from_rdd.unionByName(df2)
df_union_by_name.show()
```

18. UDF (User Defined Function)

UDFs are used to extend the functionality of Spark DataFrame with custom logic.

```
from pyspark.sql.functions import udf
from pyspark.sql.types import StringType

def add_exclamation(name):
    return name + "!"

add_udf = udf(add_exclamation, StringType())
df_udf = df_from_rdd.withColumn("excited_name", add_udf("name"))
df_udf.show()
```

19. transform()

transform() is used to apply transformations to a DataFrame.

```
df_transformed = df_from_rdd.transform(lambda df: df.withColumn("id_squared", df["id"]**2))
df_transformed.show()
```

20. apply()

Similar to transform(), apply() can be used for element-wise operations.

```
# Applying to every row

df_applied = df_from_rdd.rdd.map(lambda row: (row.id * 2, row.name)).toDF(["id", "name"])

df_applied.show()
```

21. map()

map() is used on an RDD to apply a function on each element.

```
rdd_mapped = df_from_rdd.rdd.map(lambda x: (x.id * 2, x.name))

df_mapped = rdd_mapped.toDF(["id", "name"])

df_mapped.show()
```

22. flatMap()

Used to flatten a collection of items.

```
rdd_flat = df_from_rdd.rdd.flatMap(lambda x: [(x.id, x.name), (x.id * 10, x.name)])
df_flat = rdd_flat.toDF(["id", "name"])
df_flat.show()
```

23. foreach()

foreach() is used for applying a function to each row in the DataFrame.

```
def print_row(row):
    print(row)

df_from_rdd.foreach(print_row)
```

24. sample() vs sampleBy()

sample() is used for random sampling, while sampleBy() allows sampling with stratification.

```
# sample()
df_sample = df_from_rdd.sample(fraction=0.5)
df_sample.show()

# sampleBy()
df_sample_by = df_from_rdd.sampleBy("id", fractions={1: 0.5, 2: 0.5})
df_sample_by.show()
```

25. fillna() & fill()

Used for handling missing values.

```
# fillna()
df_filled = df_from_rdd.fillna({'id': 0, 'name': 'Unknown'})
df_filled.show()

# fill()
df_filled2 = df_from_rdd.fill({'id': 0, 'name': 'Unknown'})
df_filled2.show()
```

26. pivot() (Row to Column)

Used to pivot data (convert rows to columns).

```
df_pivoted = df_from_rdd.groupBy("id").pivot("name").agg({"id": "count"})
df_pivoted.show()
```

27. partitionBy()

Partitioning the data by one or more columns for distributed processing.

```
df_partitioned = df_from_rdd.repartitionByRange(3, "id")
df_partitioned.show()
```

28. MapType (Map/Dict)

MapType is used for columns that represent key-value pairs.

```
from pyspark.sql.types import MapType, StringType

data = [(1, {"name": "Alice", "age": "25"}), (2, {"name": "Bob", "age": "30"})]
schema = StructType([StructField("id", IntegerType(), True), StructField("info", MapType(StringType(), StringType()), True)])
df_map = spark.createDataFrame(data, schema)
df_map.show()
```

Inbuilt - Functions

1.col():

```
df.select(F.col('id')).show()
```

Output:

```
+---+
| id|
+---+
| 1|
| 2|
| 3|
| 4|
+---+
```

2. lit():

```
df.withColumn('constant', F.lit(10)).show()
```

Output:

```
+--+---+
| id| name|constant|
+---+----+
| 1|Alice| 10|
| 2| Bob| 10|
| 3|Alice| 10|
| 4|Charlie| 10|
+---+----+
```

3. when():

```
df.withColumn('is_adult', F.when(F.col('id') > 1, 'Yes').otherwise('No')).show()
```

Output:

```
+---+----+
| id| name|is_adult|
+---+----+
| 1|Alice| No|
| 2| Bob| Yes|
| 3|Alice| Yes|
| 4|Charlie| Yes|
+---+----+
```

4. isNull():

```
df.filter(F.col('name').isNull()).show()
```

```
# No output since none of the values in 'name' are null
```



df.filter(F.col('name').isNotNull()).show()

Output:

```
+---+----+
| id| name|
+---+----+
| 1|Alice|
| 2| Bob|
| 3|Alice|
| 4|Charlie|
+---+----+
```

6.between()

df.filter(F.col('id').between(2, 3)).show()

Output:

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

| id| name|

+---+----+

| 2| Bob|

| 3|Alice|

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

7.like()

df.filter(F.col('name').like('A%')).show()

Output:

```
+---+----+
| id| name|
+---+----+
| 1|Alice|
| 3|Alice|
+---+----+
```

8.rlike()

df.filter(F.col('name').rlike('A.*')).show()

```
+---+----+
| id| name|
+---+----+
| 1|Alice|
| 3|Alice|
+---+----+
```

9.alias()

```
df.select(F.col('name').alias('user_name')).show()
```

Output:

```
+-----+
|user_name|
+-----+
| Alice|
| Bob|
| Alice|
| Charlie|
+-----+
```

10.cast()

```
df.withColumn('id_str', F.col('id').cast('string')).show()
```

Output:

```
+---+----+
| id| name|id_str |
+---+----+
| 1|Alice| 1|
| 2| Bob| 2|
| 3|Alice| 3|
| 4|Charlie| 4|
+---+----+
```

11.expr()

```
df.select(F.expr('id + 1')).show()
```

Output:

```
+-----+

|(id + 1)|

+-----+

| 2|

| 3|

| 4|

| 5|

+-----+
```

12.count()

```
df.select(F.count('id')).show()
```

```
+-----+
|count(id)|
+-----+
| 4|
```

13.countDistinct() df.select(F.countDistinct('name')).show() Output: +----+ |count(DISTINCT name)| +----+ 14. approx_count_distinct() df.select(F.approx_count_distinct('name')).show() Output: +----+ |approx_count_distinct(name)| 3| 15.sum() df.select(F.sum('id')).show() Output: +----+ |sum(id)| | 10| +----+ 16.sumDistinct() df.select(F.sumDistinct('id')).show() Output: +----+ |sum(DISTINCT id)| +----+ 10|

```
17. avg()
```

```
df.select(F.avg('id')).show()
```

```
18.min()
df.select(F.min('id')).show()
Output:
+----+
|min(id)|
+----+
| 1|
+----+
19.max()
df.select(F.max('id')).show()
Output:
+----+
|max(id)|
+----+
   4|
20.first()
df.select(F.first('id')).show()
Output:
+----+
|first(id) |
| 1 | +----+
21.last()
df.select(F.last('id')).show()
Output:
+----+
|last(id) |
+----+
    | 4
+----+
22.stddev()
df.select(F.stddev('id')).show()
Output:
+----+
|stddev(id)
+----+
|1.118033988749895 |
```

23.variance()

```
df.select(F.variance('id')).show()
```

Output:

```
+-----+
|variance(id) |
+-----+
| 1.25 |
+------+
```

24.collect_list()

```
df.groupBy('name').agg(F.collect_list('id')).show()
```

Output:

```
+----+
| name|collect_list(id)|
+----+
| Alice| [1, 3]|
| Bob| [2]|
| Charlie| [4]|
+----+
```

25.collect_set()

```
df.groupBy('name').agg(F.collect_set('id')).show()
```

Output:

26. corr(): Calculates the correlation between two columns.

```
df.select(F.corr('id', 'name')).show()
```

Output:

```
+-----+
|corr(id, name) |
+-----+
| null|
+------+
```

(Note: Since 'id' is numeric and 'name' is string, the correlation cannot be computed, so the result is null.)

27. covar_pop(): Calculates the population covariance between two columns.

```
df.select(F.covar_pop('id', 'name')).show()
```

Output:

```
+-----+
|covar_pop(id, name) |
+-----+
| null |
+------+
```

(Again, the covariance can't be computed between a numeric and a string column, so the result is null.)

28. covar_samp(): Calculates the sample covariance between two columns.

```
df.select(F.covar_samp('id', 'name')).show()
```

Output:

```
.

+-----+

|covar_samp(id, name) |

+-----+

| null |

+------+
```

(Similar to the previous functions, this too would return null for numeric and string columns.)

29. kurtosis(): Calculates the kurtosis of a column.

```
df.select(F.kurtosis('id')).show()
```

Output:

(The result indicates the kurtosis of the id values in the DataFrame.)

30. skewness(): Calculates the skewness of a column.

```
df.select(F.skewness('id')).show()
```

Output:

(Skewness is 0, indicating a symmetric distribution of the id values.)

31. approxQuantile(): Calculates approximate quantiles of a column.

```
df.approxQuantile('id', [0.25, 0.5, 0.75], 0.1)
```

Output:

```
[1.0, 2.5, 3.5]
```

(This returns the 25th, 50th, and 75th percentiles (or quantiles) of the id column.)

32. rank(): Assigns a rank to each row based on the window specification.

```
windowSpec = Window.orderBy('id')
df.withColumn('rank', F.rank().over(windowSpec)).show()
```

Output:

```
+--+---+
| id| name|rank|
+---+---+
| 1|Alice| 1|
| 2| Bob| 2|
| 3|Alice| 3|
| 4|Charlie| 4|
+---+---+
```

33. dense_rank(): Assigns a dense rank (without gaps) to each row based on the window specification.

df.withColumn('dense_rank', F.dense_rank().over(windowSpec)).show()

Output:

```
+--+---+
| id| name|dense_rank |
+--+---+
| 1|Alice| 1|
| 2| Bob| 2|
| 3|Alice| 3|
| 4|Charlie| 4|
+--+----+
```

34. ntile(): Divides the data into n buckets and assigns a bucket number to each row.

df.withColumn('ntile', F.ntile(2).over(windowSpec)).show()

Output:

```
+--+---+
| id| name|ntile|
+--+---+
| 1|Alice| 1|
| 2| Bob| 1|
| 3|Alice| 2|
| 4|Charlie| 2|
+--+---+
```

35. row_number(): Assigns a unique row number to each row.

df.withColumn('row_number', F.row_number().over(windowSpec)).show()

```
+--+---+
| id| name|row_number|
+--+----+
| 1|Alice| 1|
| 2| Bob| 2|
| 3|Alice| 3|
| 4|Charlie| 4|
+--+----+
```

36. lead(): Accesses the value of a column in the next row.

```
df.withColumn('next_id', F.lead('id').over(windowSpec)).show()
```

Output:

```
+---+----+
| id| name|next_id|
+---+----+
| 1|Alice| 2|
| 2| Bob| 3|
| 3|Alice| 4|
| 4|Charlie| null|
+---+----+
```

37. lag(): Accesses the value of a column in the previous row.

```
df.withColumn('prev_id', F.lag('id').over(windowSpec)).show()
```

Output:

```
+---+----+
| id| name|prev_id|
+---+----+
| 1|Alice| null|
| 2| Bob| 1|
| 3|Alice| 2|
| 4|Charlie| 3|
+---+----+
```

38. percent_rank(): Calculates the percentile rank of each row based on the window specification.

```
df.withColumn('percent_rank', F.percent_rank().over(windowSpec)).show()
```

Output:

39. window()

This function is used to define a window specification, which is typically used in conjunction with window functions such as rank(), dense_rank(), row_number(), etc. This has already been shown in some examples above (e.g., Window.orderBy('id')).

String Functions:

40. concat(): Concatenates multiple columns into a single column.

```
df.withColumn('full_name', F.concat(F.col('name'), F.lit('Doe'))).show()
```

Output:

```
+---+----+
| id| name|full_name|
+---+----+
| 1|Alice|Alice Doe|
| 2| Bob| Bob Doe|
| 3|Alice|Alice Doe|
| 4|Charlie|Charlie Doe|
+---+----+
```

41. concat_ws(): Concatenates multiple columns with a separator.

```
df.withColumn('full name', F.concat ws(' ', F.col('name'), F.lit('Doe'))).show()
```

Output:

```
+---+----+
| id| name|full_name|
+---+----+
| 1|Alice|Alice Doe|
| 2| Bob| Bob Doe|
| 3|Alice|Alice Doe|
| 4|Charlie|Charlie Doe|
| +---+----+
```

42. length(): Returns the length of a string.

```
df.withColumn('name length', F.length(F.col('name'))).show()
```

Output:

```
+---+----+
| id| name|name_length|
+---+----+
| 1|Alice| 5|
| 2| Bob| 3|
| 3|Alice| 5|
| 4|Charlie| 7|
+---+----+
```

43. lower(): Converts a string to lowercase.

```
df.withColumn('lower_name', F.lower(F.col('name'))).show()
```

```
+---+----+
| id | name | lower_name |
+---+----+
| 1 | Alice | alice |
| 2 | Bob | bob |
| 3 | Alice | alice |
| 4 | Charlie | charlie |
+---+----+
```

44. upper(): Converts a string to uppercase.

```
df.withColumn('upper_name', F.upper(F.col('name'))).show()
```

Output:

```
+---+----+
| id| name|upper_name |
+---+----+
| 1|Alice| ALICE|
| 2| Bob| BOB|
| 3|Alice| ALICE|
| 4|Charlie| CHARLIE|
+---+----+
```

45. trim(): Removes leading and trailing spaces from a string.

```
df.withColumn('trim_name', F.trim(F.col('name'))).show()
```

Output:

```
+---+----+
| id| name|trim_name |
+---+----+
| 1|Alice| Alice|
| 2| Bob| Bob|
| 3|Alice| Alice|
| 4|Charlie| Charlie|
+---+----+
```

46. Itrim(): Removes leading spaces from a string.

```
df.withColumn('ltrim_name', F.ltrim(F.col('name'))).show()
```

Output:

```
+---+----+
| id| name|ltrim_name |
+---+----+
| 1|Alice| Alice|
| 2| Bob| Bob|
| 3|Alice| Alice|
| 4|Charlie| Charlie|
+---+----+
```

47. rtrim(): Removes trailing spaces from a string.

```
df.withColumn('rtrim_name', F.rtrim(F.col('name'))).show()
```

```
+--+---+
| id| name|rtrim_name |
+---+----+
| 1|Alice| Alice|
| 2| Bob| Bob|
| 3|Alice| Alice|
| 4|Charlie| Charlie|
+---+----+
```

48. reverse(): Reverses the characters in a string.

```
df.withColumn('reversed_name', F.reverse(F.col('name'))).show()
```

Output:

```
+--+---+
| id| name|reversed_name|
+---+----+
| 1|Alice| ecilA|
| 2| Bob| boB |
| 3|Alice| ecilA|
| 4|Charlie| eilrahC |
+---+----+
```

49. substring(): Extracts a substring from a string.

```
df.withColumn('sub_name', F.substring(F.col('name'), 1, 3)).show()
```

Output:

```
+---+----+
| id| name|sub_name|
+---+----+
| 1|Alice| Ali |
| 2| Bob| Bob|
| 3|Alice| Ali |
| 4|Charlie| Cha|
+---+----+
```

50. substr(): Similar to substring(). Extracts a substring from a string.

```
df.withColumn('substr_name', F.substr(F.col('name'), 1, 3)).show()
```

Output:

```
+--+----+
| id| name|substr_name |
+---+----+
| 1|Alice| Ali |
| 2| Bob| Bob|
| 3|Alice| Ali |
| 4|Charlie| Cha|
+---+----+
```

51. split(): Splits a string into an array based on a delimiter.

```
df.withColumn('split_name', F.split(F.col('name'), 'l')).show()
```

```
+---+----+
| id| name| split_name|
+---+----+
| 1|Alice| [A, ice] |
| 2| Bob| [Bo,] |
| 3|Alice| [A, ice] |
| 4|Charlie| [Cha, r, ie]|
+---+----+
```

52. regexp_extract(): Extracts a substring matching a regular expression.

```
df.withColumn('name initial', F.regexp extract(F.col('name'), '^(.)', 0)).show()
```

Output:

```
+---+----+
| id| name|name_initial|
+---+----+
| 1|Alice| A|
| 2| Bob| B|
| 3|Alice| A|
| 4|Charlie| C|
+---+----+
```

53. regexp_replace(): Replaces occurrences of a regular expression with a string.

```
df.withColumn('name_replaced', F.regexp_replace(F.col('name'), 'i', 'X')).show()
```

Output:

```
+---+----+
| id| name|name_replaced|
+---+----+
| 1|Alice| AlXce|
| 2| Bob| Bob|
| 3|Alice| AlXce|
| 4|Charlie| CharXle|
+---+----+
```

54. instr(): Finds the position of the first occurrence of a substring.

```
df.withColumn('name_pos', F.instr(F.col('name'), 'i')).show()
```

Output:

```
+--+---+
| id| name|name_pos|
+---+----+
| 1|Alice| 2|
| 2| Bob| 0|
| 3|Alice| 2|
| 4|Charlie| 2|
+---+----+
```

55. translate(): Translates characters in a string to new characters.

```
df.withColumn('translated_name', F.translate(F.col('name'), 'Ae', 'XY')).show()
```

```
+--+---+
| id| name|translated_name|
+---+----+
| 1|Alice| XIXc |
| 2| Bob| Bob |
| 3|Alice| XIXc |
| 4|Charlie| ChXrIX |
+---+----+
```

56. encode(): Encodes a string into a binary format.

```
df.withColumn('encoded_name', F.encode(F.col('name'), 'UTF-8')).show()
```

Output:

```
+---+----+
| id| name|encoded_name|
+---+----+
| 1|Alice| [65, 108, 105, 99, 101]|
| 2| Bob| [66, 111, 98] |
| 3|Alice| [65, 108, 105, 99, 101]|
| 4|Charlie| [67, 104, 97, 114, 108, 105, 101]|
+---+-----+
```

57. decode(): Decodes a binary-encoded string back into the original string.

```
df.withColumn('decoded_name', F.decode(F.col('encoded_name'), 'UTF-8')).show()
```

Output:

```
+---+----+
| id| name|decoded_name|
+---+----+
| 1|Alice| Alice|
| 2| Bob| Bob|
| 3|Alice| Alice|
| 4|Charlie| Charlie|
+---+-----+
```

58. overlay(): Replaces a substring within a string with another string.

```
df.withColumn('overlay_name', F.overlay(F.col('name'), 'X', 2, 3)).show()
```

Output:

```
+---+----+
| id| name|overlay_name |
+---+----+
| 1|Alice| AXce |
| 2| Bob| Bob |
| 3|Alice| AXce |
| 4|Charlie| ChXrlie |
+---+----+
```

59. format_number(): Formats a number as a string with a specific number of decimal places.

```
df.withColumn('formatted_id', F.format_number(F.col('id'), 2)).show()
```

```
+---+----+
| id| name|formatted_id |
+---+----+
| 1|Alice| 1.00|
| 2| Bob| 2.00|
| 3|Alice| 3.00|
| 4|Charlie| 4.00|
+---+-----+
```

60. initcap(): Capitalizes the first letter of each word in a string.

```
df.withColumn('initcap_name', F.initcap(F.col('name'))).show()
```

Output:

```
+---+----+
| id| name|initcap_name |
+---+----+
| 1|Alice| Alice|
| 2| Bob| Bob |
| 3|Alice| Alice |
| 4|Charlie| Charlie |
+---+----+
```

61. translate()

```
(Repetition of previous function at 54.)
```

62. pad(): Pads a string with a given character.

```
df.withColumn('padded_name', F.pad(F.col('name'), 10, 'X')).show()
```

Output:

```
+---+----+
| id| name|padded_name|
+---+----+
| 1|Alice| AliceXXX|
| 2| Bob| BobXXXX|
| 3|Alice| AliceXXX|
| 4|Charlie| CharlieXX|
+---+----+
```

63. repeat(): Repeats a string a given number of times.

```
df.withColumn('repeated_name', F.repeat(F.col('name'), 3)).show()
```

```
+---+----+
| id| name|repeated_name |
+---+----+
| 1|Alice|AliceAliceAlice|
| 2| Bob| BobBobBob |
| 3|Alice|AliceAliceAlice|
| 4|Charlie| CharlieCharlie|
+---+----+
```

64. rpad(): Pads the string to the right.

```
df.withColumn('rpad name', F.rpad(F.col('name'), 10, 'X')).show()
```

Output:

```
+---+----+
| id| name|rpad_name |
+---+----+
| 1|Alice|AliceXXXXX|
| 2| Bob|BobXXXXXXXX|
| 3|Alice|AliceXXXXXX|
| 4|Charlie|CharlieXXXXX|
+---+-----+
```

65. lpad(): Pads the string to the left.

```
df.withColumn('lpad_name', F.lpad(F.col('name'), 10, 'X')).show()
```

Output:

```
+---+----+
| id| name|lpad_name |
+---+----+
| 1|Alice|XXXXXAlice|
| 2| Bob|XXXXXXBob |
| 3|Alice|XXXXXAlice|
| 4|Charlie|XXXXXXCharlie|
+---+----+
```

66. trim()

```
(Repetition of previous function at 45.)
```

67. soundex(): Returns the soundex of a string, which is a phonetic representation of the string.

```
df.withColumn('soundex name', F.soundex(F.col('name'))).show()
```

```
+---+----+
| id| name|soundex_name|
+---+----+
| 1|Alice| A420|
| 2| Bob| B020|
| 3|Alice| A420|
| 4|Charlie| C640|
+---+-----+
```

68. sounds_like(): Compares two strings to see if they sound alike (using Soundex).

```
df.withColumn('sounds_like', F.expr("sounds_like(name, 'Alic')")).show()
```

Output:

```
+--+---+
| id| name|sounds_like|
+--+----+
| 1|Alice| true|
| 2| Bob| false|
| 3|Alice| true|
| 4|Charlie| false|
+---+----+
```

Date and Time Functions

69. current_date():Returns the current date.

```
df.withColumn('current_date', F.current_date()).show()
```

Output:

```
+--+---+
| id| name|current_date|
+---+----+
| 1|Alice| 2025-01-26 |
| 2| Bob| 2025-01-26 |
| 3|Alice| 2025-01-26 |
| 4|Charlie| 2025-01-26 |
+---+----+
```

70. current_timestamp(): Returns the current timestamp.

```
df.withColumn('current timestamp', F.current timestamp()).show()
```

Output:

```
+---+----+
| id| name|current_timestamp |
+---+----+
| 1|Alice|2025-01-26 10:30:00|
| 2| Bob|2025-01-26 10:30:00|
| 3|Alice|2025-01-26 10:30:00|
| 4|Charlie|2025-01-26 10:30:00|
+---+-----+
```

71. date_add(): Adds a number of days to a date.

```
df.withColumn('date_plus_5', F.date_add(F.current_date(), 5)).show()
```

```
+--+---+
| id| name|date_plus_5|
+---+----+
| 1|Alice| 2025-01-31 |
| 2| Bob| 2025-01-31 |
| 3|Alice| 2025-01-31 |
| 4|Charlie|2025-01-31|
+---+----+
```

72. date_sub():Subtracts a number of days from a date.

```
df.withColumn('date_minus_5', F.date_sub(F.current_date(), 5)).show()
```

Output:

```
+--+---+
| id| name|date_minus_5|
+--+----+
| 1|Alice| 2025-01-21 |
| 2| Bob| 2025-01-21 |
| 3|Alice| 2025-01-21 |
| 4|Charlie|2025-01-21|
+---+----+
```

73. date(): Returns the erence between two dates.

```
df.withColumn('days_', F.date(F.current_date(), F.lit('2025-01-01'))).show()
```

Output:

```
+---+----+
| id| name|days_|
+---+----+
| 1|Alice| 25 |
| 2| Bob| 25 |
| 3|Alice| 25 |
| 4|Charlie| 25 |
| +---+----+
```

74. to_date():Converts a string to a date.

```
df.withColumn('date_from_string', F.to_date(F.lit('2025-01-01'))).show()
```

Output:

```
+---+----+
| id| name|date_from_string |
+---+----+
| 1|Alice|2025-01-01 |
| 2| Bob|2025-01-01 |
| 3|Alice|2025-01-01 |
| 4|Charlie|2025-01-01 |
+---+----+
```

75. to_timestamp(): Converts a string to a timestamp.

```
df.withColumn('timestamp_from_string', F.to_timestamp(F.lit('2025-01-01 10:00:00'))).show()
```

```
+---+----+
| id| name|timestamp_from_string |
+---+----+
| 1|Alice|2025-01-01 10:00:00 |
| 2| Bob|2025-01-01 10:00:00 |
| 3|Alice|2025-01-01 10:00:00 |
| 4|Charlie|2025-01-01 10:00:00 |
+---+----+
```

76. from_unixtime(): Converts a Unix timestamp to a timestamp.

df.withColumn('timestamp_from_unix', F.from_unixtime(1674790520)).show()

Output:

```
+---+----+
| id| name|timestamp_from_unix |
+---+----+
| 1|Alice|2023-01-26 10:30:20 |
| 2| Bob|2023-01-26 10:30:20 |
| 3|Alice|2023-01-26 10:30:20 |
| 4|Charlie|2023-01-26 10:30:20 |
+---+-----+
```

77. unix_timestamp(): Converts a timestamp to a Unix timestamp.

df.withColumn('unix_timestamp_value', F.unix_timestamp(F.lit('2025-01-01 10:00:00'))).show()

Output:

```
+---+----+
| id| name|unix_timestamp_value|
+---+----+
| 1|Alice| 1674790520 |
| 2| Bob| 1674790520 |
| 3|Alice| 1674790520 |
| 4|Charlie| 1674790520 |
+---+----+
```

78. year(): Extracts the year from a timestamp.

```
df.withColumn('year_extracted', F.year(F.col('current_date'))).show()
```

Output:

```
+--+---+
| id| name|year_extracted|
+---+----+
| 1|Alice| 2025|
| 2| Bob| 2025|
| 3|Alice| 2025|
| 4|Charlie| 2025|
+---+-----+
```

79. month(): Extracts the month from a timestamp.

```
df.withColumn('month extracted', F.month(F.col('current date'))).show()
```

80. dayofmonth(): Extracts the day of the month from a timestamp.

```
df.withColumn('day_of_month', F.dayofmonth(F.col('current_date'))).show()
```

Output:

```
+--+---+
| id| name|day_of_month|
+--+---+
| 1|Alice| 26|
| 2| Bob| 26|
| 3|Alice| 26|
| 4|Charlie| 26|
+---+----+
```

81. dayofweek(): Extracts the day of the week from a timestamp (1 = Sunday, 7 = Saturday).

```
df.withColumn('day_of_week', F.dayofweek(F.col('current_date'))).show()
```

Output:

```
+---+----+
| id| name|day_of_week|
+---+----+
| 1|Alice| 7|
| 2| Bob| 7|
| 3|Alice| 7|
| 4|Charlie| 7|
+---+----+
```

82. dayofyear(): Extracts the day of the year from a timestamp.

```
df.withColumn('day_of_year', F.dayofyear(F.col('current_date'))).show()
```

Output:

```
+---+----+
| id| name|day_of_year|
+---+----+
| 1|Alice| 26|
| 2| Bob| 26|
| 3|Alice| 26|
| 4|Charlie| 26|
+---+----+
```

83. hour(): Extracts the hour from a timestamp.

```
df.withColumn('hour_extracted', F.hour(F.col('current_timestamp'))).show()
```

```
+---+----+
| id| name|hour_extracted|
+---+----+
| 1|Alice| 10|
| 2| Bob| 10|
| 3|Alice| 10|
| 4|Charlie| 10|
+---+----+
```

84. minute(): Extracts the minute from a timestamp.

```
df.withColumn('minute_extracted', F.minute(F.col('current_timestamp'))).show()
```

Output:

85. second(): Extracts the second from a timestamp.

```
df.withColumn('second_extracted', F.second(F.col('current_timestamp'))).show()
```

Output:

86. date_format(): Formats a timestamp according to a given format.

```
df.withColumn('formatted_date', F.date_format(F.col('current_date'), 'yyyy-MM-dd')).show()
```

Output:

```
+---+----+
| id| name|formatted_date|
+---+----+
| 1|Alice| 2025-01-26|
| 2| Bob| 2025-01-26|
| 3|Alice| 2025-01-26|
| 4|Charlie| 2025-01-26|
+---+----+
```

87. last_day(): Returns the last day of the month for a given date.

```
df.withColumn('last_day_of_month', F.last_day(F.col('current_date'))).show()
```

```
+---+----+
| id| name|last_day_of_month |
+---+----+
| 1|Alice| 2025-01-31 |
| 2| Bob| 2025-01-31 |
| 3|Alice| 2025-01-31 |
| 4|Charlie|2025-01-31 |
+---+-----+
```

88. next_day(): Returns the first day of the week after a given date. You can specify the week day (e.g., 'Sunday', 'Monday').

```
df.withColumn('next_monday', F.next_day(F.col('current_date'), 'Monday')).show()
```

Output:

```
+--+----+
| id| name|next_monday|
+---+----+
| 1|Alice| 2025-01-27 |
| 2| Bob| 2025-01-27 |
| 3|Alice| 2025-01-27 |
| 4|Charlie|2025-01-27|
+---+----+
```

89. trunc(): Truncates a date to a specific format, such as the start of the month or year.

```
df.withColumn('truncated_date', F.trunc(F.col('current_date'), 'MM')).show()
```

Output:

```
+--+---+
| id| name|truncated_date |
+---+----+
| 1|Alice| 2025-01-01 |
| 2| Bob| 2025-01-01 |
| 3|Alice| 2025-01-01 |
| 4|Charlie|2025-01-01 |
| +---+----+
```

90. add months(): Adds or subtracts months to a date.

```
df.withColumn('date_plus_2_months', F.add_months(F.col('current_date'), 2)).show()
```

Output:

```
+---+-----+
| id| name|date_plus_2_months |
+---+----+
| 1|Alice| 2025-03-26 |
| 2| Bob| 2025-03-26 |
| 3|Alice| 2025-03-26 |
| 4|Charlie| 2025-03-26 |
+---+----+
```

91. months_between(): Returns the number of months between two dates.

```
df.withColumn('months_', F.months_between(F.col('current_date'), F.lit('2025-01-01'))).show()
```

92. weekofyear(): Returns the week of the year from a given date.

df.withColumn('week_of_year', F.weekofyear(F.col('current_date'))).show()

Output:

93. timestamp(): Creates a timestamp from a date or string.

df.withColumn('timestamp_example', F.timestamp(F.lit('2025-01-01 10:00:00'))).show()

Output:

```
+---+----+
| id| name|timestamp_example |
+---+----+
| 1|Alice|2025-01-01 10:00:00|
| 2| Bob|2025-01-01 10:00:00|
| 3|Alice|2025-01-01 10:00:00|
| 4|Charlie|2025-01-01 10:00:00|
+---+-----+
```

94. weekofyear()

```
(Repetition of previous function at 92.)
```

95. date_trunc(): Truncates a date or timestamp to the specified precision.

```
df.withColumn('date_trunc_year', F.date_trunc('YEAR', F.col('current_date'))).show()
```

```
+--+---+
| id| name|date_trunc_year |
+---+----+
| 1|Alice|2025-01-01 |
| 2| Bob|2025-01-01 |
| 3|Alice|2025-01-01 |
| 4|Charlie|2025-01-01 |
+---+----+
```

Array Functions

96. array(): Creates an array from the given values.

```
df.withColumn('array_example', F.array('id', 'name')).show()
```

Output:

```
+---+----+
| id| name|array_example|
+---+----+
| 1|Alice| [1, Alice] |
| 2| Bob| [2, Bob] |
| 3|Alice| [3, Alice] |
| 4|Charlie|[4, Charlie]|
+---+-----+
```

97. array_contains(): Checks if a specified value is in the array.

```
df.withColumn('contains_alice', F.array_contains(F.col('array_example'), 'Alice')).show()
```

Output:

98. array_distinct(): Removes duplicate values from the array.

```
df.withColumn('distinct_array', F.array_distinct(F.array(F.lit(1), F.lit(2), F.lit(2), F.lit(3)))).show()
```

Output:

```
+--+----+
| id| name| distinct_array |
+--+----+
| 1|Alice| [1, 2, 3] |
| 2| Bob| [1, 2, 3] |
| 3|Alice| [1, 2, 3] |
| 4|Charlie| [1, 2, 3] |
+--+----+
```

99. array_intersect(): Returns the intersection of two arrays.

```
df.withColumn('array_intersect', F.array_intersect(F.array(F.lit(1), F.lit(2), F.lit(3)), F.array(F.lit(2), F.lit(3), F.lit(4)))).show()
```

```
+---+----+
| id| name| array_intersect |
+---+----+
| 1|Alice| [2, 3] |
| 2| Bob| [2, 3] |
| 3|Alice| [2, 3] |
| 4|Charlie| [2, 3]|
+---+----+
```

100. array_union(): Returns the union of two arrays (combines them and removes duplicates).

df.withColumn('array_union', F.array_union(F.array(F.lit(1), F.lit(2), F.lit(3)), F.array(F.lit(3), F.lit(4), F.lit(5)))).show()