

# PySpark

Top 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()
```

Output:

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

### 5.isNull()

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

Output:

```
+---+-----+
| 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()
```

Output:

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



### 13.countDistinct()

```
df.select(F.countDistinct('name')).show()
```

Output:

```
+-----+
|count(DISTINCT name)|
+-----+
|          3|
+-----+
```

### 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()
```

Output:

```
+-----+
|avg(id)|
+-----+
|  2.5  |
+-----+
```

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

### 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:

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

**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:

```
+-----+
|kurtosis(id) |
+-----+
| -1.5      |
+-----+
```

(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(id) |
+-----+
| 0.0      |
+-----+
```

(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()
```

Output:

```
+---+-----+
| 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:

```
+---+-----+-----+
| id| name|percent_rank|
+---+-----+-----+
| 1| Alice| 0.0|
| 2| Bob| 0.33|
| 3| Alice| 0.67|
| 4| Charlie| 1.0|
+---+-----+-----+
```

**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()
```

Output:

```
+---+-----+
| 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. ltrim(): 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()
```

Output:

```
+---+-----+
| 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'), '')).show()
```

Output:

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

**52. `regex_extract()`:** Extracts a substring matching a regular expression.

```
df.withColumn('name_initial', F.regex_extract(F.col('name'), '^(.)', 0)).show()
```

Output:

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

**53. `regex_replace()`:** Replaces occurrences of a regular expression with a string.

```
df.withColumn('name_replaced', F.regex_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()
```

Output:

```
+---+-----+
| 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()
```

Output:

```
+---+-----+
| 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()
```

Output:

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

**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|BobXXXXXXX|
| 3| Alice|AliceXXXXX|
| 4| Charlie|CharlieXXXX|
+---+-----+
```

**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()
```

Output:

```
+---+-----+
| 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()
```

Output:

```
+---+-----+
| 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()
```

Output:

```
+---+-----+
| 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()
```

Output:

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



**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()
```

Output:

```
+---+-----+
| 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:

```
+---+-----+
| id| name|minute_extracted|
+---+-----+
| 1| Alice|      30|
| 2|  Bob|      30|
| 3| Alice|      30|
| 4| Charlie|    30 |
+---+-----+
```

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

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

Output:

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

**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()
```

Output:

```
+---+-----+
| 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()
```

Output:

```
+---+-----+
| id| name|months_|
+---+-----+
| 1| Alice| 0.85 |
| 2| Bob| 0.85 |
| 3| Alice| 0.85 |
| 4| Charlie| 0.85 |
+---+-----+
```

**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:

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

**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()
```

Output:

```
+---+-----+
| 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:

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

**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()
```

Output:

```
+---+-----+
| 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()
```

Output:

```
+---+-----+
| id| name|  array_union  |
+---+-----+
|  1| Alice| [1, 2, 3, 4, 5]|
|  2|  Bob| [1, 2, 3, 4, 5]|
|  3| Alice| [1, 2, 3, 4, 5]|
|  4| Charlie| [1, 2, 3, 4, 5]|
+---+-----+
```