# ACTIONS & BASIC TRANSFORMATIONS



### CONTENT

# Introduction & Presentation ETL Process

Actions & Basic Transformations

Advanced Transformations

Databricks

### GOAL OF THE LEARNING SECTIONS



- Perform Basic Transformation
- Perform Conditional Selection of Rows
- · Carry out basic Data Cleaning

# ACCESSING COLUMNS

Accessing columns with .select()

+			++	+	+
	petalLength	petalWidth	sepalLength	sepalWidth	species
+			+	+	+
	1.4	0.2	5.1	3.5	setosa
	1.4	0.2	4.9	3.0	setosa
	1.3	0.2	4.7	3.2	setosa
	1.5	0.2	4.6	3.1	setosa
	1.4	0.2	5.0	3.6	setosa
	1.7	0.4	5.4	3.9	setosa
	1.4	0.3	4.6	3.4	setosa
	1.5	0.2	5.0	3.4	setosa
	1.4	0.2	4.4	2.9	setosa
	1.5	0.1	4.9	3.1	setosa

| df1.select("petalLength").show(5) +-----+ |petalLength| +-----+ | 1.4| | 1.4| | 1.3| | 1.5| | 1.4|

#### ACCESSING ROWS

• Accessing columns with workaround .collect() then access with print()

1.	0.2	5.1	3.5	setosa
1	1 0.2	4.9	3.0	setosa
1.	0.2	4.7	3.2	setosa
1.	0.2	4.6	3.1	setosa
1.	1 0.2	5.0	3.6	setosa
1.	7 0.4	5.4	3.9	setosa
1.	1 0.3	4.6	3.4	setosa
1.	0.2	5.0	3.4	setosa
1.	1 0.2	4.4	2.9	setosa
1.	0.1	4.9	3.1	setosa

```
[ ] # Returns list of Row objects
  local_df1 = df1.collect()
  print(f"Type of entries: {type(local_df1[0])}\n")
  print(f"Entries: {local_df1[:5]}")
```

Type of entries: <class 'pyspark.sql.types.Row'>

Entries: [Row(petalLength=1.4, petalWidth=0.2, sepalLength=5.1, sepalWidth=3.5, species='setosa'), Row(petalLength=1.4, petalWidth=1.4, petalWidth=1.4, petalWidth=1.4, sepalWidth=1.4, se

# ADDING COLUMNS

Adding columns with .withColumn()

petalLength	petalWidth	sepalLength	sepalWidth	species
1.4	0.2	5.1	3.5	setosa
1.4	0.2	4.9	3.0	setosa
1.3	0.2	4.7	3.2	setosa
1.5	0.2	4.6	3.1	setosa
1.4	0.2	5.0	3.6	setosa
+				

[ ] df\_extraCol = df1.withColumn('newColumn', df1.petalWidth + df1.petalLength)
 df\_extraCol.show(5)

4						·+
į	petalLength	petalWidth	sepalLength	sepalWidth	species	newColumn
+						++
	1.4	0.2	5.1	3.5	setosa	1.5999999999999999
	1.4	0.2	4.9	3.0	setosa	1.5999999999999999
	1.3	0.2	4.7	3.2	setosa	1.5
	1.5	0.2	4.6	3.1	setosa	1.7
	1.4	0.2	5.0	3.6	setosa	1.5999999999999999
4						
1						

# REMOVING COLUMNS

• Removing columns with .drop()

+					
petalLength	petalWidth	sepalLength	sepalWidth	species	petalSum
+		+	+	+	
1.4	0.2	5.1	3.5	setosa	1.5999999999999999
1.4	0.2	4.9	3.0	setosa	1.5999999999999999
1.3	0.2	4.7	3.2	setosa	1.5
1.5	0.2	4.6	3.1	setosa	1.7
1.4	0.2	5.0	3.6	setosa	1.5999999999999999
+		· 	· 		



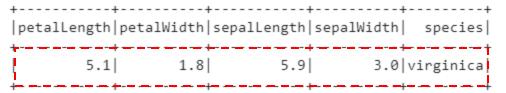
```
[ ] df1 = df_extraCol.drop(df_extraCol.petalSum)
df1.show(5)
```

petalLength				
1.4    1.4    1.3    1.5	0.2 0.2 0.2	4.9 4.7 4.6	3.0 3.2 3.1	setosa   setosa   setosa   setosa   setosa
++		+	+	+

### CONCATENATING DATAFRAMES

Concatenating dataframes with .union()

		+				+	<b>-</b> -				+					+-					+	
nį	gth	ре	eta	alw	idt	h	se	pal	.Le	ng	h	sep	all	Nic	lth	İ		spe	eci	les	İ	
		+				+					+					+-					+	
าเ	ull				nu]	11				nu.	11			nι	111				nι	111		
-	1.4				0.	. 2				5	.1			3	3.5			S	eto	sa		
-	1.4				0.	2				4	9			3	3.0			S	eto	sa		
4	4.1				1.	.3				5	7			2	2.8	Ιv	er	si	col	lor		
4	4.1				1.	3				5	.7			2	2.8	ĺν	er	si	col	lor	ĺ	
(	6.0				2.	5				6	3			3	3.3	ĺ	vi	rg:	ini	ica	ĺ	
	5.1				1.	9				5	8			2	2.7		vi	rg:	ini	ica		/
าเ	ull				nu]	11				nu.	11			nι	111				nι	111		/
_		+-				-+				_	+			_		+		_			+	
																					L	
																					!	
				_		_			_					_			_	-		_		



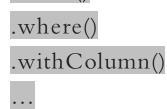
df1 union(df2) show()

ati.union(	atz).snow()			
+  petalLeng	+th petalWidt	h sepalLength		++   species
nu	11   nu]	.1  null	null	null
1	.4  0.	2 5.1	1 3.5	setosa
1	.4  0.	2 4.9	3.0	setosa
4	.1  1.	3 5.7	7 2.8	versicolor
4	.1  1.	3 5.7	7 2.8	versicolor
6	.0  2.	5 6.3	3.3	virginica
5	.1  1.	9  5.8	3 2.7	virginica
nu	11  nu]	.1  null	l  null	null
5	.1  1.	8 5.9	3.0	virginica
+	+	-+	+	++

#### PERFORM CONDITIONAL SELECTION OF ROWS

• Combining aggregate functions, to achieve exact selection: .filter()

petalLength	petalWidth	sepalLength sepa	alWidth	species
4.2   4.2   4.3   3.0   4.1   6.0	1.2 1.3 1.3 1.1 1.3 2.5	5.7  5.7  6.2  5.1  5.7	2.9 2.9 2.5	versicolor versicolor versicolor versicolor versicolor virginica
5.1 5.9 5.6 5.8	1.9 2.1 1.8 2.2	5.8 7.1 6.3 6.5	2.7 3.0 2.9 3.0	virginica



|virginica|

### PERFORM CONDITIONAL SELECTION OF ROWS



```
[ ] (df1.select("species", "petalWidth", "petalLength")
    .where((df1.species == "setosa") & (df1.petalLength > 1.3))
    .withColumn("petalSum", df1.petalWidth + df1.petalLength)
    .dropDuplicates()
    .describe()
    .show(5))
```

+-					+
s	ummary	species	petalWidth	petalLength	petalSum
+-	+				+
	count	16	16	16	16
	mean		•		1.90000000000000001
	stddev	null	0.1414213562373095	0.15916448515084428	0.24221202832779934
	min	setosa	0.1	1.4	1.5
	max	setosa	0.6	1.9	2.3
+-					+

# BASIC DATA CLEANING

- Removing NAs with .dropna()
- Removing duplicates with .dropDuplicates()

