Pridwen Datasets

Alexis Guyot

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Introduction to Pridwen datasets and their functionalities through examples. Internally, Pridwen datasets remain Spark datasets/dataframes. Therefore, they benefit from the same "internal functionalities" as Spark structures: lazy evaluation, optimisation of execution plans, distributed execution, etc. Some provided functions make it easy to switch from a Spark dataset to a Pridwen dataset (and vice versa) at any time.

Additional resources:

- link to the tests source code: https://github.com/AlexisGuyot/Pridwen/blob/master/src/test/scala/DatasetTests.scala;
- link to dataset impelmentation source code: https://github.com/AlexisGuyot/Pridwen/tree/master/src/main/scala/dataset.

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1 Creating a dataset

1.1 Starting from a Spark dataset

```
val spark = SparkSession.builder.master("local").appName("PridwenDataset").getOrCreate
   import spark.implicits._
   case class Base(att1: String, att2: Int, att3: Boolean)
   case class Multivalued(att1: String, att2: Int, att3: Boolean, att4: List[Double])
   case class Nested(att4: String, att5: Int, att6: Boolean, att7: Base)
   val dataset = Seq(
9
      Base("A", 1, true),
Base("B", 2, false),
10
      Base("C", 3, true)
   ).toDS
13
   val datasetM = Seq(
15
      Multivalued("A", 1, true, List(1.1, 1.2, 1.3)),
16
      Multivalued("B", 2, false, List(2.1, 2.2)),
17
      Multivalued("C", 3, true, List())
18
   ).toDS
19
20
   val datasetN = Seq(
21
      Nested("A", 1, true, Base("B", 2, false)),
      Nested("C", 3, true, Base("D", 4, false)),
      Nested("E", 5, true, Base("E", 6, false))
24
   ).toDS
```

1.2 Transforming it into a Pridwen dataset

```
val data = dataset.asModel[Relation]
val dataM = datasetM.asModel[JSON]
val dataN = datasetN.asModel[JSON]

// Syntaxes alternatives
val data = Data(dataset).as[Relation]
val data = Data[Relation](dataset)
```

```
data.describe("MyRelation", true)
```

```
- att2: Int
- att3: Boolean

- att3: Boolean

- att1: Int
- att3: Boolean

- att3: Boolean
```

```
dataM.describe("MyMultiJSON", true)
```

```
======= MyMultiJSON
2
  Model: JSON
  Schema:
  - att1: String
6
  - att2: Int
  - att3: Boolean
8
  - att4: List[Double]
  +---+
11
  |att1|att2| att3| att4|
  +----+
  | A| 1| true|[1.1, 1.2, 1.3]|
  | B| 2|false| [2.1, 2.2]|
15
  | C| 3| true| []|
16
17
  +----+
18
  _____
19
```

```
dataN.describe("MyNestedJSON", true)
```

```
- att3: Boolean
12
13
  +----+
14
  |att4|att5|att6| att7|
15
  +---+
16
17
  | A| 1|true|{B, 2, false}|
  | C| 3|true|{D, 4, false}|
| E| 5|true|{E, 6, false}|
18
19
  +----+
  _____
```

1.3 If needed, go back to a Spark dataset

```
val sparkDS = data.toDS

// Syntaxe alternative
val sparkDS = data.toDS.withSchema[Base]
```

2 Transforming a dataset

2.1 Model change

Internally, this transformation does not modify the structure containing the dataset data. If it is a Spark dataset parameterised by a schema S (as a class), then the internal structure of the new Pridwen dataset obtained after the model change is still a Spark dataset parameterised by the class S.

```
data.as[JSON].describe(true)
```

```
===== Dataset
2
  Model: JSON
  Schema:
  - att1: String
   - att2: Int
   - att3: Boolean
  +---+
10
  |att1|att2| att3|
11
  +----+
12
    A | 1 | true|
13
     B | 2 | false |
14
    C | 3 | true|
15
16
17
   _____
```

2.1.1 Safety brought by typing

```
dataM.as[Relation] // Does not compile because the schema of dataM does not conform to the relational model

data.as[Graph] // Does not compile because the schema of data does not conform to the graph model
```

2.2 Attribute projection

The new dataset schema obtained after projection is automatically inferred from the initial schema of the data, and the names and types of the selected attributes passed as parameters. The underlying dataset becomes a dataframe (Dataset[Row]) since Spark loses track of the schema after applying a projection. The schema of the resulting dataframe (as a StructType¹) is updated to match the schema of the Pridwen dataset.

 $^{^1 \}texttt{https://spark.apache.org/docs/1.5.0/api/java/org/apache/spark/sql/types/StructType.html}$

2.2.1 Simple projection

```
data.select(col('att1)).describe(true)
```

2.2.2 Projecting a nested attribute

```
dataN.select(col('att7) -> col('att2)).describe(true)
```

2.2.3 Projecting multiple attributes

```
dataN.select(col('att4) && col('att7) -> col('att2)).describe(true)
```

```
====== Dataset
   Model: JSON
   Schema:
   - att4: String
   - att2: Int
   +---+
   |att4|att2|
10
   +----+
11
   | A | 2 |
| C | 4 |
| E | 6 |
12
13
14
15
16
```

2.2.4 Projecting with an alias

```
data.select(col('att1).as('test)).describe(true)
```

```
dataN.select(col('att4) && (col('att7) -> col('att1)).as('test)).describe(true)
```

```
------ Dataset

Model: JSON

Schema:
- att4: String
```

```
7 - test: String
8
9 +---+
10 |att4|test|
11 +---+
12 | A | B |
13 | C | D |
14 | E | E |
15 +---+
16
17
```

2.2.5 Safety brought by typing

```
data.select(col('fail)) // Does not compile because the schema of data does not include
    an attribute named fail
```

2.3 Row selection

Internally, this transformation does not modify the structure containing the dataset data. If it is a Spark dataset parameterised by a schema S (as a class), then the internal structure of the new Pridwen dataset obtained after the selection is still a Spark dataset parameterised by the class S.

2.3.1 By comparing the values of some attributes

```
dataN.filter(col('att4) === col('att7) -> col('att1)).describe(true)
```

```
===== Dataset
  Model: JSON
  Schema:
  - att4: String
  - att5: Int
  - att6: Boolean
  - att7:
9
    - att1: String
10
    - att2: Int
     - att3: Boolean
13
  +---+
14
  |att4|att5|att6| att7
15
  +---+
  | E| 5 |true|{E, 6, false}|
17
  +---+
18
19
20
```

```
dataN.select(col('att4) && col('att7) -> col('att1))
.filter(col('att4) === col('att1)).describe(true)
```

2.3.2 By comparing the values of an attribute with a constant value

```
dataN.filter(col('att7) -> col('att1) === v("D")).describe(true)
```

```
===== Dataset
  Model: JSON
  Schema:
  - att4: String
  - att5: Int
  - att6: Boolean
  - att7:
    - att1: String
10
    - att2: Int
11
    - att3: Boolean
12
13
  +---+
14
  |att4|att5|att6| att7
  +---+
  | C | 3 |true|{D, 4, false}|
17
  +---+
18
19
20
```

```
// Other predefined comparison operations
data.filter(col('att2) =!= v(2)).describe(true)
data.filter(col('att2) > v(2)).describe(true)
```

```
data.filter(col('att2) >= v(2)).describe(true)
data.filter(col('att2) < v(2)).describe(true)
data.filter(col('att2) <= v(2)).describe(true)
data.filter(col('att3).isNull).describe(true)
data.filter(col('att3).isNotNull).describe(true)</pre>
```

See section B.1 for the console outputs of the above transformations.

2.3.3 With multiple conditions

```
dataN.filter(
    col('att7) -> col('att1) === v("D") ||
    col('att4) === col('att7) -> col('att1)
    ).describe(true)
```

```
===== Dataset
  Model: JSON
  Schema:
  - att4: String
  - att5: Int
  - att6: Boolean
  - att7:
9
    - att1: String
10
    - att2: Int
11
    - att3: Boolean
12
  +---+
  |att4|att5|att6| att7
15
  +---+
  | C | 3 |true|{D, 4, false}|
17
  | E | 5 |true|{E, 6, false}|
18
  +----+
19
20
```

2.3.4 Comparison based on a UDF

```
// Applied on some attributes
data.filter(
    col('att1) && col('att2),
    (x: String, y: Int) => x == "A" || y == 2
).describe(true)
```

```
1 ====== Dataset
2
```

```
// Applied on all attributes
data.filter((x: String, y: Int, z: Boolean) => x == "A" || y == 2 || z).describe(true)
```

```
===== Dataset
1
  Model: Relation
3
   Schema:
   - att1: String
   - att2: Int
   - att3: Boolean
   +---+
10
   |att1|att2| att3|
11
12
   | A | 1 | true|
13
   | B | 2 | false |
14
   | C | 3 | true|
15
   +----+
16
```

2.3.5 Safety brought by typing

```
dataN.filter(col('att4) === col('att7) -> col('att2)) // Does not compile because
    attributes att4 and att4.att2 do not have the same type

dataN.filter(col('att7) -> col('att1) === v(1)) // Does not compile because attribute
    att7.att1 does not have the same type as value 1

data.filter(col('att1) && col('att2), (x: String, y: Int) => 0) // Does not compile
    because the return type of the UDF is not a boolean
```

```
data.filter(col('att1) && col('att2), (x: String, y: String) => true) // Does not
    compile because the type of attribute att2 (Int) is not String

data.filter(col('att1) && col('att2), (x: String, y: Int, z: Boolean) => y == 2) //
    Does not compile because the UDF has three input parameters whereas only two
    attributes are selected

dataN.filter(col('fail) === col('att7) -> col('att1)) // Does not compile because the
    attribute fail does not exist in the schema of dataN
```

2.4 Adding a new attribute

The new dataset schema obtained after adding the attribute is automatically inferred from the initial schema of the data, the name passed as a parameter and the value or return type of the operator/function used to compute the values of the new attribute. The underlying dataset becomes a dataframe (Dataset[Row]) since Spark loses track of the schema after adding a new attribute. The schema of the resulting dataframe (as a StructType²) is updated to match the schema of the Pridwen dataset.

2.4.1 New attribute with constant values

```
data.add(col('test), 0).keepModel.describe(true)
```

```
===== Dataset
  Model: Relation
   Schema:
   - att1: String
   - att2: Int
   - att3: Boolean
   - test: Int
   +----+
11
   |att1|att2| att3|test|
12
   +---+
13
   | A | 1 | true | 0 |
14
     B | 2 | false | 0 |
15
   | C | 3 | true| 0 |
17
18
19
```

```
// Nested addition
dataN.add(col('att7) -> col('test), 0).keepModel.describe(true)
```

²https://spark.apache.org/docs/1.5.0/api/java/org/apache/spark/sql/types/StructType.html

```
====== Dataset
  Model: JSON
  Schema:
  - att4: String
  - att5: Int
  - att6: Boolean
  - att7:
    - att1: String
10
    - att2: Int
11
    - att3: Boolean
12
     - test: Int
13
14
  +----+
15
  |att4|att5|att6| att7
16
  +----+
17
  | A | 1 |true|{B, 2, false, 0}|
18
  | C | 3 |true|{D, 4, false, 0}|
19
  | E | 5 |true|{E, 6, false, 0}|
20
  +---+
21
22
23
```

```
// Addition requiring a model change
data.add(col('test), List[Int](0)).changeModel[JSON].describe(true)
```

```
===== Dataset
1
2
  Model: JSON
3
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - test: List[Int]
10
  +---+
11
  |att1|att2| att3|test|
12
  +---+
13
  | A | 1 | true| [0]|
14
  | B | 2 | false | [0] |
15
  | C | 3 | true| [0]|
  +---+
```

2.4.2 New attribute with values computed from the values of other attributes

```
data.add(col('test), col('att2) + v(10)).keepModel.describe(true)

// Other operations than + that are predefined: -, *, %, /
```

```
===== Dataset
  Model: Relation
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - test: Int
  +---+
11
  |att1|att2| att3|test|
12
  +---+
13
  | A | 1 | true| 11 |
14
  | B | 2 | false | 12 |
15
  | C | 3 | true | 13 |
16
   +---+
17
18
```

```
dataN.add(col('test), col('att5) + (col('att7) -> col('att2))).keepModel.describe(true)
```

```
===== Dataset
  Model: JSON
  Schema:
  - att4: String
  - att5: Int
  - att6: Boolean
  - att7:
9
    - att1: String
10
    - att2: Int
11
    - att3: Boolean
12
  - test: Int
13
  +---+
15
  |att4|att5|att6| att7 |test|
  +---+
  | A | 1 |true|{B, 2, false}| 3 |
18
  | C | 3 |true|{D, 4, false}| 7 |
19
  | E | 5 |true|{E, 6, false}| 11 |
20
21 +---+---+
```

2.4.3 New attribute with values computed with a UDF

```
// From a single attribute (without nesting)
data.add(col('test), col('att2), (x: Int) => x + 1).keepModel.describe(true)
```

```
===== Dataset
  Model: Relation
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - test: Int
9
10
  +---+
11
12
  |att1|att2| att3|test|
13
  +---+
  | A | 1 | true| 2 |
14
  | B | 2 | false | 3 |
  +---+
17
19
```

```
// From a single attribute (with nesting)
dataN.add(
    col('att7) -> col('test),
    col('att7) -> col('att2),
    (x: Int) => x + 1
) .keepModel.describe(true)
```

```
model: JSON

Schema:
- att4: String
- att5: Int
- att6: Boolean
- att7:
- att1: String
- att2: Int
- att3: Boolean
```

```
- test: Int
13
14
  +---+
15
  |att4|att5|att6| att7 |
16
  +----+
17
  | A | 1 |true|{B, 2, false, 3}|
18
  | C | 3 |true|{D, 4, false, 5}|
19
  | E | 5 |true|{E, 6, false, 7}|
20
  +----+
  _____
```

```
// From multiple attributes
data.add(
    col('test), col('att1) && col('att2),
    (x: String, y: Int) => s"$x$y"
).keepModel.describe(true)
```

```
===== Dataset
2
  Model: Relation
3
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - test: String
  +----+
11
  |att1|att2| att3|test|
12
  +----+
13
  | A | 1 | true| A1 |
14
  | B | 2 | false | B2 |
15
  | C | 3 | true| C3 |
16
  +----+
```

```
// From all attributes
data.add(col('test), "*", (x: String, y: Int, z: Boolean) => 0).keepModel.describe(true)
```

```
1 ======== Dataset
2 
3  Model: Relation
4 
5  Schema:
6  - att1: String
7  - att2: Int
```

```
- att3: Boolean
  - test: Int
9
10
  +---+
11
  |att1|att2| att3|test|
12
   +---+
13
    A | 1 | true | 0 |
14
     B | 2 | false |
                 0 |
    C | 3 | true| 0 |
17
    ---+---+
```

2.4.4 Safety brought by typing

```
data.add(col('test), List[Int]()).keepModel // Does not compile because a multivalued
    attribute cannot be added to a relation

data.add(col('test), col('att1) + v(10)).keepModel // Does not compile because att1 is
    not a numerical attribute

data.add(col('test), col('att1) + col('att2)).keepModel // Same, only att2 is a
    numerical attribute

data.add(col('test), col('fail) + v(10)).keepModel // Does not compile because
    attribute fail does not exist in the schema of data

data.add(col('test), col('att1) && col('att2), (x: String, y: String) =>
    s"$x$y").keepModel // Does not compile because the signature of the UDF passed as
    parameter is erroneous (the type of att2 is Int whereas y is a String)
```

2.5 Deleting an attribute

The new dataset schema obtained after deleting the attribute is automatically inferred from the initial schema of the data, from which the attribute whose name is passed as a parameter is removed. The underlying dataset becomes a dataframe (Dataset[Row]) since Spark loses track of the schema after deleting an attribute. The schema of the resulting dataframe (as a StructType³) is updated to match the schema of the Pridwen dataset.

```
data.drop(col('att1)).describe(true)
```

```
To a section when the section and the section are section as the section ar
```

 $^{^3}$ https://spark.apache.org/docs/1.5.0/api/java/org/apache/spark/sql/types/StructType.html

```
- att2: Int
- att3: Boolean

- att3: Boo
```

2.5.1 Safety brought by typing

```
data.drop(col('fail)) // Does not compile because attribute fail does not exist in the
    schema of data
```

2.6 Attribute update

The new dataset schema obtained after updating the attribute is automatically inferred from the initial schema of the data, the attribute names passed as parameters and, if applicable, the return type of the function used to modify the attribute values. The underlying dataset becomes a dataframe (Dataset[Row]) since Spark loses track of the schema after updating an attribute. The schema of the resulting dataframe (as a StructType⁴) is updated to match the schema of the Pridwen dataset.

2.6.1 Renaming an attribute

```
data.withColumnRenamed('att1, 'test).describe(true)
```

 $^{^4}$ https://spark.apache.org/docs/1.5.0/api/java/org/apache/spark/sql/types/StructType.html

2.6.2 Modifying the values of an attribute

```
// With name change
data.update('att1, 'test, (x: String) => 0).keepModel.describe(true)
```

```
===== Dataset
  Model: Relation
  Schema:
   - test: Int
   - att2: Int
   - att3: Boolean
   +----+
10
11
   |test|att2| att3|
12
   +---+
   | 0 | 1 | true|
13
   | 0 | 2 | false |
   | 0 | 3 | true|
15
   +----+
16
```

```
// Without name change data.update('att1, (x: String) => 0).keepModel.describe(true)
```

```
===== Dataset
2
3
  Model: Relation
  Schema:
  - att1: Int
  - att2: Int
  - att3: Boolean
  +---+
10
  |att1|att2| att3|
11
  +---+
12
  | 0 | 1 | true|
13
  | | 0 | 2 |false|
14
  | 0 | 3 | true|
15
16 +----+
```

2.6.3 Safety brought by typing

```
data.update('fail, 'test, (x: String) => 0).keepModel // Does not compile because
    attribute fail does not exist in the schema of data

data.update('att1, 'test, (x: Int) => 0).keepModel // Does not compile because the
    signature of the UDF passed as parameter is erroneous (the type of att1 is String
    whereas x is an Int)
```

2.7 Sorting a dataset

Internally, this transformation does not modify the structure containing the dataset data. If it is a Spark dataset parameterised by a schema S (as a class), then the internal structure of the new Pridwen dataset obtained after the sorting is still a Spark dataset parameterised by the class S.

```
// Only one ordering criterion. Supported orders: asc (ascending), desc (descending) data.orderBy(col('att1).desc).describe(true)
```

```
====== Dataset
  Model: Relation
   Schema:
   - att1: String
   - att2: Int
   - att3: Boolean
   +----+
10
   |att1|att2| att3|
11
   +----+
12
   | C | 3 | true|
13
    B | 2 |false|
   | A | 1 | true|
```

```
// Multiple ordering criteria data.orderBy(col('att1).desc && col('att2).asc).describe(true)
```

```
----- Dataset
Model: Relation
```

2.7.1 Safety brought by typing

```
data.orderBy(col('fail).desc) // Does not compile because attribute fail does not exist
  in the schema of data
```

2.8 Joins between datasets

The new dataset schema obtained after the join is automatically inferred by merging the original data schemas. The underlying dataset becomes a dataframe (Dataset[Row]) since Spark loses track of the schema after a join. The schema of the resulting dataframe (as a StructType⁵) is updated to match the schema of the Pridwen dataset.

 $^{^5} https://spark.apache.org/docs/1.5.0/api/java/org/apache/spark/sql/types/StructType.html \\$

```
// Inner join with a single condition
data.join(data2, col('att1) === col('truc)).keepLeftModel.describe(true)
```

```
===== Dataset
2
  Model: Relation
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - truc: String
9
  - test: Int
10
11
  +---+
12
  |att1|att2| att3|truc|test|
13
  +---+
14
  | A | 1 | true| A | 1 |
15
  | B | 2 | false | B | 0 |
  | C | 3 | true| C | 1 |
17
  +---+
18
19
20
```

```
// Inner join with multiple conditions preserving the left dataset model (data).
    Supports the same comparison operations as filter

data.join(
    data2,
    col('att1) === col('truc) && col('att2) > col('test)
).keepLeftModel.describe(true)
```

```
Model: Relation

Schema:
- attl: String
- att2: Int
- att3: Boolean
- truc: String
- test: Int
```

```
// Inner join preserving the right dataset model (dataM)
data.join(dataM, col('att1) === col('att1)).keepRightModel.describe(true)
```

```
===== Dataset
  Model: JSON
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - att1: String
9
  - att2: Int
10
  - att3: Boolean
11
  - att4: List[Double]
  +---+
  |att1|att2| att3|att1|att2| att3| att4
  +---+---+
  | A | 1 | true | A | 1 | true | [1.1, 1.2, 1.3] |
  | B | 2 | false | B | 2 | false | [2.1, 2.2] |
18
  | C | 3 | true| C | 3 | true| []
19
  +---+---+----
20
21
```

```
// Inner join requiring a model change
data.join(dataM, col('att1) === col('att1)).changeModel[JSON].describe(true)
```

```
Model: JSON

Model: JSON

Schema:
    - att1: String
    - att2: Int
    - att3: Boolean
    - att1: String
```

```
10 | - att2: Int
  - att3: Boolean
  - att4: List[Double]
12
  +----+
14
  |att1|att2| att3|att1|att2| att3| att4
15
  +---+
16
  | A | 1 | true| A | 1 | true|[1.1, 1.2, 1.3]|
| B | 2 | false| B | 2 | false| [2.1, 2.2] |
17
  | C | 3 | true | C | 3 | true |
                          []
  +----+
  _____
```

```
// With an explicit mention of the join mode. Supported modes: inner, cross, outer,
   full, fullouter, full_outer, left, leftouter, left_outer, right, rightouter,
   right_outer (all aside semi and anti joins, for now)
data.join(data2, col('att1) === col('truc), "left").keepLeftModel.describe(true)
```

```
====== Dataset
  Model: Relation
  Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
  - truc: String
  - test: Int
  +---+
12
  |att1|att2| att3|truc|test|
13
  +---+
14
  | A | 1 | true| A | 1 |
15
  | B | 2 | false | B | 0 |
  | C | 3 | true| C | 1 |
  +---+
18
19
  _____
```

2.8.1 Safety brought by typing

```
data.join(data2, col('att1) === col('truc), "fail").keepLeftModel // Does not compile
  because "fail" is not a supported join mode

data.join(dataM, col('att1) === col('att1)).keepLeftModel // Does not compile because
  the result would be a relation (keepLeftModel) including a multivalued attribute
  (att4 from dataM)
```

```
data.join(data2, col('fail) === col('truc)).keepLeftModel // Does not compile because attribute fail does not exist in the schema of data
```

2.9 Aggregation of a dataset

The new dataset schema obtained after the aggregation is automatically inferred from the initial schema of the data, the attribute names passed as parameters to the groupBy function, and the return types of the operators passed as parameters to the agg function. The underlying dataset becomes a dataframe (Dataset[Row]) since Spark loses track of the schema after applying an aggregation. The schema of the resulting dataframe (as a StructType⁶) is updated to match the schema of the Pridwen dataset.

```
data.groupBy(col('att3))
.agg(col('att3).count && col('att2).avg && col('att2).max)
.keepModel.describe(true)
```

```
====== Dataset
   Model: Relation
   Schema:
   - att3: Boolean
   - max: Int
   - avg: Double
   - count: Long
10
11
   | att3|count|avg|max|
12
   +----+
13
14
   | true| 2 |2.0| 3 |
   |false| 1 |2.0| 2 |
   +----+
16
17
18
```

2.9.1 Safety brought by typing

```
data.groupBy(col('att3)).agg(col('att1).avg).keepModel // Does not compile because att1
   is not a numerical attribute whereas the avg operator can only be applied on this kind of attributes

data.groupBy(col('fail)).agg(col('fail).count).keepModel // Does not compile because attribute fail does not exist in the schema of data
```

 $^{^6 \}texttt{https://spark.apache.org/docs/1.5.0/api/java/org/apache/spark/sql/types/StructType.html}$

A Complete example

```
val datasetN = Seg(
                 SchemaDepartNested("A", 1, true, SchemaDepart("B", 2, false)),
  2
                 SchemaDepartNested("C", 3, true, SchemaDepart("D", 4, false)),
  3
                 SchemaDepartNested("E", 5, true, SchemaDepart("E", 6, false)),
                 SchemaDepartNested("E", 10, true, SchemaDepart("E", 6, false))
         ).toDS
         val dataN = datasetN.asModel[JSON]
         val lds = dataN.add(col('one), List[Int](1)).keepModel
10
                 .add(col('double), col('att5) * v(2)).keepModel
                 .add(col('udf),\ col('double)\ \&\&\ col('one)\ \&\&\ col('att6),\ (x:\ Int,\ y:\ List[Int],\ z:\ List[Int],\ Li
                            Boolean) \Rightarrow if(z) x + y(0) else x - y(0)).keepModel
                 .drop(col('att6))
13
                 .update('one, (x: List[Int]) \Rightarrow x(0)).keepModel
14
                  .withColumnRenamed('att4, 'letter)
15
                  .filter(col('letter) === col('att7) -> col('att1) || col('udf) === v(3))
16
                 .orderBy(col('letter).desc)
                  .select(col('letter) && col('att7) && col('udf).as('res))
         lds.describe(true)
19
         val rds = lds.select(col('att7) -> col('att1) && col('att7) -> col('att2)).as[Relation]
21
         rds.describe(true)
22
         val pridwenDS = lds.join(rds, col('letter) === col('att1), "left").keepLeftModel
24
                 .groupBy(col('letter))
25
                 .agg(col('res).avg)
26
                  .changeModel[Relation]
27
         pridwenDS.describe(true)
         case class FinalSchema(letter: String, avg: Double)
        pridwenDS.toDS.withSchema[FinalSchema].show
```

Applied transformations:

- 1. starting from a Spark dataset;
- 2. transforming it into a Pridwen dataset;
- 3. adding a new attribute (x3);
- 4. deleting an attribute;
- 5. modifying the values of an attribute;
- 6. renaming an attribute;
- 7. rows selection;
- 8. sorting a dataset;
- 9. attribute projection (x2);

- 10. model change;
- 11. join between datasets;
- 12. aggregation of a dataset;
- 13. go back to a Spark dataset.

```
===== Dataset
2
  Model: JSON
3
  Schema:
  - letter: String
6
  - att7:
    - att1: String
     - att2: Int
9
    - att3: Boolean
10
  - res: Int
11
12
13
  |letter| att7 |res|
14
15
  | E |{E, 6, false}| 11|
16
  | E |{E, 6, false}| 21|
17
  | A |{B, 2, false}| 3 |
18
19
  +----+
20
21
  _____
22
  ===== Dataset
23
24
  Model: Relation
25
26
27
  Schema:
   - att1: String
28
  - att2: Int
29
  +---+
  |att1|att2|
32
  +---+
  | E | 6 |
  | E | 6 |
35
  | B | 2 |
36
37
38
39
  _____
40
  ===== Dataset
41
42
  Model: Relation
43
44
  Schema:
45
```

```
- letter: String
  - avg: Double
47
48
  +----+
49
  |letter| avg|
50
  +----+
51
    E |16.0|
52
    A | 3.0|
53
  +---+
  _____
57
  +----+
  |letter| avg|
59
  +----+
    E |16.0|
61
    A | 3.0|
62
  +----+
```

B Other console outputs

B.1 Row(s) selection

```
data.filter(col('att2) =!= v(2)).describe(true)
```

```
data.filter(col('att2) > v(2)).describe(true)
```

```
1 ======= Dataset
2
```

```
Schema:
  - att1: String
  - att2: Int
  - att3: Boolean
9
10
  +----+
  |att1|att2|att3|
11
  +----+
  | C | 3 |true|
  +---+
14
15
16
```

```
data.filter(col('att2) >= v(2)).describe(true)
```

```
===== Dataset
1
  Model: Relation
3
  Schema:
  - att1: String
6
  - att2: Int
  - att3: Boolean
  +---+
  |att1|att2| att3|
11
  +---+
12
  | B | 2 | false |
13
  | C | 3 | true|
14
  +---+
15
16
  _____
17
```

```
data.filter(col('att2) < v(2)).describe(true)</pre>
```

```
12 +----+
  | A | 1 |true|
13
  +---+
14
15
16
   data.filter(col('att2) <= v(2)).describe(true)</pre>
   ===== Dataset
  Model: Relation
  Schema:
  - att1: String
   - att2: Int
   - att3: Boolean
  +---+
10
  |att1|att2| att3|
11
  +---+
12
   | A | 1 | true|
13
   | B | 2 | false |
14
   +----+
15
  data.filter(col('att3).isNull).describe(true)
   ===== Dataset
  Model: Relation
  Schema:
   - att1: String
   - att2: Int
   - att3: Boolean
10
   +----+
11
   |att1|att2|att3|
   +----+
   +----+
  data.filter(col('att3).isNotNull).describe(true)
```

```
1 ===== Dataset
  Model: Relation
  Schema:
  - att1: String
- att2: Int
  - att3: Boolean
  +----+
  |att1|att2| att3|
11
  +----+
  | A | 1 | true|
13
  | | B | 2 | false |
14
  | C | 3 | true|
15
  +----+
16
17
18
```