

ROLLINGSTDEV Function

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Computes the rolling standard deviation of values forward or backward of the current row within the specified column.

- If an input value is missing or null, it is not factored in the computation. For example, for the first row in the dataset, the rolling standard deviation of previous values is undefined.
- The row from which to extract a value is determined by the order in which the rows are organized based on the `order` parameter. If you are working on a randomly generated sample of your dataset, the values that you see for this function might not correspond to the values that are generated on the full dataset during job execution.
- The function takes a column name and two optional integer parameters that determine the window backward and forward of the current row.
 - The default integer parameter values are `-1` and `0`, which computes the rolling function from the current row back to the first row of the dataset.
- This function works with the following transforms:
 - *Window Transform*
 - *Set Transform*
 - *Derive Transform*

For more information on a non-rolling version of this function, see *STDEV Function*.

Basic Usage

Column example:

```
derive type:single value:ROLLINGSTDEV(myCol)
```

Output: Generates a new column containing the rolling standard deviation of all values in the `myCol` column from the first row of the dataset to the current one.

Rows before example:

```
window value:ROLLINGSTDEV(myNumber, 100)
```

Output: Generates the new column, which contains the rolling standard deviation of the current row and the 100 previous row values in the `myNumber` column.

Rows before and after example:

```
window value:ROLLINGSTDEV(myNumber, 3, 2)
```

Output: Generates the new column, which contains the rolling standard deviation of the two previous row values, the current row value, and the two rows after the current one in the `myNumber` column.

Syntax and Arguments

```
window value:ROLLINGSTDEV(col_ref, rowsBefore_integer, rowsAfter_integer) order: order_col  
[group: group_col]
```

Argument	Required?	Data Type	Description
col_ref	Y	string	Name of column whose values are applied to the function
rowsBefore_integer	N	integer	Number of rows before the current one to include in the computation
rowsAfter_integer	N	integer	Number of rows after the current one to include in the computation

For more information on the `order` and `group` parameters, see *Window Transform*.

For more information on syntax standards, see *Language Documentation Syntax Notes*.

col_ref

Name of the column whose values are used to compute the function.

- Multiple columns and wildcards are not supported.

Usage Notes:

Required?	Data Type	Example Value
Yes	String (column reference to Integer or Decimal values)	myColumn

rowsBefore_integer, rowsAfter_integer

Integers representing the number of rows before or after the current one from which to compute the rolling function, including the current row. For example, if the first value is 5, the current row and the four rows after it are used in the computation. Negative values for k compute the rolling average from rows preceding the current one.

- `rowBefore=1` generates the current row value only.
- `rowBefore=-1` uses all rows preceding the current one.
- If `rowsAfter` is not specified, then the value 0 is applied.
- If a `group` parameter is applied, then these parameter values should be no more than the maximum number of rows in the groups.

Usage Notes:

Required?	Data Type	Example Value
No	Integer	4

Examples

Tip: For additional examples, see *Common Tasks*.

Example - Rolling computations for racing splits

This example describes how to use the rolling computational functions:

- **ROLLINGAVERAGE** - computes a rolling average from a window of rows before and after the current row. See *ROLLINGAVERAGE Function*.
- **ROLLINGMIN** - computes a rolling minimum from a window of rows. See *ROLLINGMIN Function*.
- **ROLLINGMAX** - computes a rolling maximum from a window of rows. See *ROLLINGMAX Function*.
- **ROLLINGSTDEV** - computes a rolling standard deviation from a window of rows. See *ROLLINGSTDEV Function*.
- **ROLLINGVAR** - computes a rolling variance from a window of rows. See *ROLLINGVAR Function*.

Source:

In this example, the following data comes from times recorded at regular intervals during a three-lap race around a track. The source data is in cumulative time in seconds (*time_sc*). You can use **ROLLING** and other windowing functions to break down the data into more meaningful metrics.

lap	quarter	time_sc
1	0	0.000
1	1	19.554
1	2	39.785
1	3	60.021
2	0	80.950
2	1	101.785
2	2	121.005
2	3	141.185
3	0	162.008
3	1	181.887
3	2	200.945
3	3	220.856

Transform:

Primary key: Since the quarter information repeats every lap, there is no unique identifier for each row. The following steps create this identifier:

```
settype col: lap,quarter type: 'String'
```

```
derive type:single value: MERGE(['l',lap,'q',quarter]) as: 'splitId'
```

Get split times: Use the following transform to break down the splits for each quarter of the race:

```
derive type:single value: ROUND(time_sc - PREV(time_sc, 1), 3) order: splitId as: 'split_time_sc'
```

Compute rolling computations: You can use the following types of computations to provide rolling metrics on the current and three previous splits:

```
derive type:single value: ROLLINGAVERAGE(split_time_sc, 3) order: splitId as: 'ravg'
```

```
derive type:single value: ROLLINGMAX(split_time_sc, 3) order: splitId as: 'rmax'
```

```
derive type:single value: ROLLINGMIN(split_time_sc, 3) order: splitId as: 'rmin'
```

```
derive type:single value: ROUND(ROLLINGSTDEV(split_time_sc, 3), 3) order: splitId as: 'rstdev'
```

```
derive type:single value: ROUND(ROLLINGVAR(split_time_sc, 3), 3) order: splitId as: 'rvar'
```

Results:

When the above transforms have been completed, the results look like the following:

lap	quarter	splitId	time_sc	split_time_sc	rvar	rstdev	rmin	rmax	ravg
1	0	l1q0	0						
1	1	l1q1	20.096	20.096	0	0	20.096	20.096	20.096
1	2	l1q2	40.53	20.434	0.029	0.169	20.096	20.434	20.265
1	3	l1q3	61.031	20.501	0.031	0.177	20.096	20.501	20.344
2	0	l2q0	81.087	20.056	0.039	0.198	20.056	20.501	20.272
2	1	l2q1	101.383	20.296	0.029	0.17	20.056	20.501	20.322
2	2	l2q2	122.092	20.709	0.059	0.242	20.056	20.709	20.39
2	3	l2q3	141.886	19.794	0.113	0.337	19.794	20.709	20.214
3	0	l3q0	162.581	20.695	0.139	0.373	19.794	20.709	20.373
3	1	l3q1	183.018	20.437	0.138	0.371	19.794	20.709	20.409
3	2	l3q2	203.493	20.475	0.113	0.336	19.794	20.695	20.35
3	3	l3q3	222.893	19.4	0.252	0.502	19.4	20.695	20.252

You can reduce the number of steps by applying a window transform such as the following:

```
window value: window1 = lap,rollingaverage(split_time_sc, 0, 3), rollingmax(split_time_sc, 0, 3),rollingmin(split_time_sc, 0, 3),round(rollingstdev(split_time_sc, 0, 3), 3),round(rollingvar(split_time_sc, 0, 3), 3) group: lap order: lap
```

However, you must rename all of the generated windowX columns.