



# User Manual of IoTDB-Quality

**Author:** Data Quality Group

**Institute:** School of Software, Tsinghua University

**Date:** March 17, 2021

# Contents

<b>1</b>	<b>Get Started</b>	<b>1</b>
1.1	Introduction . . . . .	1
1.2	Comparison . . . . .	1
1.3	Q&A . . . . .	2
<b>2</b>	<b>Data Profiling</b>	<b>3</b>
2.1	Cov . . . . .	3
2.2	Distinct . . . . .	3
2.3	Histogram . . . . .	3
2.4	Integral . . . . .	3
2.5	Mad . . . . .	3
2.6	Max . . . . .	3
2.7	Mean . . . . .	3
2.8	Median . . . . .	3
2.9	Min . . . . .	3
2.10	Mode . . . . .	3
2.11	Percentile . . . . .	3
2.12	Sample . . . . .	3
2.13	Skew . . . . .	3
2.14	Spread . . . . .	3
2.15	Stddev . . . . .	3
<b>3</b>	<b>Data Quality</b>	<b>4</b>
3.1	Completeness . . . . .	4
3.2	Consistency . . . . .	6
3.3	Timeliness . . . . .	8
3.4	Validity . . . . .	11
<b>4</b>	<b>Data Repairing</b>	<b>14</b>
4.1	Fill . . . . .	14
4.2	TimestampRepair . . . . .	14
4.3	ValueRepair . . . . .	14
<b>5</b>	<b>Data Matching</b>	<b>15</b>
5.1	DTW . . . . .	15
5.2	Pearson . . . . .	15

---

5.3	SeriesAlign . . . . .	15
5.4	SeriesSimilarity . . . . .	15
5.5	ValueAlign . . . . .	15
<b>6</b>	<b>Anomaly Detection</b>	<b>16</b>
6.1	KSigma . . . . .	16
6.2	LOF . . . . .	17
6.3	Range . . . . .	17
<b>7</b>	<b>Complex Event Processing</b>	<b>19</b>
7.1	AND . . . . .	19
7.2	EventMatching . . . . .	19
7.3	EventNameRepair . . . . .	19
7.4	EventTag . . . . .	19
7.5	EventTimeRepair . . . . .	19
7.6	MissingEventRecovery . . . . .	19
7.7	SEQ . . . . .	19

# Chapter 1 Get Started

## 1.1 Introduction

### 1.1.1 What is IoTDB-Quality

**Apache IoTDB** (Internet of Things Database) is a data management system for time series data, which can provide users specific services, such as, data collection, storage and analysis.

For applications based on time series data, data quality is vital. **IoTDB-Quality** is User Defined Functions (UDF) about data quality, including data profiling, data quality evaluation and data repairing. It effectively meets the demand for data quality in the industrial field.

### 1.1.2 Quick Start

1. Download the JAR with all dependencies.
2. Copy the JAR package to `ext\udf` under the directory of IoTDB server.
3. Register the UDFs with the following SQL statements in IoTDB:

```
create function completeness as 'cn.edu.thu.dquality.udf.UDTFCompleteness'  
create function consistency as 'cn.edu.thu.dquality.udf.UDTFConsistency'  
create function timeliness as 'cn.edu.thu.dquality.udf.UDFTimeliness'  
create function validity as 'cn.edu.thu.dquality.udf.UDTFValidity'
```

## 1.2 Comparison

### 1.2.1 InfluxDB

**InfluxDB** is a popular time series database. InfluxQL is its query language, some of whose universal functions are related to data profiling. The comparison is shown below. *Native* means this function has been the native function of IoTDB and *Built-in UDF* means this function has been the built-in UDF of IoTDB.

Data profiling functions of IoTDB-Quality	Univeral functions of InfluxQL
<i>Native</i>	COUNT()
<b>Distinct</b>	DISTINCT()
<b>Integral</b>	INTEGRAL()
<b>Mean</b>	MEAN()
<b>Median</b>	MEDIAN()
<b>Mode</b>	MODE()
<b>Spread</b>	SPREAD()
<b>Stddev</b>	STDDEV()
<i>Native</i>	SUM()
<i>Built-in UDF</i>	BOTTOM()
<i>Native</i>	FIRST()
<i>Native</i>	LAST()
<i>Native</i>	MAX()
<i>Native</i>	MIN()
<b>Percentile</b>	PERCENTILE()
<b>Sample</b>	SAMPLE()
<i>Built-in UDF</i>	TOP()
<b>Cov</b>	
<b>Histogram</b>	
<b>Pearson</b>	
<b>Skew</b>	

## 1.3 Q&A

## **Chapter 2 Data Profiling**

**2.1 Cov**

**2.2 Distinct**

**2.3 Histogram**

**2.4 Integral**

**2.5 Mad**

**2.6 Max**

**2.7 Mean**

**2.8 Median**

**2.9 Min**

**2.10 Mode**

**2.11 Percentile**

**2.12 Sample**

**2.13 Skew**

**2.14 Spread**

**2.15 Stddev**

## Chapter 3 Data Quality

### 3.1 Completeness

#### 3.1.1 Usage

This function is used to calculate the completeness of time series. The input series are divided into several continuous and non overlapping windows. The timestamp of the first data point and the completeness of each window will be output.

**Name:** COMPLETENESS

**Input Series:** Only support a single input series. The type is INT32 / INT64 / FLOAT / DOUBLE.

**Parameters:**

- **window**: The number of data points in each window. The number of data points in the last window may be less than it. By default, all input data belongs to the same window.

**Output Series:** Output a single series. The type is DOUBLE. The range of each value is [0,1].

**Note:** Only when the number of data points in the window exceeds 10, the calculation will be performed. Otherwise, the window will be ignored and nothing will be output.

#### 3.1.2 Examples

##### 3.1.2.1 Default Parameters

With default parameters, this function will regard all input data as the same window.

Input series:

Time root.test.d1.s1	
[2020-01-01T00:00:02.000+08:00]	100.0
[2020-01-01T00:00:03.000+08:00]	101.0
[2020-01-01T00:00:04.000+08:00]	102.0
[2020-01-01T00:00:06.000+08:00]	104.0
[2020-01-01T00:00:08.000+08:00]	126.0
[2020-01-01T00:00:10.000+08:00]	108.0
[2020-01-01T00:00:14.000+08:00]	112.0
[2020-01-01T00:00:15.000+08:00]	113.0
[2020-01-01T00:00:16.000+08:00]	114.0
[2020-01-01T00:00:18.000+08:00]	116.0
[2020-01-01T00:00:20.000+08:00]	118.0
[2020-01-01T00:00:22.000+08:00]	120.0
[2020-01-01T00:00:26.000+08:00]	124.0

[2020-01-01T00:00:28.000+08:00]	126.0]
[2020-01-01T00:00:30.000+08:00]	NaN]
+-----+	

SQL for query:

```
select completeness(s1) from root.test.d1 where time <= 2020-01-01 00:00:30
```

Output series:

+-----+	
	Time completeness(root.test.d1.s1)
+-----+	
[2020-01-01T00:00:02.000+08:00]	0.875]
+-----+	

### 3.1.2.2 Specific Window Size

When the window size is given, this function will divide the input data as multiple windows.

Input series:

+-----+	
	Time root.test.d1.s1
+-----+	
[2020-01-01T00:00:02.000+08:00]	100.0]
[2020-01-01T00:00:03.000+08:00]	101.0]
[2020-01-01T00:00:04.000+08:00]	102.0]
[2020-01-01T00:00:06.000+08:00]	104.0]
[2020-01-01T00:00:08.000+08:00]	126.0]
[2020-01-01T00:00:10.000+08:00]	108.0]
[2020-01-01T00:00:14.000+08:00]	112.0]
[2020-01-01T00:00:15.000+08:00]	113.0]
[2020-01-01T00:00:16.000+08:00]	114.0]
[2020-01-01T00:00:18.000+08:00]	116.0]
[2020-01-01T00:00:20.000+08:00]	118.0]
[2020-01-01T00:00:22.000+08:00]	120.0]
[2020-01-01T00:00:26.000+08:00]	124.0]
[2020-01-01T00:00:28.000+08:00]	126.0]
[2020-01-01T00:00:30.000+08:00]	NaN]
[2020-01-01T00:00:32.000+08:00]	130.0]
[2020-01-01T00:00:34.000+08:00]	132.0]
[2020-01-01T00:00:36.000+08:00]	134.0]
[2020-01-01T00:00:38.000+08:00]	136.0]
[2020-01-01T00:00:40.000+08:00]	138.0]
[2020-01-01T00:00:42.000+08:00]	140.0]
[2020-01-01T00:00:44.000+08:00]	142.0]
[2020-01-01T00:00:46.000+08:00]	144.0]
[2020-01-01T00:00:48.000+08:00]	146.0]



[2020-01-01T00:00:50.000+08:00]	148.0
[2020-01-01T00:00:52.000+08:00]	150.0
[2020-01-01T00:00:54.000+08:00]	152.0
[2020-01-01T00:00:56.000+08:00]	154.0
[2020-01-01T00:00:58.000+08:00]	156.0
[2020-01-01T00:01:00.000+08:00]	158.0

SQL for query:

```
select completeness(s1, "window"="15") from root.test.d1 where time <= 2020-01-01 00:01:00
```

Output series:

Time completeness(root.test.d1.s1, "window"="15")
[2020-01-01T00:00:02.000+08:00] 0.875
[2020-01-01T00:00:32.000+08:00] 1.0

## 3.2 Consistency

### 3.2.1 Usage

This function is used to calculate the consistency of time series. The input series are divided into several continuous and non overlapping windows. The timestamp of the first data point and the consistency of each window will be output.

**Name:** CONSISTENCY

**Input Series:** Only support a single input series. The type is INT32 / INT64 / FLOAT / DOUBLE.

**Parameters:**

- **window**: The number of data points in each window. The number of data points in the last window may be less than it. By default, all input data belongs to the same window.

**Output Series:** Output a single series. The type is DOUBLE. The range of each value is [0,1].

**Note:** Only when the number of data points in the window exceeds 10, the calculation will be performed. Otherwise, the window will be ignored and nothing will be output.

### 3.2.2 Examples

#### 3.2.2.1 Default Parameters

With default parameters, this function will regard all input data as the same window.

Input series:

Time	root.test.d1.s1
[2020-01-01T00:00:02.000+08:00]	100.0]
[2020-01-01T00:00:03.000+08:00]	101.0]
[2020-01-01T00:00:04.000+08:00]	102.0]
[2020-01-01T00:00:06.000+08:00]	104.0]
[2020-01-01T00:00:08.000+08:00]	126.0]
[2020-01-01T00:00:10.000+08:00]	108.0]
[2020-01-01T00:00:14.000+08:00]	112.0]
[2020-01-01T00:00:15.000+08:00]	113.0]
[2020-01-01T00:00:16.000+08:00]	114.0]
[2020-01-01T00:00:18.000+08:00]	116.0]
[2020-01-01T00:00:20.000+08:00]	118.0]
[2020-01-01T00:00:22.000+08:00]	120.0]
[2020-01-01T00:00:26.000+08:00]	124.0]
[2020-01-01T00:00:28.000+08:00]	126.0]
[2020-01-01T00:00:30.000+08:00]	NaN]

SQL for query:

```
select consistency(s1) from root.test.d1 where time <= 2020-01-01 00:00:30
```

Output series:

Time	consistency(root.test.d1.s1)
[2020-01-01T00:00:02.000+08:00]	0.9333333333333333]

### 3.2.2.2 Specific Window Size

When the window size is given, this function will divide the input data as multiple windows.

Input series:

Time	root.test.d1.s1
[2020-01-01T00:00:02.000+08:00]	100.0]
[2020-01-01T00:00:03.000+08:00]	101.0]
[2020-01-01T00:00:04.000+08:00]	102.0]
[2020-01-01T00:00:06.000+08:00]	104.0]
[2020-01-01T00:00:08.000+08:00]	126.0]
[2020-01-01T00:00:10.000+08:00]	108.0]
[2020-01-01T00:00:14.000+08:00]	112.0]

[2020-01-01T00:00:15.000+08:00]	113.0
[2020-01-01T00:00:16.000+08:00]	114.0
[2020-01-01T00:00:18.000+08:00]	116.0
[2020-01-01T00:00:20.000+08:00]	118.0
[2020-01-01T00:00:22.000+08:00]	120.0
[2020-01-01T00:00:26.000+08:00]	124.0
[2020-01-01T00:00:28.000+08:00]	126.0
[2020-01-01T00:00:30.000+08:00]	NaN
[2020-01-01T00:00:32.000+08:00]	130.0
[2020-01-01T00:00:34.000+08:00]	132.0
[2020-01-01T00:00:36.000+08:00]	134.0
[2020-01-01T00:00:38.000+08:00]	136.0
[2020-01-01T00:00:40.000+08:00]	138.0
[2020-01-01T00:00:42.000+08:00]	140.0
[2020-01-01T00:00:44.000+08:00]	142.0
[2020-01-01T00:00:46.000+08:00]	144.0
[2020-01-01T00:00:48.000+08:00]	146.0
[2020-01-01T00:00:50.000+08:00]	148.0
[2020-01-01T00:00:52.000+08:00]	150.0
[2020-01-01T00:00:54.000+08:00]	152.0
[2020-01-01T00:00:56.000+08:00]	154.0
[2020-01-01T00:00:58.000+08:00]	156.0
[2020-01-01T00:01:00.000+08:00]	158.0
+-----+-----+	

SQL for query:

```
select consistency(s1,"window"="15") from root.test.d1 where time <= 2020-01-01 00:01:00
```

Output series:

+-----+-----+	
Time consistency(root.test.d1.s1, "window"="15")	
+-----+-----+	
[2020-01-01T00:00:02.000+08:00]	0.9333333333333333
[2020-01-01T00:00:32.000+08:00]	1.0
+-----+-----+	

## 3.3 Timeliness

### 3.3.1 Usage

This function is used to calculate the timeliness of time series. The input series are divided into several continuous and non overlapping windows. The timestamp of the first data point and the timeliness of each window will be output.

**Name:** TIMELINESS

**Input Series:** Only support a single input series. The type is INT32 / INT64 / FLOAT / DOUBLE.

**Parameters:**

- **window** : The number of data points in each window. The number of data points in the last window may be less than it. By default, all input data belongs to the same window.

**Output Series:** Output a single series. The type is DOUBLE. The range of each value is [0,1].

**Note:** Only when the number of data points in the window exceeds 10, the calculation will be performed. Otherwise, the window will be ignored and nothing will be output.

### 3.3.2 Examples

#### 3.3.2.1 Default Parameters

With default parameters, this function will regard all input data as the same window.

Input series:

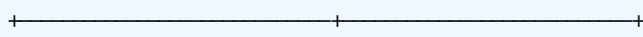
Time root.test.d1.s1	
[2020-01-01T00:00:02.000+08:00]	100.0
[2020-01-01T00:00:03.000+08:00]	101.0
[2020-01-01T00:00:04.000+08:00]	102.0
[2020-01-01T00:00:06.000+08:00]	104.0
[2020-01-01T00:00:08.000+08:00]	126.0
[2020-01-01T00:00:10.000+08:00]	108.0
[2020-01-01T00:00:14.000+08:00]	112.0
[2020-01-01T00:00:15.000+08:00]	113.0
[2020-01-01T00:00:16.000+08:00]	114.0
[2020-01-01T00:00:18.000+08:00]	116.0
[2020-01-01T00:00:20.000+08:00]	118.0
[2020-01-01T00:00:22.000+08:00]	120.0
[2020-01-01T00:00:26.000+08:00]	124.0
[2020-01-01T00:00:28.000+08:00]	126.0
[2020-01-01T00:00:30.000+08:00]	NaN

SQL for query:

```
select timeliness(s1) from root.test.d1 where time <= 2020-01-01 00:00:30
```

Output series:

Time timeliness(root.test.d1.s1)	
[2020-01-01T00:00:02.000+08:00]	0.9333333333333333



### 3.3.2.2 Specific Window Size

When the window size is given, this function will divide the input data as multiple windows.

Input series:

Time	root.test.d1.s1
[2020-01-01T00:00:02.000+08:00]	100.0]
[2020-01-01T00:00:03.000+08:00]	101.0]
[2020-01-01T00:00:04.000+08:00]	102.0]
[2020-01-01T00:00:06.000+08:00]	104.0]
[2020-01-01T00:00:08.000+08:00]	126.0]
[2020-01-01T00:00:10.000+08:00]	108.0]
[2020-01-01T00:00:14.000+08:00]	112.0]
[2020-01-01T00:00:15.000+08:00]	113.0]
[2020-01-01T00:00:16.000+08:00]	114.0]
[2020-01-01T00:00:18.000+08:00]	116.0]
[2020-01-01T00:00:20.000+08:00]	118.0]
[2020-01-01T00:00:22.000+08:00]	120.0]
[2020-01-01T00:00:26.000+08:00]	124.0]
[2020-01-01T00:00:28.000+08:00]	126.0]
[2020-01-01T00:00:30.000+08:00]	NaN]
[2020-01-01T00:00:32.000+08:00]	130.0]
[2020-01-01T00:00:34.000+08:00]	132.0]
[2020-01-01T00:00:36.000+08:00]	134.0]
[2020-01-01T00:00:38.000+08:00]	136.0]
[2020-01-01T00:00:40.000+08:00]	138.0]
[2020-01-01T00:00:42.000+08:00]	140.0]
[2020-01-01T00:00:44.000+08:00]	142.0]
[2020-01-01T00:00:46.000+08:00]	144.0]
[2020-01-01T00:00:48.000+08:00]	146.0]
[2020-01-01T00:00:50.000+08:00]	148.0]
[2020-01-01T00:00:52.000+08:00]	150.0]
[2020-01-01T00:00:54.000+08:00]	152.0]
[2020-01-01T00:00:56.000+08:00]	154.0]
[2020-01-01T00:00:58.000+08:00]	156.0]
[2020-01-01T00:01:00.000+08:00]	158.0]

SQL for query:

```
select timeliness(s1,"window"="15") from root.test.d1 where time <= 2020-01-01 00:01:00
```

Output series:

Time	timeliness(root.test.d1.s1, "window"="15")
[2020-01-01T00:00:02.000+08:00]	0.9333333333333333
[2020-01-01T00:00:32.000+08:00]	1.0

## 3.4 Validity

### 3.4.1 Usage

This function is used to calculate the Validity of time series. The input series are divided into several continuous and non overlapping windows. The timestamp of the first data point and the Validity of each window will be output.

**Name:** VALIDITY

**Input Series:** Only support a single input series. The type is INT32 / INT64 / FLOAT / DOUBLE.

**Parameters:**

- **window**: The number of data points in each window. The number of data points in the last window may be less than it. By default, all input data belongs to the same window.

**Output Series:** Output a single series. The type is DOUBLE. The range of each value is [0,1].

**Note:** Only when the number of data points in the window exceeds 10, the calculation will be performed. Otherwise, the window will be ignored and nothing will be output.

### 3.4.2 Examples

#### 3.4.2.1 Default Parameters

With default parameters, this function will regard all input data as the same window.

Input series:

Time	root.test.d1.s1
[2020-01-01T00:00:02.000+08:00]	100.0
[2020-01-01T00:00:03.000+08:00]	101.0
[2020-01-01T00:00:04.000+08:00]	102.0
[2020-01-01T00:00:06.000+08:00]	104.0
[2020-01-01T00:00:08.000+08:00]	126.0
[2020-01-01T00:00:10.000+08:00]	108.0
[2020-01-01T00:00:14.000+08:00]	112.0
[2020-01-01T00:00:15.000+08:00]	113.0

2020-01-01T00:00:16.000+08:00	114.0
2020-01-01T00:00:18.000+08:00	116.0
2020-01-01T00:00:20.000+08:00	118.0
2020-01-01T00:00:22.000+08:00	120.0
2020-01-01T00:00:26.000+08:00	124.0
2020-01-01T00:00:28.000+08:00	126.0
2020-01-01T00:00:30.000+08:00	NaN
+-----+-----+	

SQL for query:

```
select Validity(s1) from root.test.d1 where time <= 2020-01-01 00:00:30
```

Output series:

+-----+-----+	
	Time  validity(root.test.d1.s1)
+-----+-----+	
2020-01-01T00:00:02.000+08:00	0.8833333333333333
+-----+-----+	

### 3.4.2.2 Specific Window Size

When the window size is given, this function will divide the input data as multiple windows.

Input series:

+-----+-----+	
	Time  root.test.d1.s1
+-----+-----+	
2020-01-01T00:00:02.000+08:00	100.0
2020-01-01T00:00:03.000+08:00	101.0
2020-01-01T00:00:04.000+08:00	102.0
2020-01-01T00:00:06.000+08:00	104.0
2020-01-01T00:00:08.000+08:00	126.0
2020-01-01T00:00:10.000+08:00	108.0
2020-01-01T00:00:14.000+08:00	112.0
2020-01-01T00:00:15.000+08:00	113.0
2020-01-01T00:00:16.000+08:00	114.0
2020-01-01T00:00:18.000+08:00	116.0
2020-01-01T00:00:20.000+08:00	118.0
2020-01-01T00:00:22.000+08:00	120.0
2020-01-01T00:00:26.000+08:00	124.0
2020-01-01T00:00:28.000+08:00	126.0
2020-01-01T00:00:30.000+08:00	NaN
2020-01-01T00:00:32.000+08:00	130.0
2020-01-01T00:00:34.000+08:00	132.0
2020-01-01T00:00:36.000+08:00	134.0
2020-01-01T00:00:38.000+08:00	136.0

[2020-01-01T00:00:40.000+08:00]	138.0
[2020-01-01T00:00:42.000+08:00]	140.0
[2020-01-01T00:00:44.000+08:00]	142.0
[2020-01-01T00:00:46.000+08:00]	144.0
[2020-01-01T00:00:48.000+08:00]	146.0
[2020-01-01T00:00:50.000+08:00]	148.0
[2020-01-01T00:00:52.000+08:00]	150.0
[2020-01-01T00:00:54.000+08:00]	152.0
[2020-01-01T00:00:56.000+08:00]	154.0
[2020-01-01T00:00:58.000+08:00]	156.0
[2020-01-01T00:01:00.000+08:00]	158.0
+-----+-----+	

SQL for query:

```
select Validity(s1,"window"="15") from root.test.d1 where time <= 2020-01-01 00:01:00
```

Output series:

+-----+-----+	
	Time  validity (root.test.d1.s1, "window"="15")
+-----+-----+	
[2020-01-01T00:00:02.000+08:00]	0.8833333333333333
[2020-01-01T00:00:32.000+08:00]	1.0
+-----+-----+	



## **Chapter 4 Data Repairing**

### **4.1 Fill**

### **4.2 TimestampRepair**

### **4.3 ValueRepair**

## **Chapter 5 Data Matching**

### **5.1 DTW**

### **5.2 Pearson**

### **5.3 SeriesAlign**

### **5.4 SeriesSimilarity**

### **5.5 ValueAlign**

## Chapter 6 Anomaly Detection

### 6.1 KSigma

#### 6.1.1 Usage

This function is used to detect distribution anomaly of time series. According to k parameter, the function judges if a input value is an extreme value beyond k-sigma, aka distribution anomaly, and a new time series of anomaly will be output.

**Name:** KSIGMA

**Input Series:** Only support a single input series. The type is INT32 / INT64 / FLOAT / DOUBLE.

- **k**: how many times to multiply on standard deviation to define extreme value.

**Output Series:** Output a single series. The type is DOUBLE.

**Note:** Only when k is larger than 0, the anomaly detection will be performed. Otherwise, nothing will be output.

#### 6.1.2 Examples

##### 6.1.2.1 Assigning k

Input series:

Time	root.test.d1.s1
[2020-01-01T00:00:02.000+08:00]	0.0
[2020-01-01T00:00:03.000+08:00]	50.0
[2020-01-01T00:00:04.000+08:00]	100.0
[2020-01-01T00:00:06.000+08:00]	150.0
[2020-01-01T00:00:08.000+08:00]	200.0
[2020-01-01T00:00:10.000+08:00]	200.0
[2020-01-01T00:00:14.000+08:00]	200.0
[2020-01-01T00:00:15.000+08:00]	200.0
[2020-01-01T00:00:16.000+08:00]	200.0
[2020-01-01T00:00:18.000+08:00]	200.0
[2020-01-01T00:00:20.000+08:00]	150.0
[2020-01-01T00:00:22.000+08:00]	100.0
[2020-01-01T00:00:26.000+08:00]	50.0
[2020-01-01T00:00:28.000+08:00]	0.0
[2020-01-01T00:00:30.000+08:00]	NaN

SQL for query:

```
select ksigma(s1,"k"="1.0") from root.test.d1 where time <= 2020-01-01 00:00:30
```

Output series:

Time	ksigma(root.test.d1.s1,"k"="3.0")
[2020-01-01T00:00:02.000+08:00]	0.0
[2020-01-01T00:00:03.000+08:00]	50.0
[2020-01-01T00:00:26.000+08:00]	50.0
[2020-01-01T00:00:28.000+08:00]	0.0

## 6.2 LOF

## 6.3 Range

### 6.3.1 Usage

This function is used to detect range anomaly of time series. According to upper bound and lower bound parameters, the function judges if a input value is beyond range, aka range anomaly, and a new time series of anomaly will be output.

**Name:** RANGE

**Input Series:** Only support a single input series. The type is INT32 / INT64 / FLOAT / DOUBLE.

- **lower\_bound**: lower bound of range anomaly detection.
- **upper\_bound**: upper bound of range anomaly detection.

**Output Series:** Output a single series. The type is DOUBLE.

**Note:** Only when upper\_bound is larger than lower\_bound, the anomaly detection will be performed. Otherwise, nothing will be output.

### 6.3.2 Examples

#### 6.3.2.1 Assigning Lower and Upper Bound

Input series:

Time	root.test.d1.s1
[2020-01-01T00:00:02.000+08:00]	100.0
[2020-01-01T00:00:03.000+08:00]	101.0
[2020-01-01T00:00:04.000+08:00]	102.0
[2020-01-01T00:00:06.000+08:00]	104.0

[2020-01-01T00:00:08.000+08:00]	126.0]
[2020-01-01T00:00:10.000+08:00]	108.0]
[2020-01-01T00:00:14.000+08:00]	112.0]
[2020-01-01T00:00:15.000+08:00]	113.0]
[2020-01-01T00:00:16.000+08:00]	114.0]
[2020-01-01T00:00:18.000+08:00]	116.0]
[2020-01-01T00:00:20.000+08:00]	118.0]
[2020-01-01T00:00:22.000+08:00]	120.0]
[2020-01-01T00:00:26.000+08:00]	124.0]
[2020-01-01T00:00:28.000+08:00]	126.0]
[2020-01-01T00:00:30.000+08:00]	NaN]

SQL for query:

```
select range(s1,"lower_bound"="101.0","upper_bound"="125.0") from root.test.d1 where time <= 2020-01-01 00:00:30
```

Output series:

Time	range(root.test.d1.s1,"lower_bound"="101.0","upper_bound"="125.0")]
[2020-01-01T00:00:02.000+08:00]	100.0]
[2020-01-01T00:00:28.000+08:00]	126.0]

## **Chapter 7 Complex Event Processing**

### **7.1 AND**

### **7.2 EventMatching**

### **7.3 EventNameRepair**

### **7.4 EventTag**

### **7.5 EventTimeRepair**

### **7.6 MissingEventRecovery**

### **7.7 SEQ**