

# Distributed Time Series Analysis flint Framework For Spark

Larisa Sawyer Two Sigma



## Larisa Sawyer

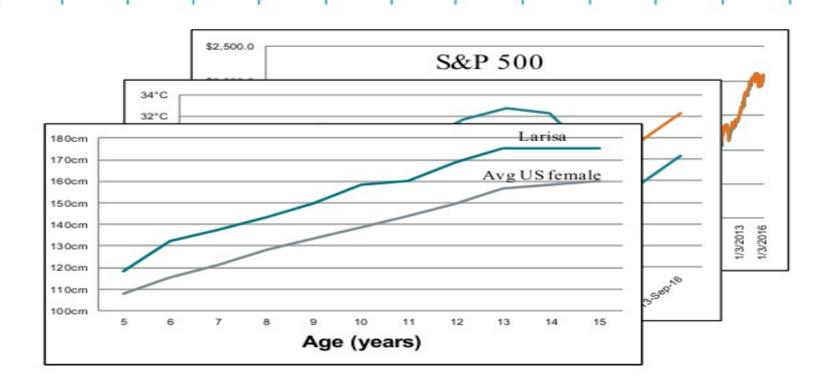






### Time series examples

- Stock market prices
- Temperatures
- Height
- ..







#### What do we do with time series data?

Forecast future values given past observations

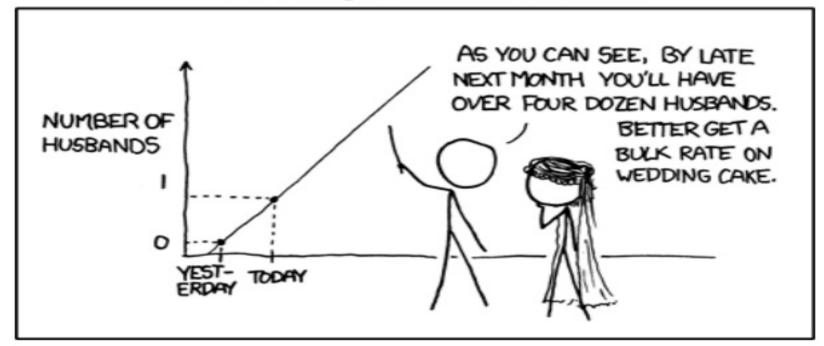
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#### Univariate time series

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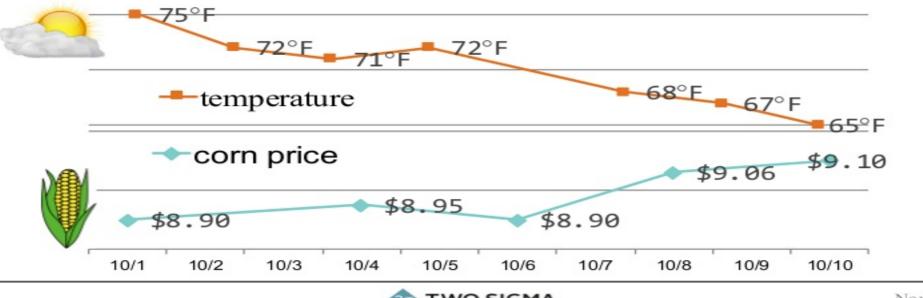
#### MY HOBBY: EXTRAPOLATING



#### Multivariate time series

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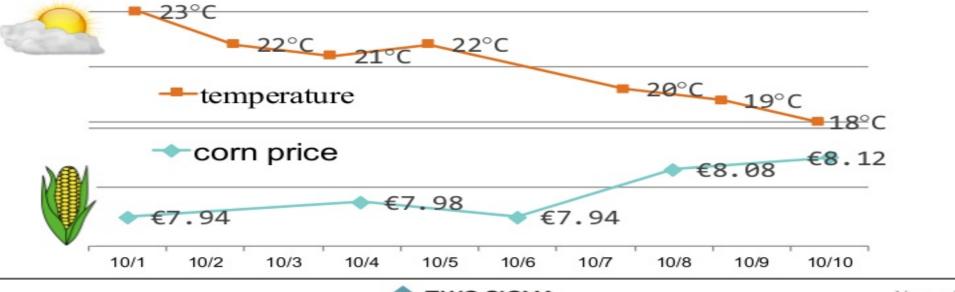
- · We can forecast better by joining multiple time series
- Our framework enables fast distributed temporal join of large scale unaligned time series
- Temporal join is a fundamental operation for time series analysis



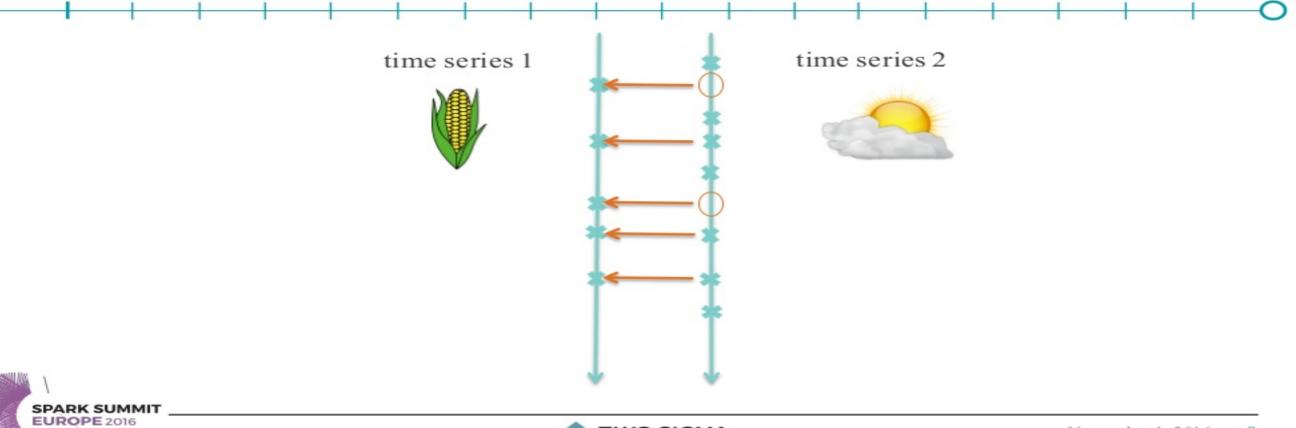
#### Multivariate time series

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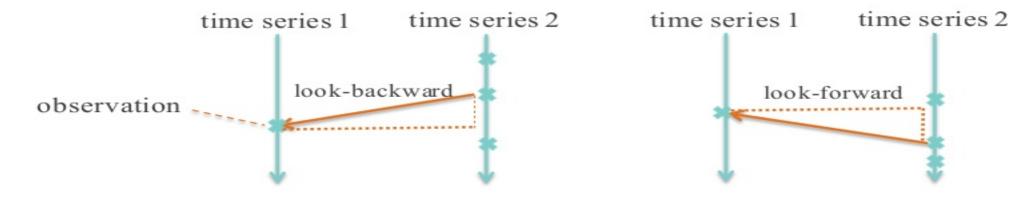


# What is a left join?



### What is temporal join?

- A particular join function defined by a matching criteria over time
- Examples of criteria
  - look-backward
  - look-forward



time	tweets
08:00 AM	8 8 8 8
10:00 AM	<b>6</b>
12:00 PM	000

time	BRK.A
08:00 AM	
11:00 AM	

#### Important Legal Information

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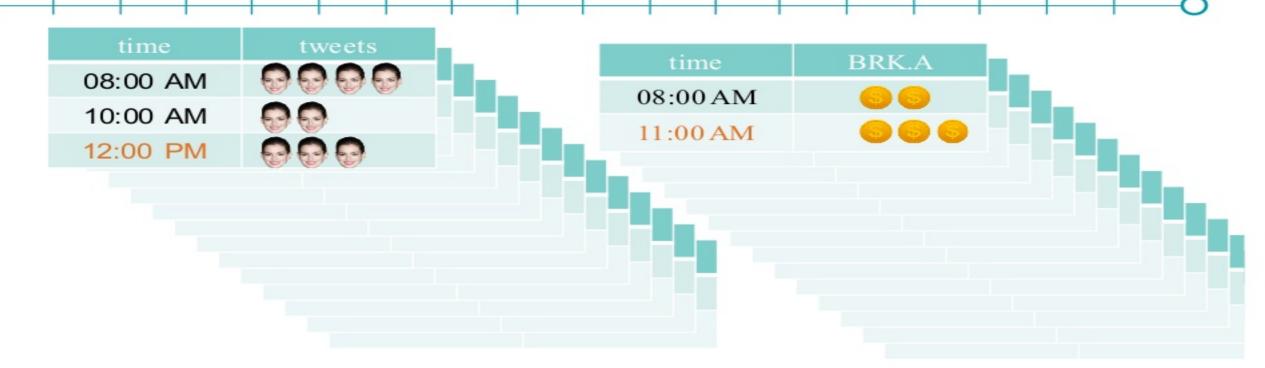
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# Temporal joins in practice





#### **Existing solutions**

- Existing packages don't support temporal join or can't handle large time series
  - Pandas / R / Matlab
    - Limited to single machine
  - Spark
    - · Does scale, but all data is unordered
  - spark-ts
    - Expects univariate time series to fit on single machine
    - Splits by col
    - Supports only snapshot data



#### Flint: A new time series library for Spark

- Goal
  - Provide a collection of functions to manipulate and analyze time series at scale
    - Group, temporal join, summarize, aggregate ...
- How
  - Build a time series aware data structure
    - TimeSeriesRDD extends RDD
  - Optimize using temporal locality
    - Reduce shuffling
    - Reduce memory pressure by streaming



#### What is a TimeSeriesRDD?

- TimeSeriesRDD vs RDD
  - Associate time range on each partition
  - Track partition time-ranges
  - Preserve temporal order



### **RDD**

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Raw Data	
time	temperature
6:00 AM	60°F
6:01 AM	61°F
	••••
7:00 AM	70°F
7:01 AM	71°F
	••••
8:00 AM	80°F
8:01 AM	81°F

#### RDD

(6:00	AM,	60°F)
(6:01	AM,	61°F)
(7:00 (7:01	AM,	70°F) 71°F)
(6:58	AM,	64°F)
(6:59	AM,	65°F)
(7:34	AM,	74°F)
(7:35	AM,	74°F)
(7:58	AM,	76°F)
(7:59	AM,	77°F)
(8:00	AM,	80°F)
(8:01	AM,	81°F)

#### **TimeSeriesRDD**

Raw Data		RDD	TSRDD	
time	temperature	(6:00 AM, <b>60</b> °F) (6:01 AM, <b>61</b> °F)	(6:00 AM, 60°F) (6:01 AM, 61°F)	[06:00]
6:00 AM	60°F	(7:00 AM, <b>70</b> °F)	(6:01 AM, <b>61</b> °F)	07:00 A
6:01 AM	61°F	(7:01 AM, 71°F)		
		(6:58 AM, <b>64</b> °F)	(7:00 AM, <b>70</b> °F)	[07:00
7:00 AM	70°F	(6:59 AM, <b>65</b> °F)	(7:01 AM, <b>71</b> °F)	
7:01 AM	71°F	(7:34 AM, <b>74</b> °F) (7:35 AM, <b>74</b> °F)		
	•••			
8:00 AM	80°F	(7:58 AM, <b>76°F</b> ) (7:59 AM, <b>77°F</b> )	(8:00 AM, <b>80</b> °F) (8:01 AM, <b>81</b> °F)	[8:00 A
8:01 AM	81°F	(8:00 AM, <b>80</b> °F)		∞)
FUDODE sous	((****)	(8:01 AM, <b>81</b> °F)		

TWO SIGMA

November 1, 2016 22

## Group function

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A group function groups rows with exactly the same timestamps

time	city	temperature	
1:00 PM	New York	70°F	group 1
1:00 PM	Brussels	60°F	group
2:00 PM	New York	71°F	group 2
2:00 PM	Brussels	61°F	J gloup 2
3:00 PM	New York	72°F	group 3
3:00 PM	Brussels	62°F	J gloup 3
4:00 PM	New York	73°F	group 4
4:00 PM	Brussels	63°F	J gloup 4

## Group function

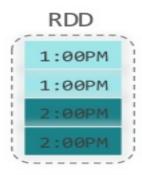
A group function groups rows with nearby timestamps

time	city	temperature	
1:00 PM	New York	70°F	
1:00 PM	Brussels	60°F	group 1
2:00 PM	New York	71°F	group
2:00 PM	Brussels	61°F	
3:00 PM	New York	72°F	
3:00 PM	Brussels	62°F	group 2
4:00 PM	New York	73°F	group 2
4:00 PM	Brussels	63°F	



### Group in Spark

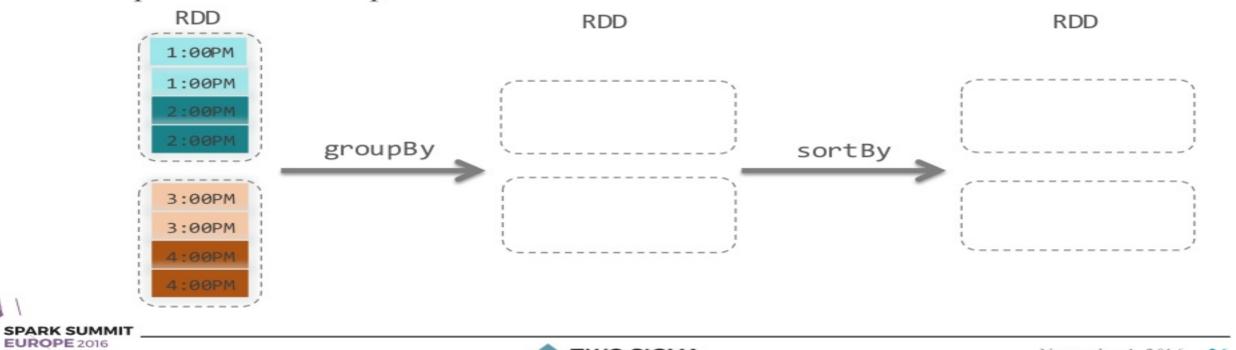
Groups rows with exactly the same timestamps





## Group in Spark

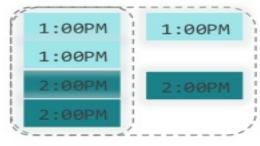
Back to Temporal Order . Datamporal forden is more relized on the workers

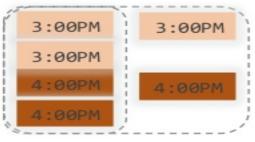


### Group in TimeSeriesRDD

Data is grouped per partition locally as streams

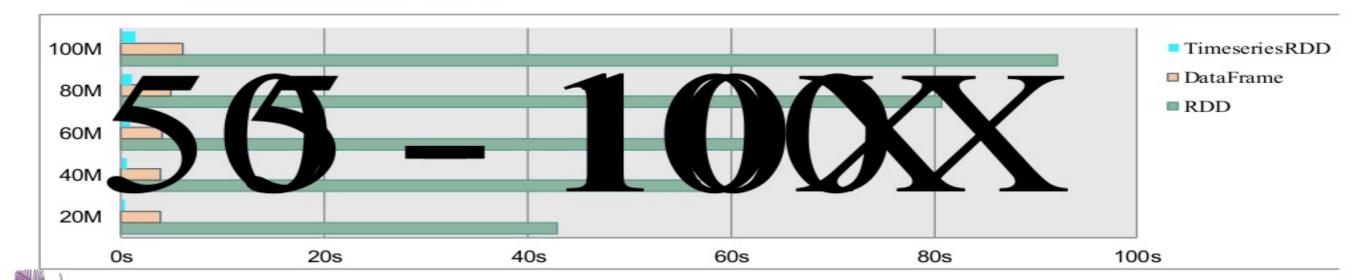




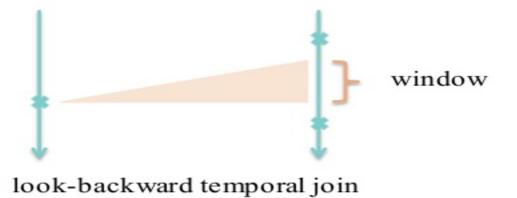


#### Benchmark for group + count

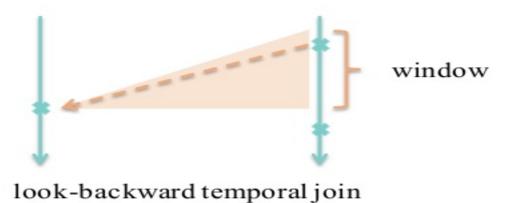
- Running time of count after group
  - 16 executors (10G memory and 4 cores per executor)
  - Data read from HDFS



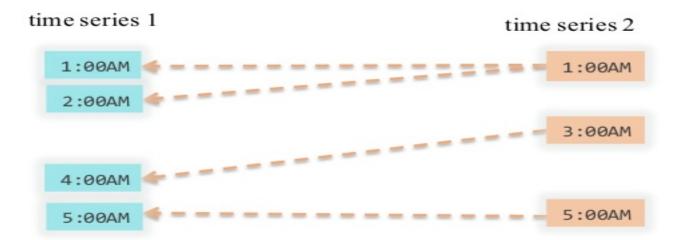
- A temporal join function is defined by a matching criteria over time
- A typical matching criteria has two parameters
  - direction look-backward or look-forward
  - window how much to look-backward or look-forward



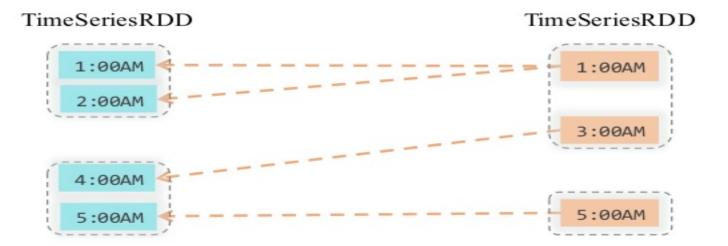
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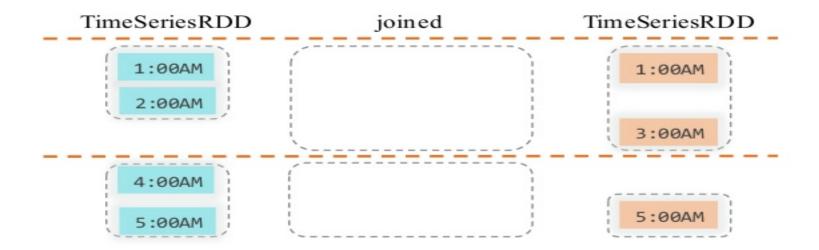
· Temporal join with criteria look-back and window of 1 hour



- Temporal join with criteria look-back and window of 1 hour
  - How do we do temporal join in TimeSeriesRDD?

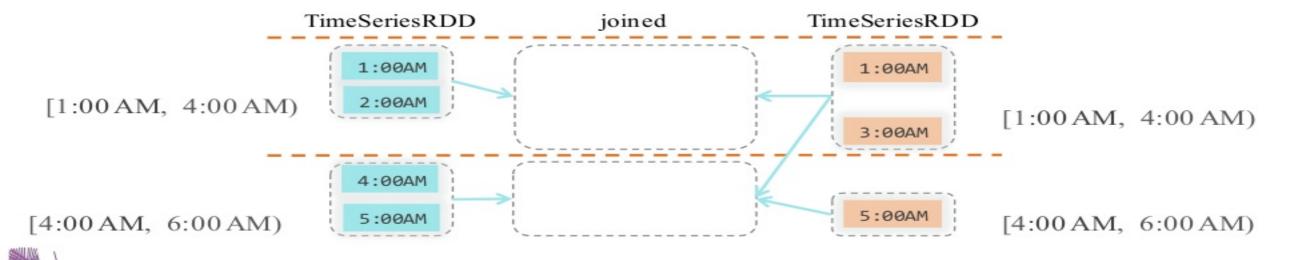


- Temporal join with criteria look-back and window of 1 hour
  - partition time space into disjoint intervals

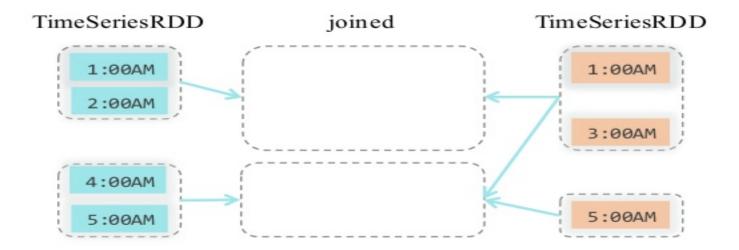


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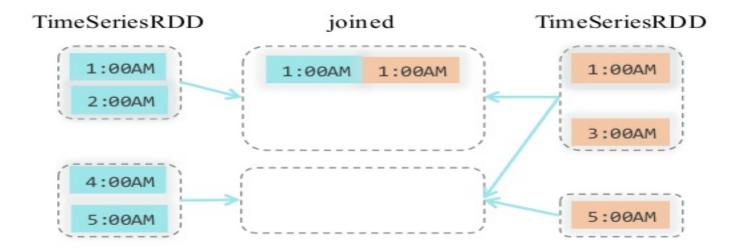
- Temporal join with criteria look-back and window of 1 hour
  - Build dependency graph for the joined TimeSeries RDD



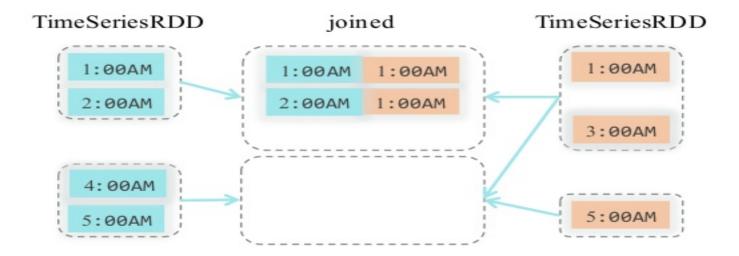
- Temporal join with criteria look-back and window of 1 hour
  - Join data as streams per partition



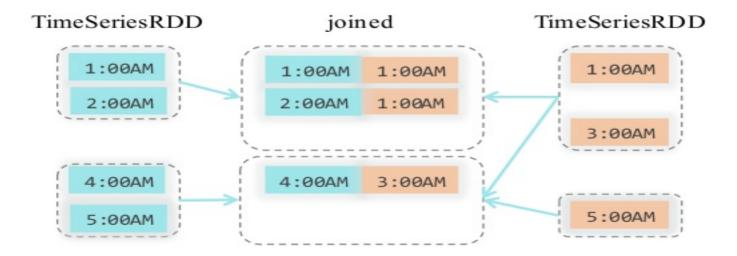
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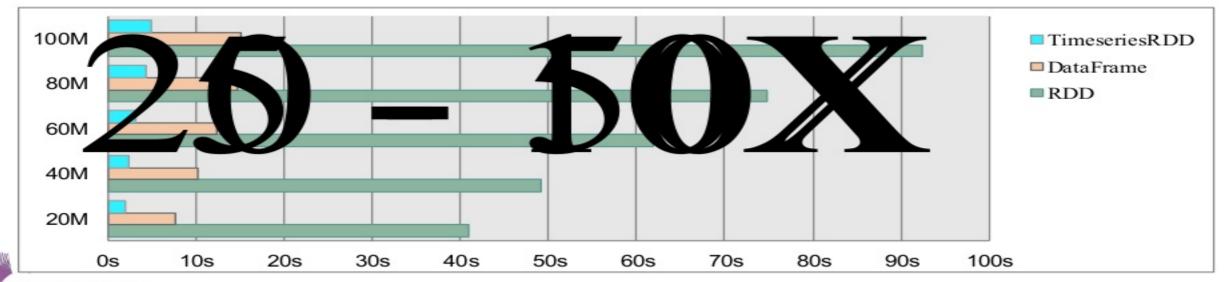


- Temporal join with criteria look-back and window of 1 hour
  - Join data as streams



## Benchmark for temporal join + count

- Running time of count after temporal join
  - 16 executors (10G memory and 4 cores per executor)
  - Data read from HDFS



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#### Functions over TimeSeriesRDD

- Grouping functions
- Temporal joins such as look-forward, look-backward etc.
- Summarizers such as average, variance, z-score etc. over grouping functions





# Open Source

- True!
- https://github.com/twosigma/flint





#### What's next?

- TimeSeriesDataframe / TimeSeriesDataset
- Speed up
  - · Richer APIs
- Python bindings
- Additional summarizers



#### Key contributors

- Christopher Aycock
- Yuri Bogomolov
- Jonathan Coveney
- Li Jin
- David Medina
- Julia Meinwald
- David Palaitis
- Larisa Sawyer
- Leif Walsh

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Wenbo Zhao





## Flint: Time Series For Spark

A library to solve for general time series analysis operations at massive scale Anne Hathaway has nothing to do with Berkshire Hathaway

Check it out in open source, and contribute

https://github.com/twosigma/flint



# THANK YOU.

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