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Spark Streaming

Large-scale near-real-time stream processing

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Requirements

- Scalable to large clusters
- Second-scale latencies
- Simple programming model
- Integrated with batch & interactive processing

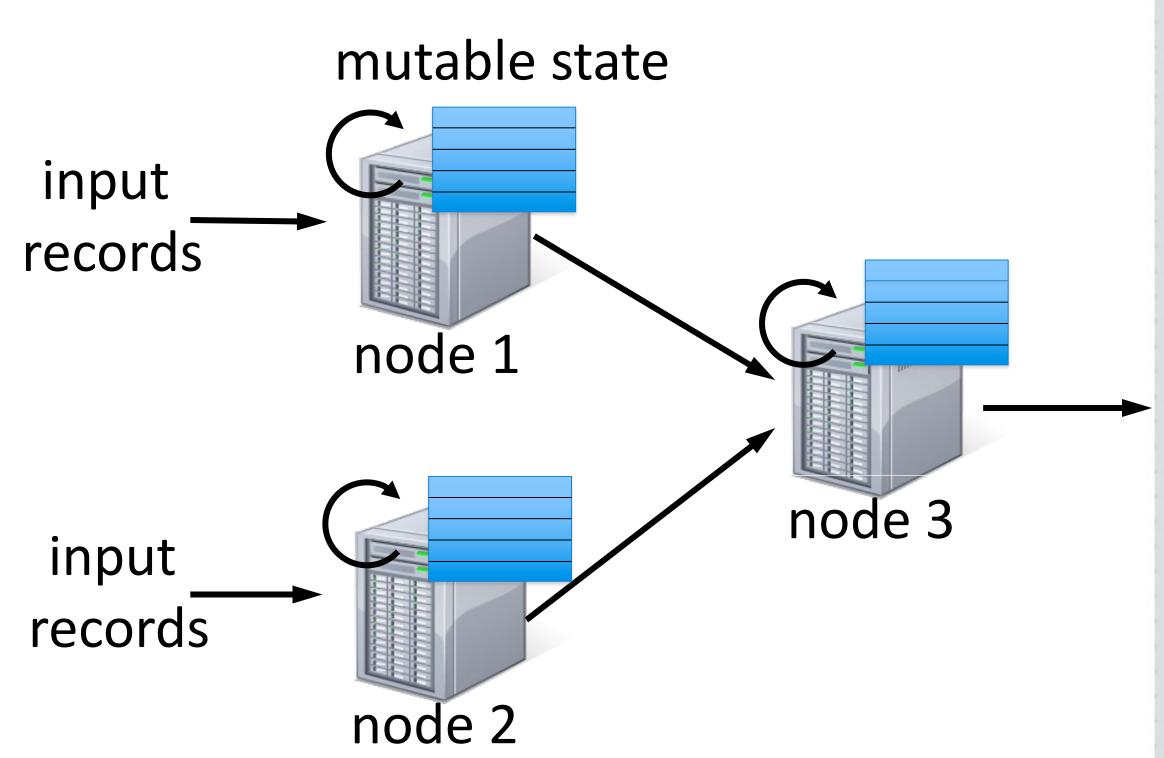


Stateful Stream Processing

- Traditional streaming systems have a eventdriven record-at-a-time processing model
 - Each node has mutable state
 - For each record, update state & send new records

State is lost if node dies!

 Making stateful stream processing be faulttolerant is challenging



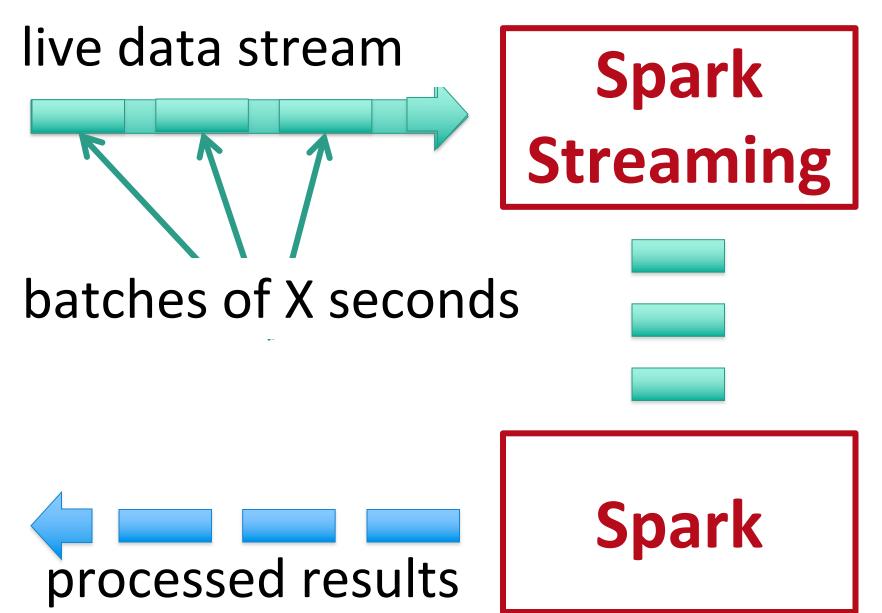
Requirements

- Scalable to large clusters
- Second-scale latencies
- Simple programming model
- Integrated with batch & interactive processing
- Efficient fault-tolerance in stateful computations

Discretized Stream Processing

Run a streaming computation as a series of very small, deterministic batch jobs

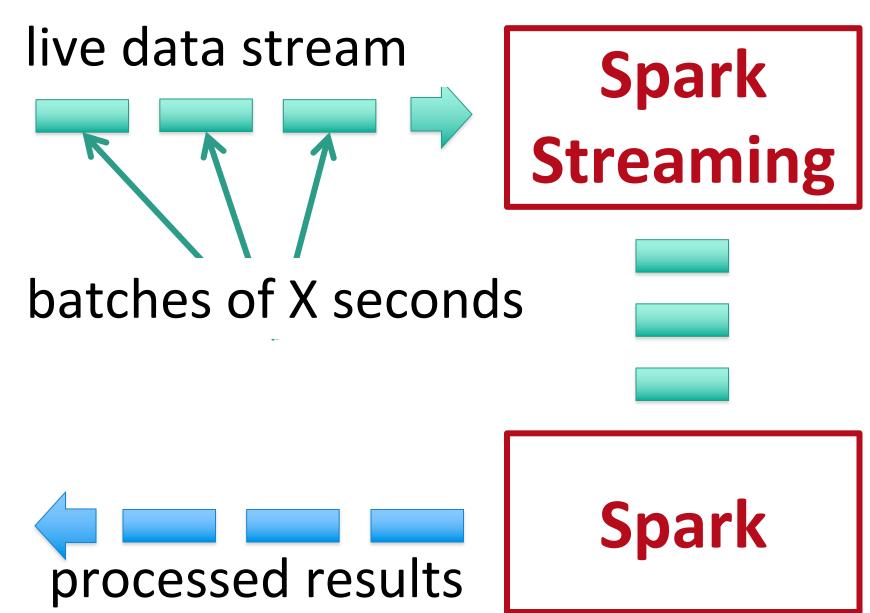
- Chop up the live stream into batches of X seconds
- Spark treats each batch of data as RDDs and processes them using RDD operations
- Finally, the processed results of the RDD operations are returned in batches



Discretized Stream Processing

Run a streaming computation as a series of very small, deterministic batch jobs

- Batch sizes as low as ½ second, latency ~ 1 second
- Potential for combining batch processing and streaming processing in the same system





Example 1 – Get hashtags from Twitter

val tweets = ssc.twitterStream(<Twitter username>, <Twitter password>)

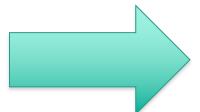
DStream: a sequence of RDD representing a stream of data

Twitter Streaming API

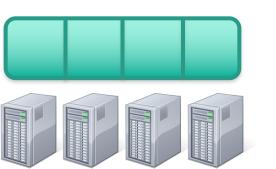


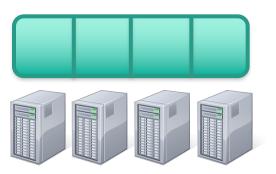
batch @ t+1

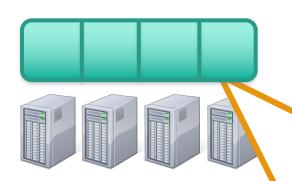
batch @ t+2



tweets DStream





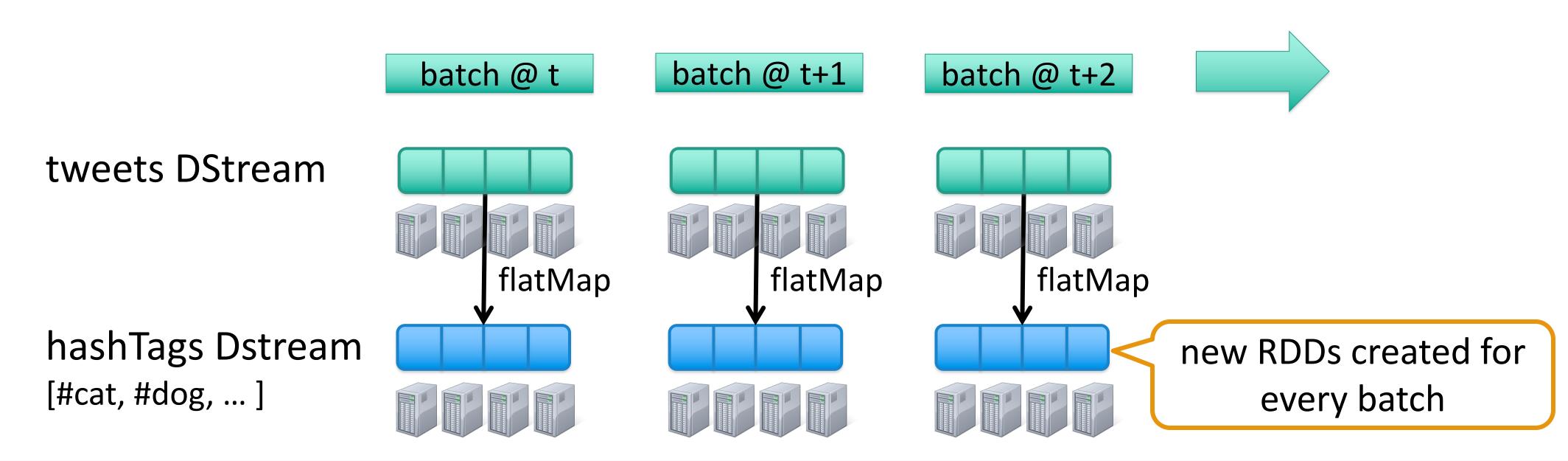


stored in memory as an RDD (immutable, distributed)



Example 1 – Get hashtags from Twitter

```
val tweets = ssc.twitterStream(<Twitter username>, <Twitter password>)
val hashTags = tweets.flatMap (status => getTags(status))
new DStream
transformation: modify data in one Dstream to create another DStream
```





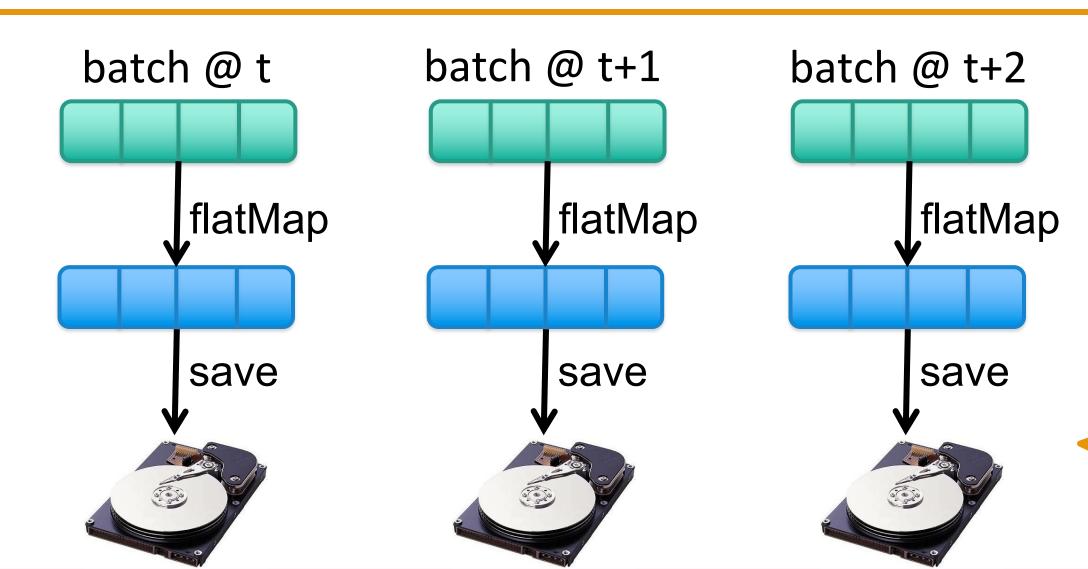
Example 1 – Get hashtags from Twitter

```
val tweets = ssc.twitterStream(<Twitter username>, <Twitter password>)
val hashTags = tweets.flatMap (status => getTags(status))
hashTags.saveAsHadoopFiles("hdfs://...")
```

output operation: to push data to external storage

tweets DStream

hashTags DStream



every batch saved to HDFS



Java Example

Scala

```
val tweets = ssc.twitterStream(<Twitter username>, <Twitter password>)
val hashTags = tweets.flatMap (status => getTags(status))
hashTags.saveAsHadoopFiles("hdfs://...")
```

Java

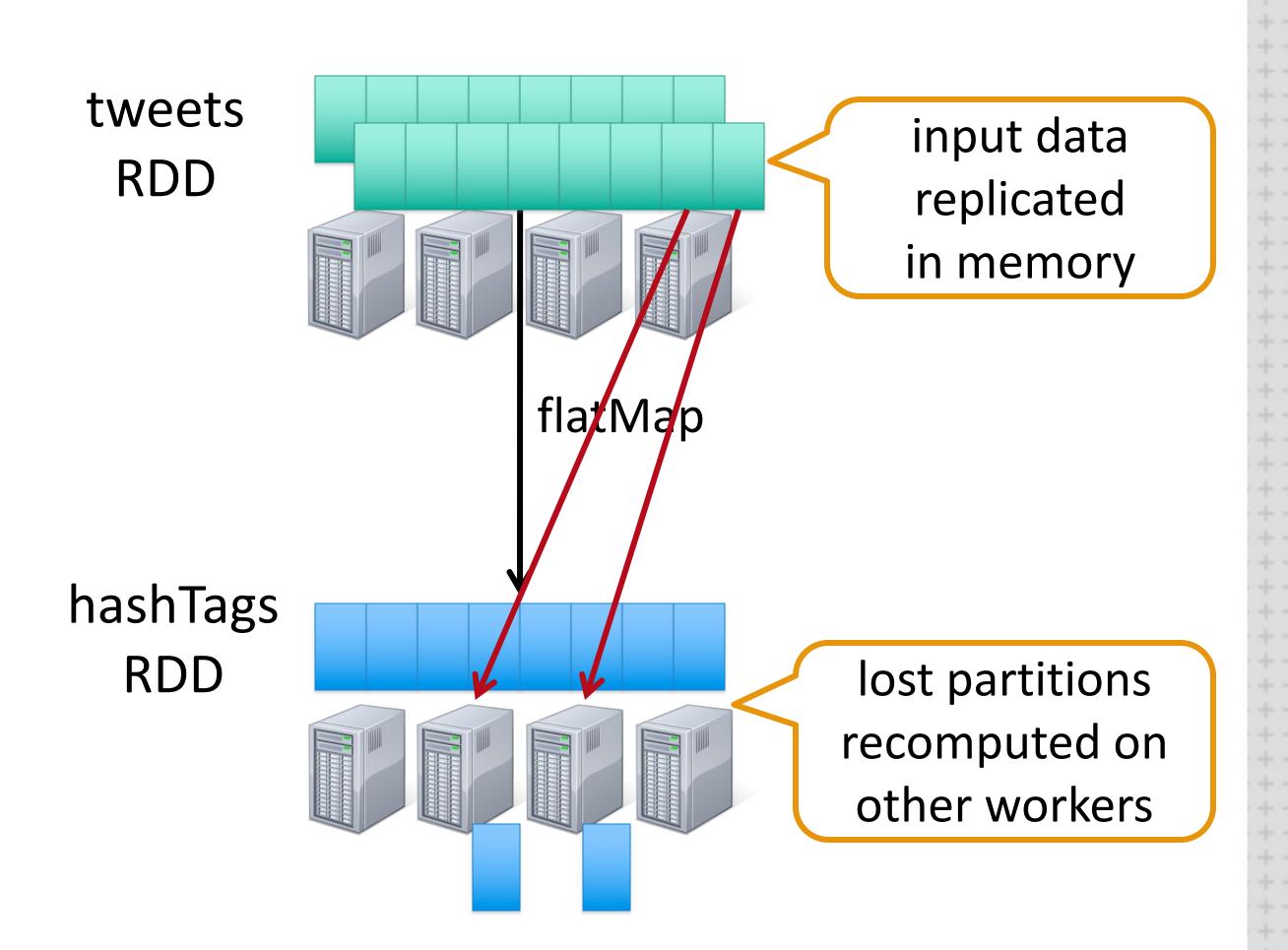


Fault-tolerance

 RDDs are remember the sequence of operations that created it from the original fault-tolerant input data

 Batches of input data are replicated in memory of multiple worker nodes, therefore fault-tolerant

 Data lost due to worker failure, can be recomputed from input data



Key concepts

- DStream sequence of RDDs representing a stream of data
 - Twitter, HDFS, Kafka, Flume, ZeroMQ, Akka Actor, TCP sockets
- Transformations modify data from on DStream to another
 - Standard RDD operations map, countByValue, reduce, join, ...
 - Stateful operations window, countByValueAndWindow, ...
- Output Operations send data to external entity
 - saveAsHadoopFiles saves to HDFS
 - foreach do anything with each batch of results

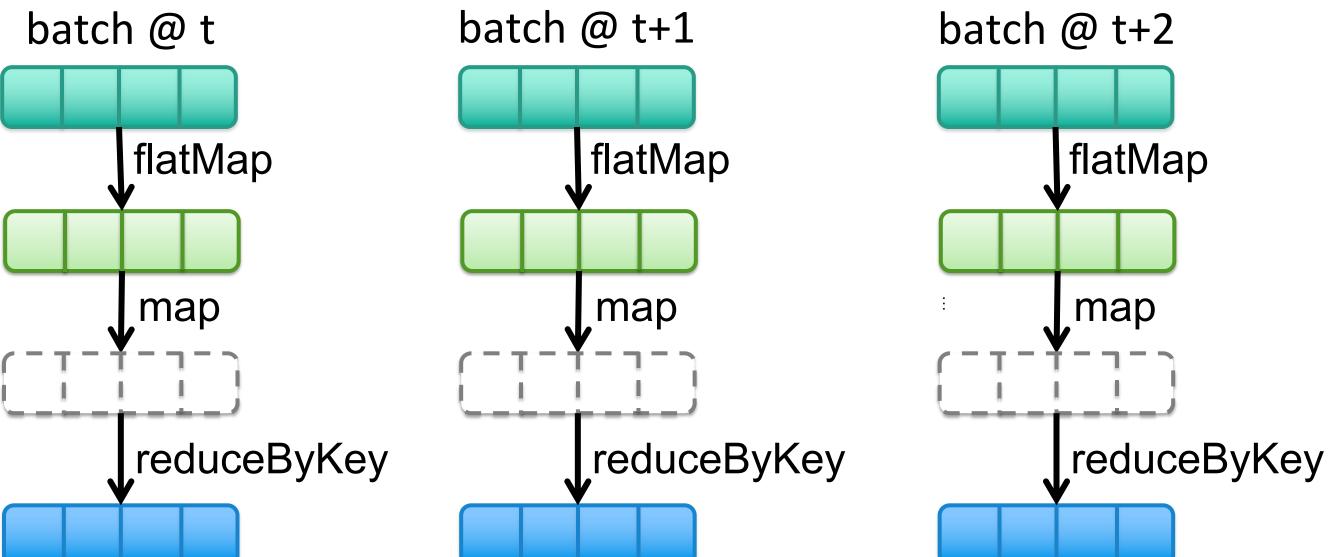


Example 2 – Count the hashtags

tweets

hashTags

tagCounts [(#cat, 10), (#dog, 25), ...]



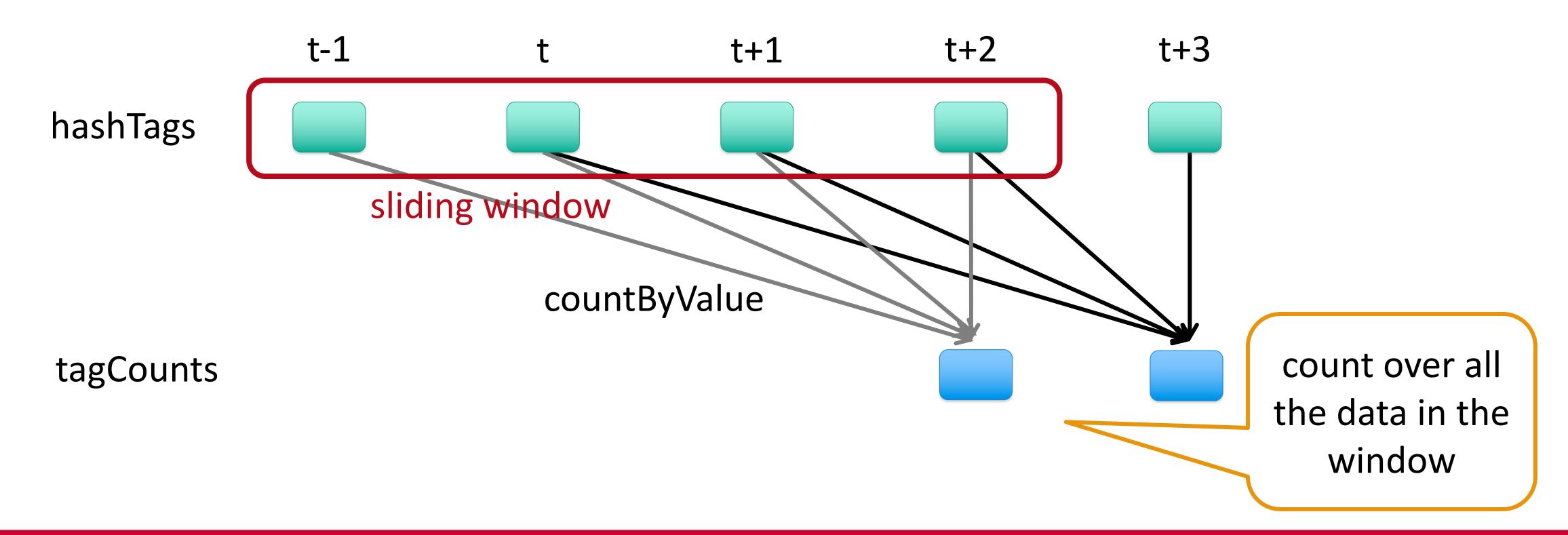


Example 3 – Count the hashtags over last 10 mins



Example 3 – Counting the hashtags over last 10 mins

val tagCounts = hashTags.window(Minutes(10), Seconds(1)).countByValue()





Smart window-based countByValue

val tagCounts = hashtags.countByValueAndWindow(Minutes(10), Seconds(1))

