

# Advanced Streaming Operations

---



**Ahmad Alkilani**

DATA ARCHITECT

@akizl



# Advanced Streaming Operations

**Checkpointing  
Window Operations**

**Stateful Transformations  
Streaming Cardinality Estimation**



# Advanced Streaming Operations

**Checkpointing**

**Window Operations**

**Stateful Transformations**

**Streaming Cardinality Estimation**

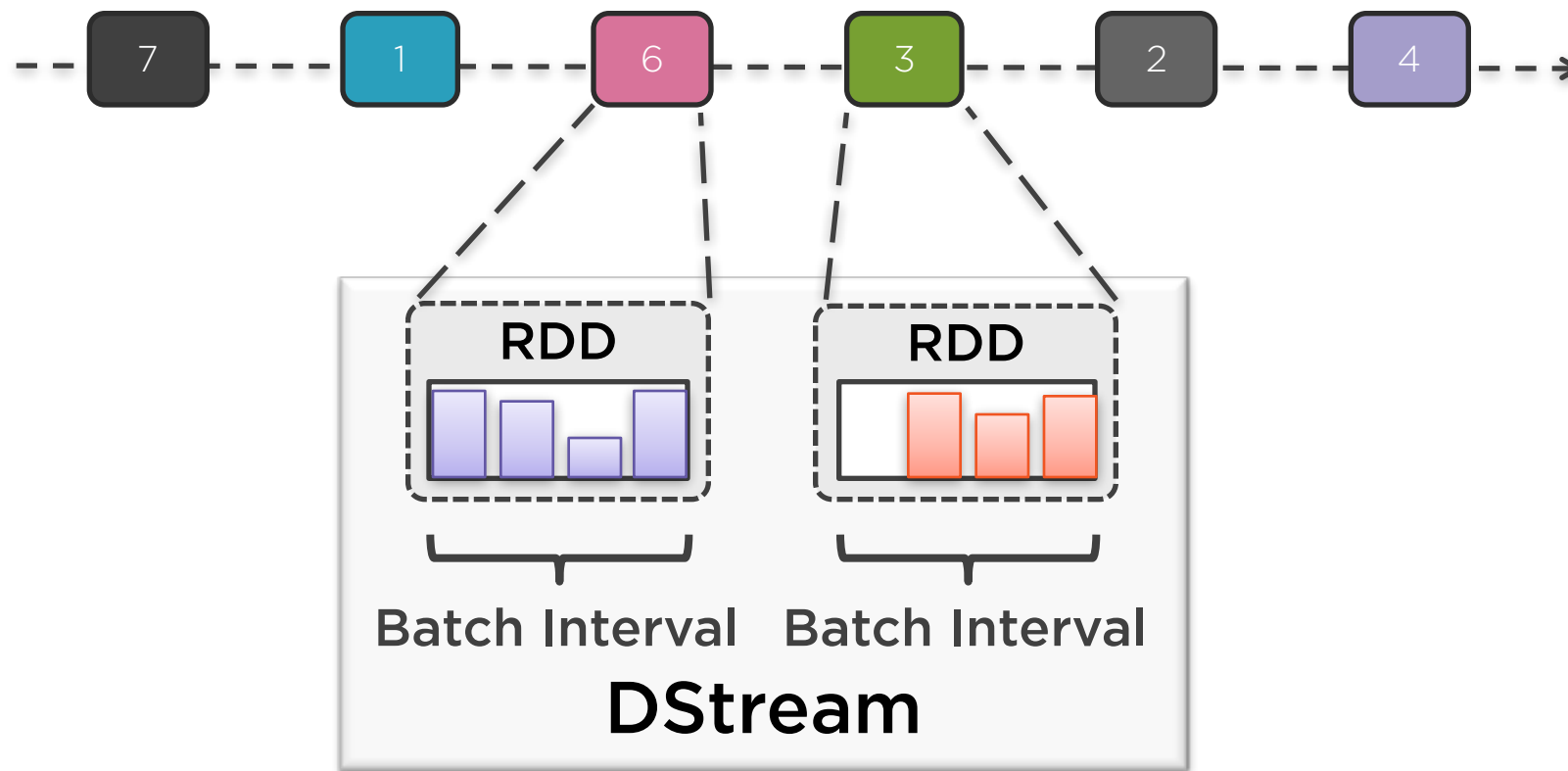


# Checkpointing

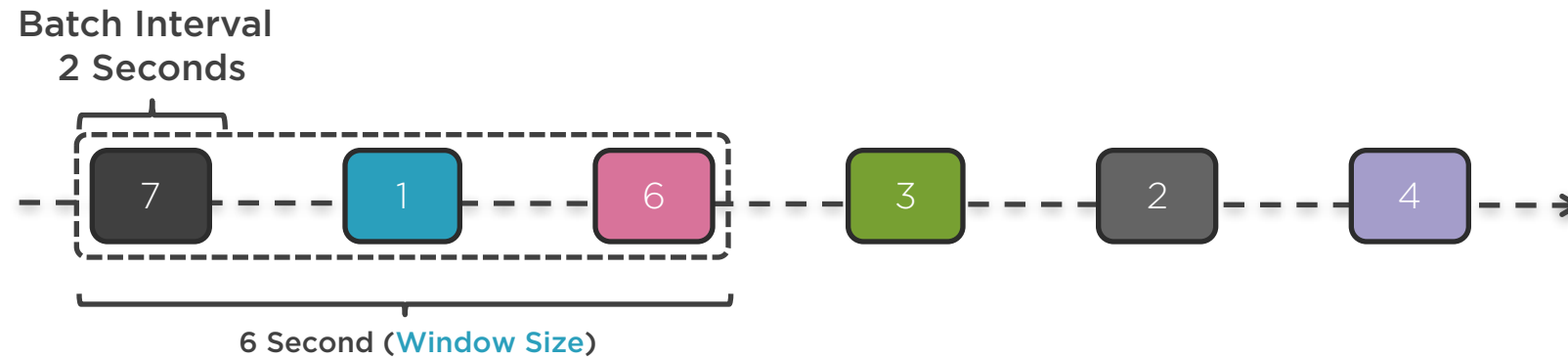
```
def getSQLContext(sc: SparkContext) = {  
  val sqlContext = SQLContext.getOrCreate(sc)  
  sqlContext  
}  
  
def getStreamingContext(streamingApp : (SparkContext, Duration) => StreamingContext, sc : SparkContext, batchDuration: Duration) = {  
  val creatingFunc = () => streamingApp(sc, batchDuration)  
  val ssc = sc.getCheckpointDir match {  
    case Some(checkpointDir) => StreamingContext.getActiveOrCreate(checkpointDir, creatingFunc, sc.hadoopConfiguration, createOnError = true)  
    case None => StreamingContext.getActiveOrCreate(creatingFunc)  
  }  
  sc.getCheckpointDir.foreach( cp => ssc.checkpoint(cp))  
  ssc  
}
```



# Window Operations



# Window Operations



To create a window of a stream of data in Spark define:

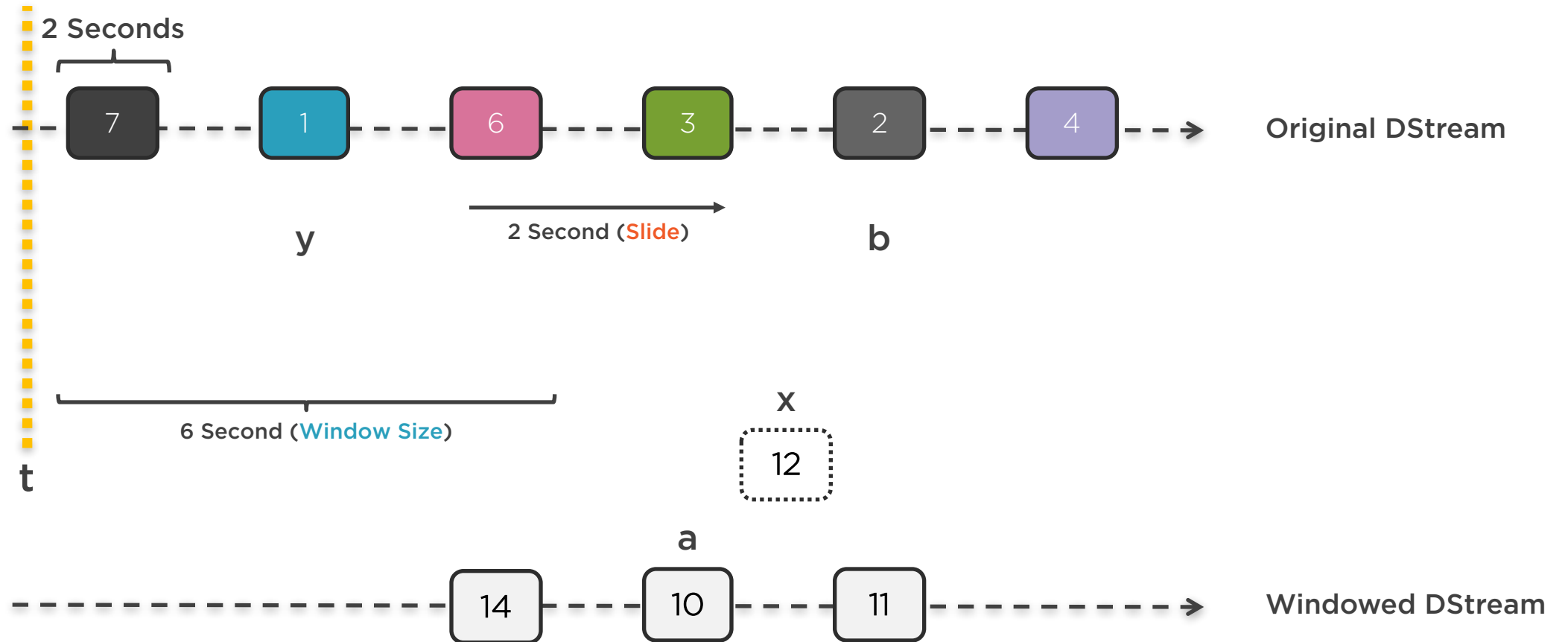
- Window Size
- Slide Interval

Must be multiples of batch interval

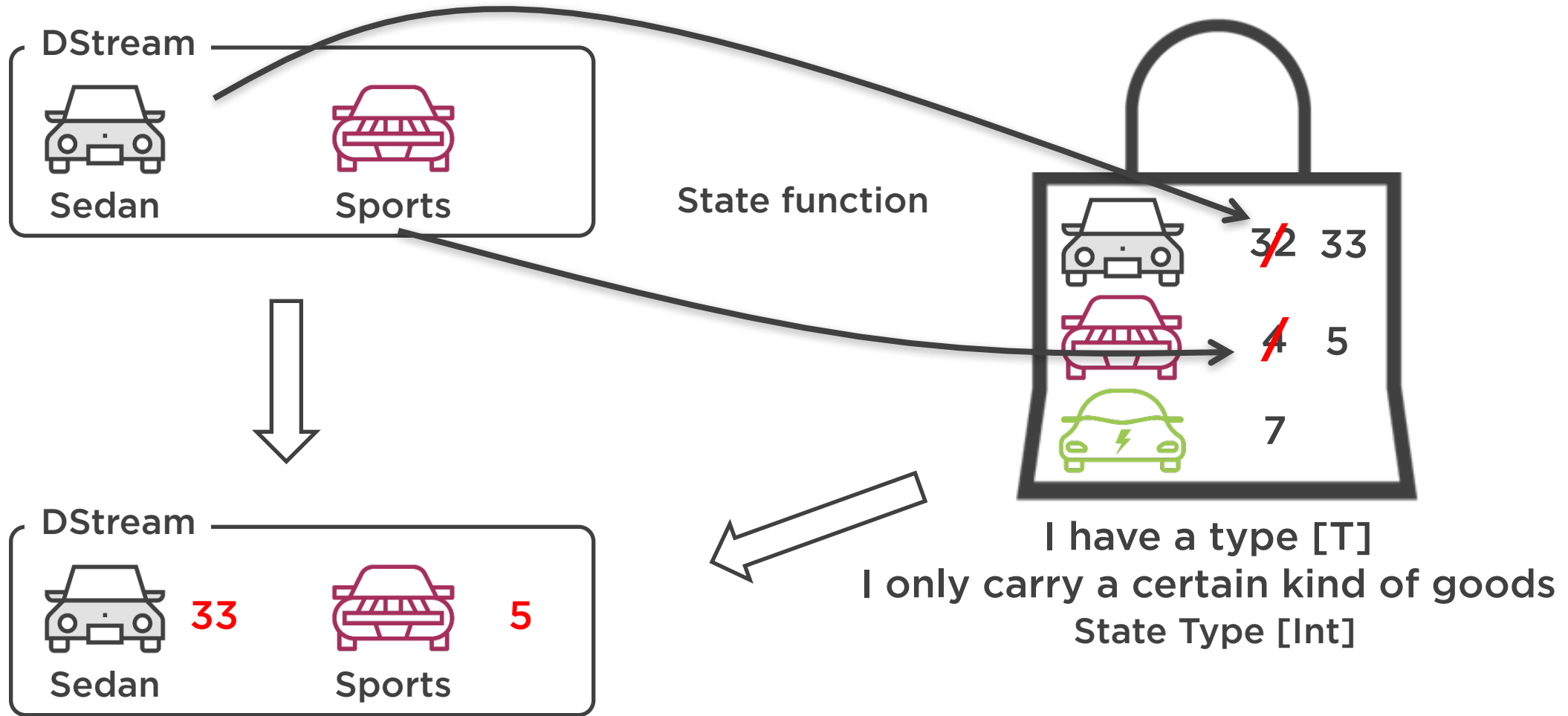


# Window Operations

```
clicksDStream.windowByWindow((a, b) => a + b, (x, y) => x), y, Seconds(20), Seconds(2))
```



# Stateful Transformations





# Stateful Transformations

- `updateStateByKey`
- `mapWithState`

```
activityStream.transform( rdd => {  
  val df = rdd.toDF()  
  df.registerTempTable("activity")  
  val activityByProduct = sqlContext.sql("""SELECT
```

```
    sum(case when action = 'add_to_cart' then 1 else 0 end) as add_to_cart_count,  
    sum(case when action = 'page_view' then 1 else 0 end) as page_view_count  
  from activity  
  group by product, timestamp_hour """)  
  activityByProduct  
    .map { r => ((r.getString(0), r.getLong(1)),  
      ActivityByProduct(r.getString(0), r.getLong(1), r.getLong(2), r.getLong(3), r.getLong(4))  
    ) }  
} )
```



# Stateful Transformations

- **updateStateByKey**
- mapWithState

```
activityByProduct
  .map { r => (r.getString(0), r.getLong(1)),
    ActivityByProduct(r.getString(0), r.getLong(1), r.getLong(2), r.getLong(3), r.getLong(4))
  }
```

Key	State
A	( 3 , 8 , 6 )
B	( 12 , 4 , 7 )

State is a tuple of 3 integers  
[(Int, Int, Int)]

```
.updateStateByKey[(Int, Int, Int)] (  
  (newItemsPerKey: Seq[ActivityByProduct], currentState: Option[(Int, Int, Int)]) => {  
    ↙  
  }  
)
```

Option[(Int, Int, Int)]



# Stateful Operations

- `updateStateByKey`
- `mapWithState`

```
activityByProduct
  .map { r => ((r.getString(0), r.getLong(1)),
    ActivityByProduct(r.getString(0), r.getLong(1), r.getLong(2), r.getLong(3), r.getLong(4))
  ) }
```

`mapWithState[StateType, MappedType](spec : StateSpec[K, V, ...., ....])`  
: `MapWithStateDStream[K, V, StateType, MappedType]`  
(Int, Int, Int)

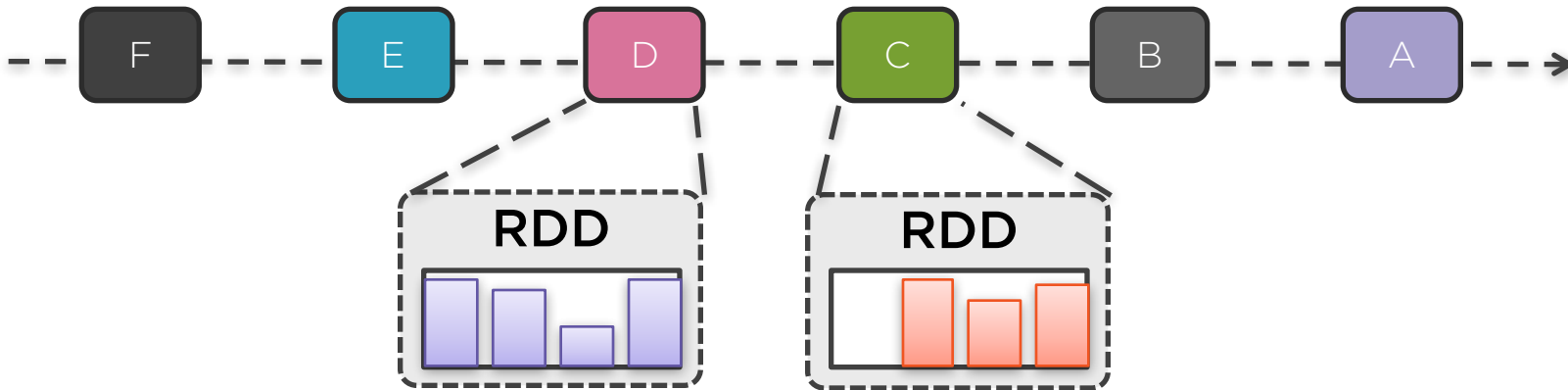
`val spec = StateSpec`  
  `.function((K, Option[V], State[StateType]) => Option[MappedType])`  
  `.timeout(t: Duration)`  
(Int, Int, Int)



# Stateful Operations

- `updateStateByKey`
- **`mapWithState`**

`mapWithState[StateType, MappedType](spec : StateSpec[K, V, ....., ..])`  
: `MapWithStateDStream[K, V, StateType, MappedType]`



`myStream.mapWithState(...).print()`

`myStream.mapWithState(...).stateSnapshots().print()`

Current State	
Key	State
C	( 3 , 8 , 6 )
D	( 12 , 4 , 7 )



# Cardinality Estimation using HyperLogLog



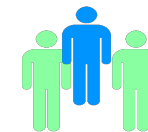
“Let’s shrink Big Data into Small Data ...  
and hope it magically becomes Great Data.”

## HLL - Cardinality estimation

Unique observations

Product “X” unique visitors  
in a certain timeframe

Timeframe expanded



# Cardinality Estimation using HyperLogLog



"Let's shrink Big Data into Small Data ...  
and hope it magically becomes Great Data."

## HLL - Cardinality estimation

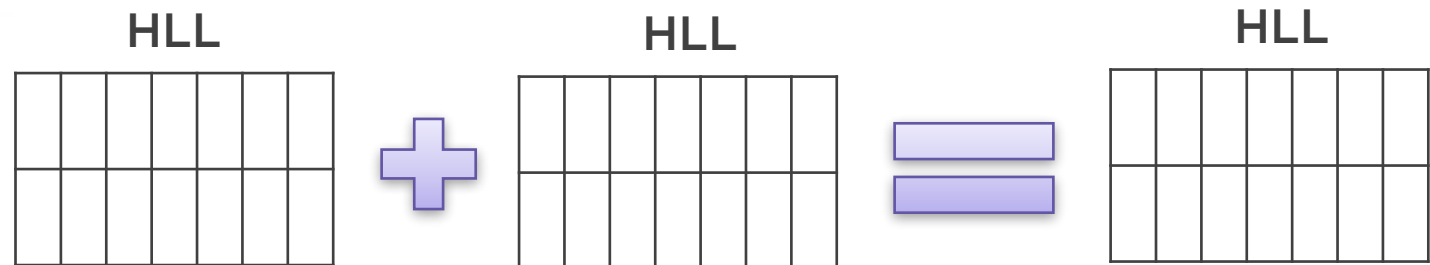
Unique observations

Naïve solution 3 Billion events > 60 GB of memory

HLL solves this in < 100s of KB

Based on bit pattern observables

Associative Data Structure



# Summary

- **Window Operations**
- **Stateful Transformations**
  - `updateStateByKey`
  - `mapWithState`
- **Cardinality Estimation using HyperLogLog**
- **Algebird Library**
  - `CountMinSketch`
  - `Bloom Filters`
  - `Priority Queues`

