



Apache Beam: 流处理与批处理的统一编程框架



何沛
阿里巴巴-中间件-实时计算
Apache Beam committer

Apache Beam

定义了流处理与批处理 **统一的编程模型**，
它生成的流水线具有 **很好的可移植性**，
它在执行层面定义了 **高效的执行协议**。



1. Apache Beam 编程模型

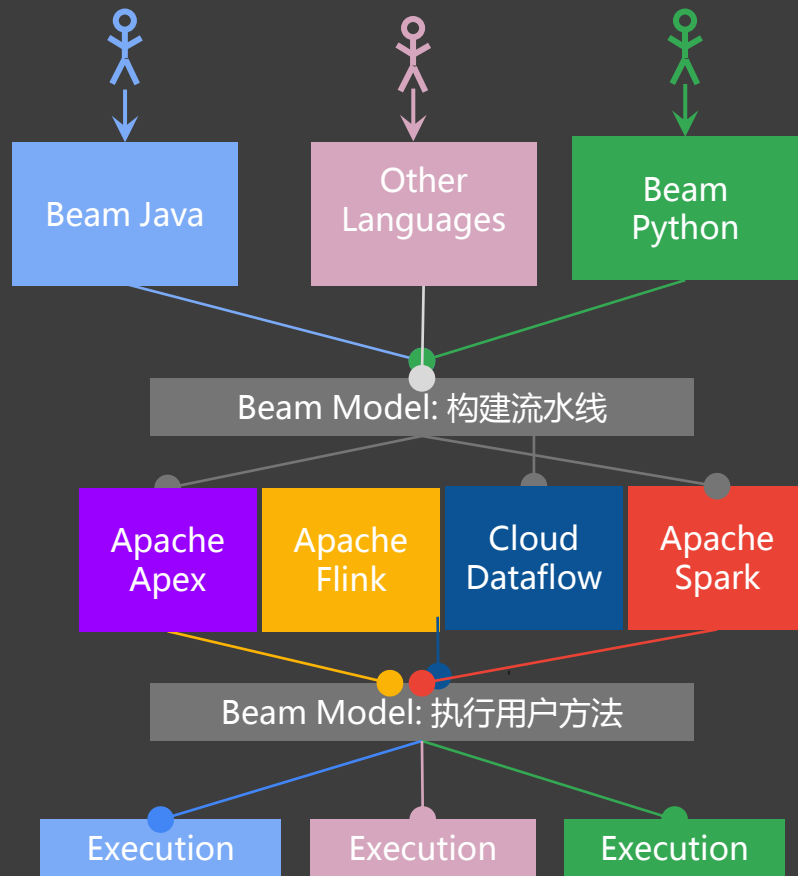
- What Where When How

2. SDK 构建流水线

- Java
- Python

3. 执行引擎 Runners

- Apex
- Spark
- Flink
- Google Cloud Dataflow



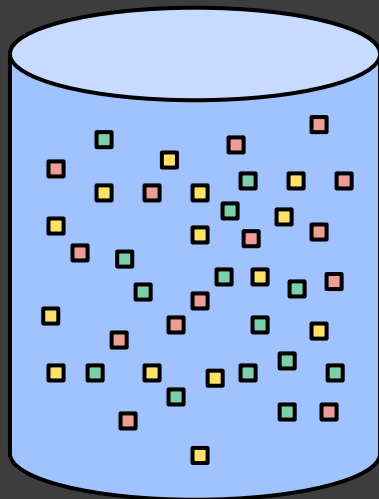


流处理与批处理统一编程模型

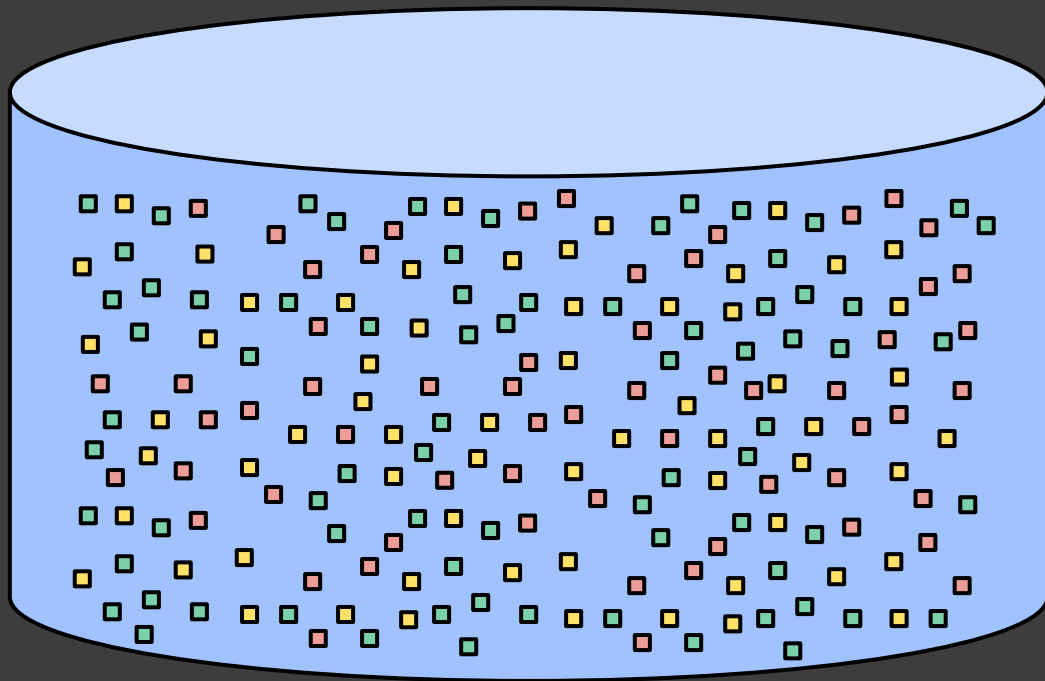




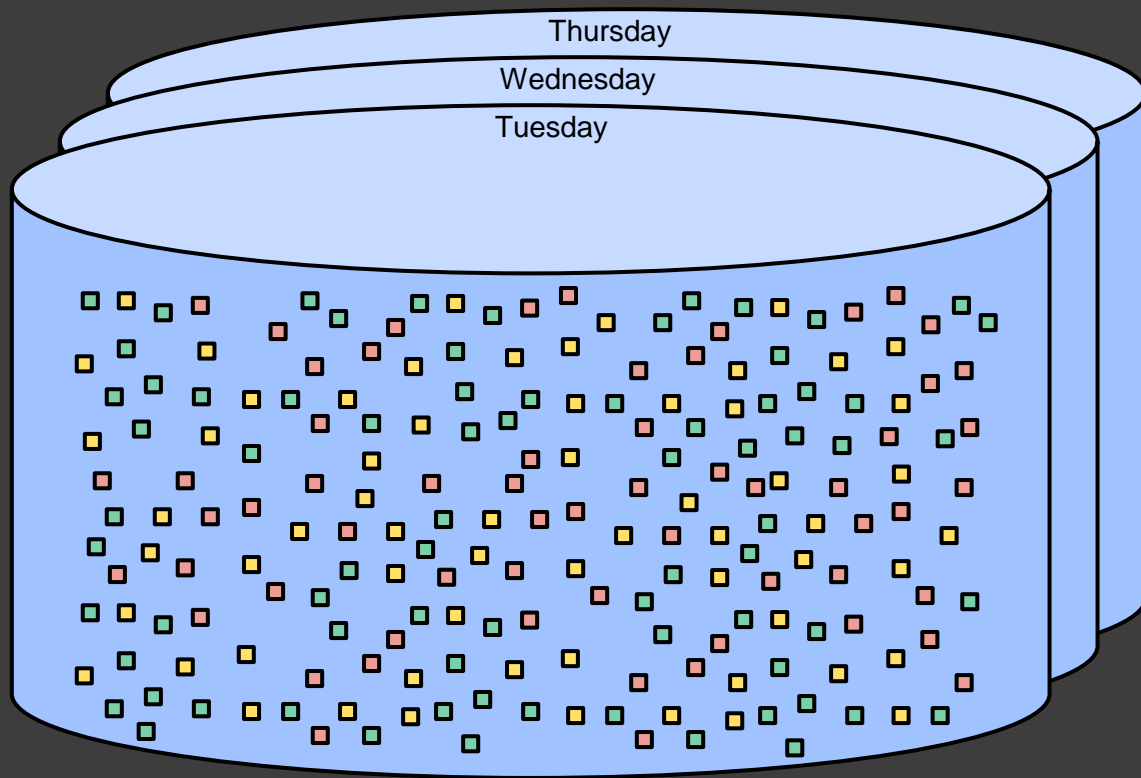
数据...



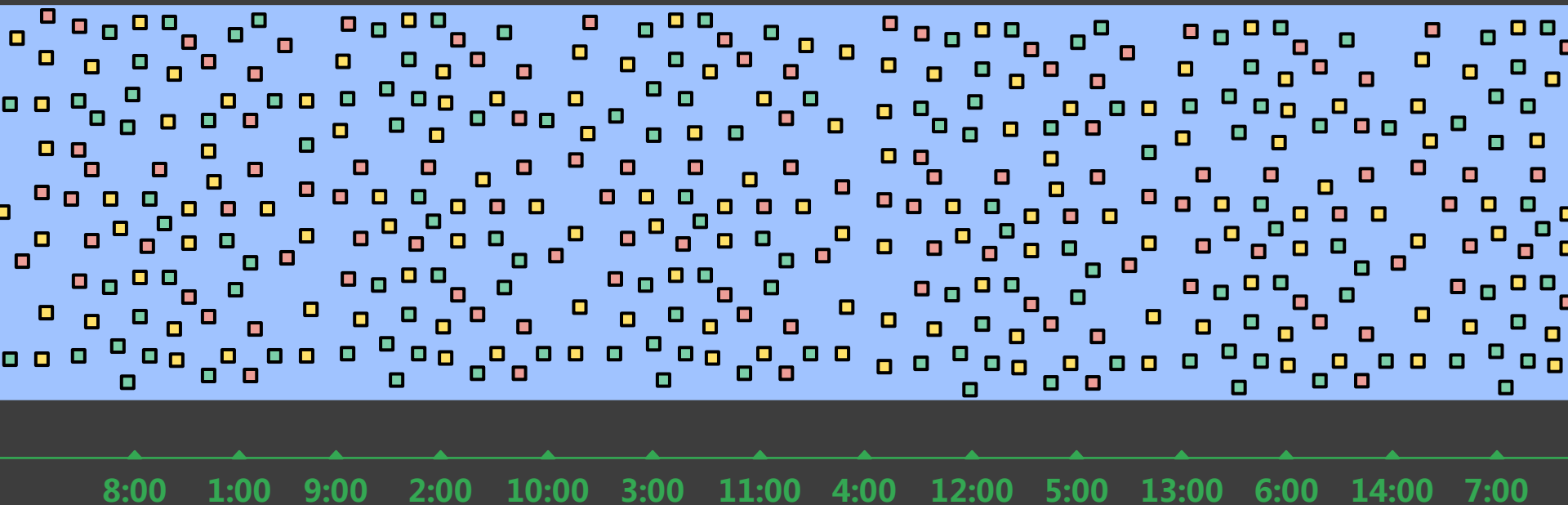
...很多数据...



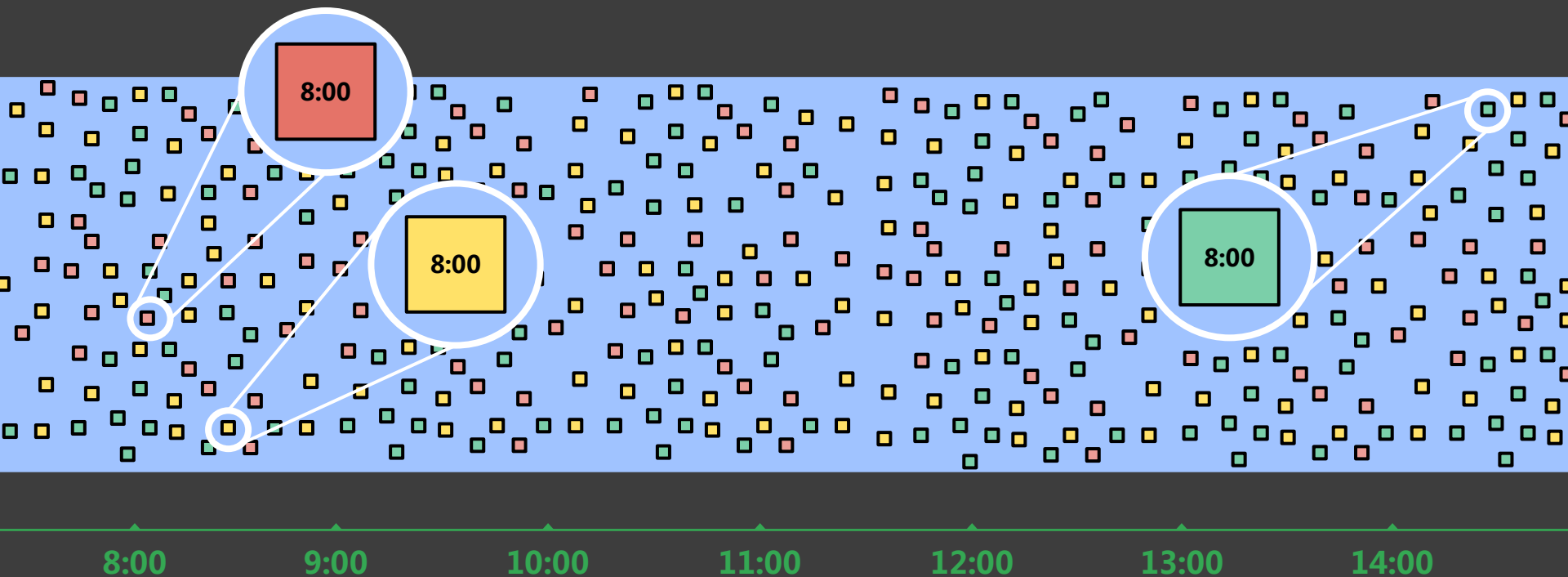
...越来越多的数据...



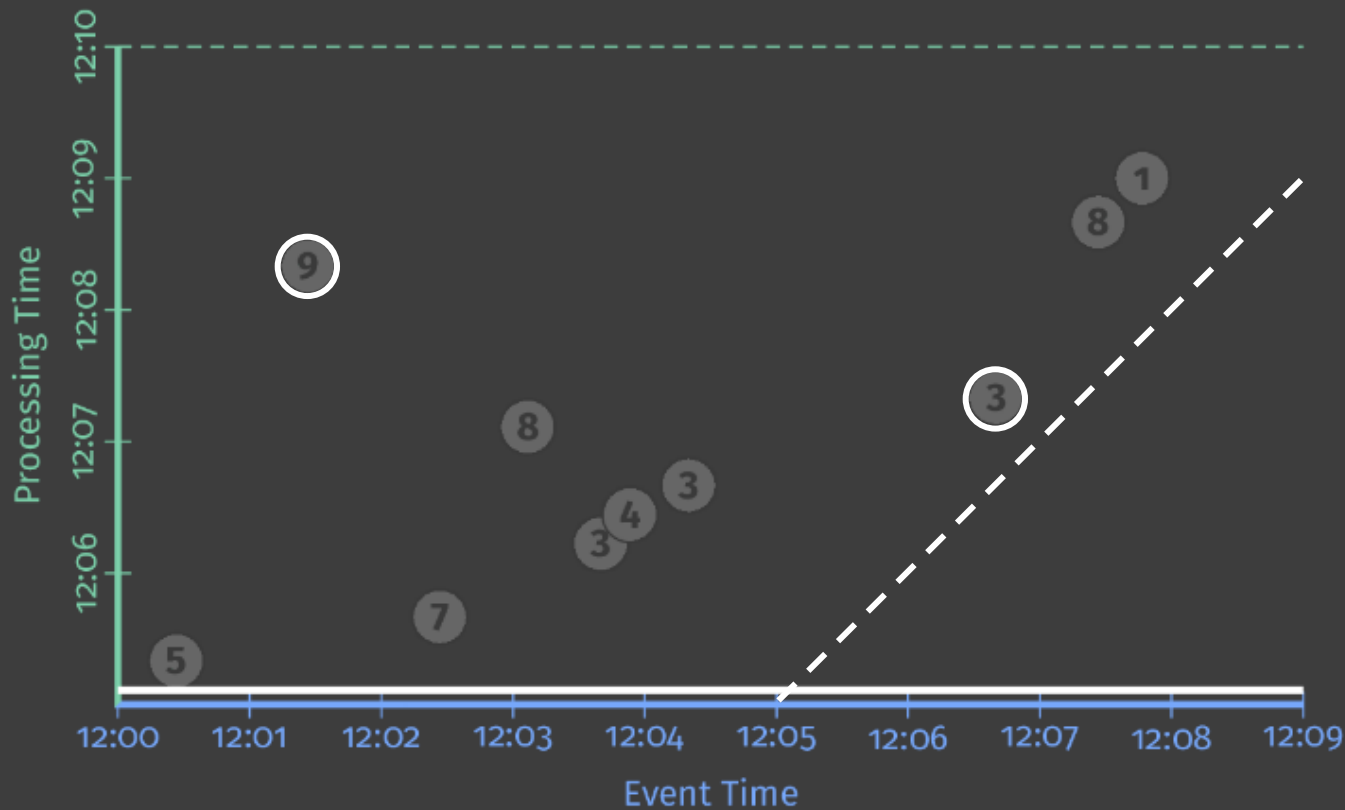
... 无边界的流数据...



...无序、有延迟的流数据.



事件时间 (Event time) 与 处理时间 (Processing Time)



What 业务逻辑是什么? (ParDo, Combine)

Where 在事件时间轴上对应哪里? (Windowing)

When 何时输出结果? (Triggering)

How 如何更新结果? (Accumulating)

What: 求和 (Sum)

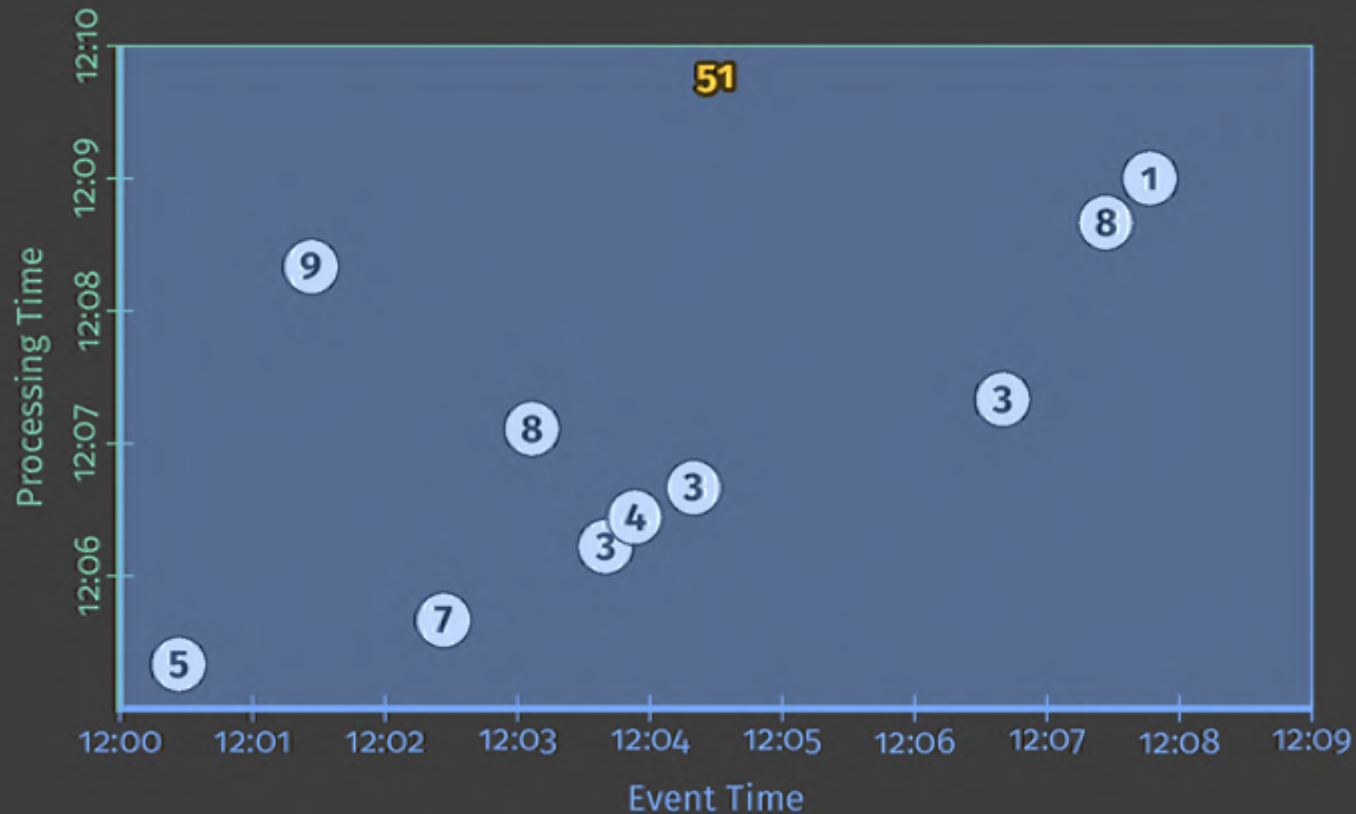
```
// Collection of raw log lines
PCollection<String> raw = IO.read(...);

// Element-wise transformation into team/score pairs
PCollection<KV<String, Integer>> input =
    raw.apply(ParDo.of(new ParseFn()));

// Composite transformation containing an aggregation
PCollection<KV<String, Integer>> scores =
    input.apply(Sum.integersPerKey());
```

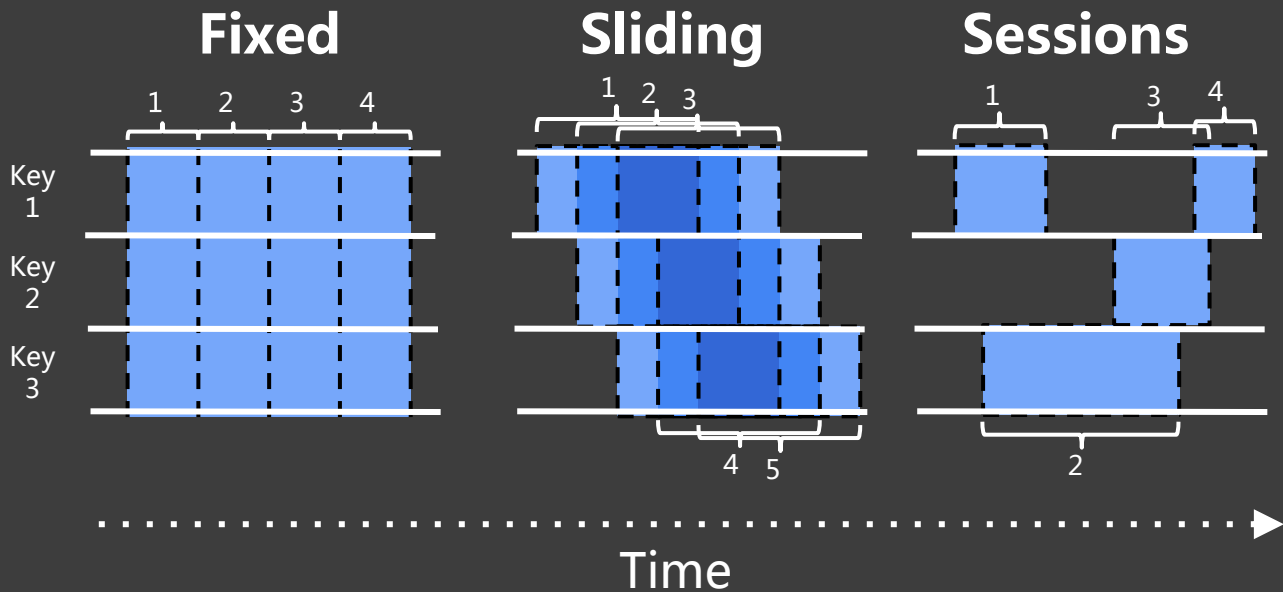
*All code snippets are pseudo-java -- details shortened or elided for clarity.

What: 求和 (Sum)



Where: 在事件时间轴上对应哪里?

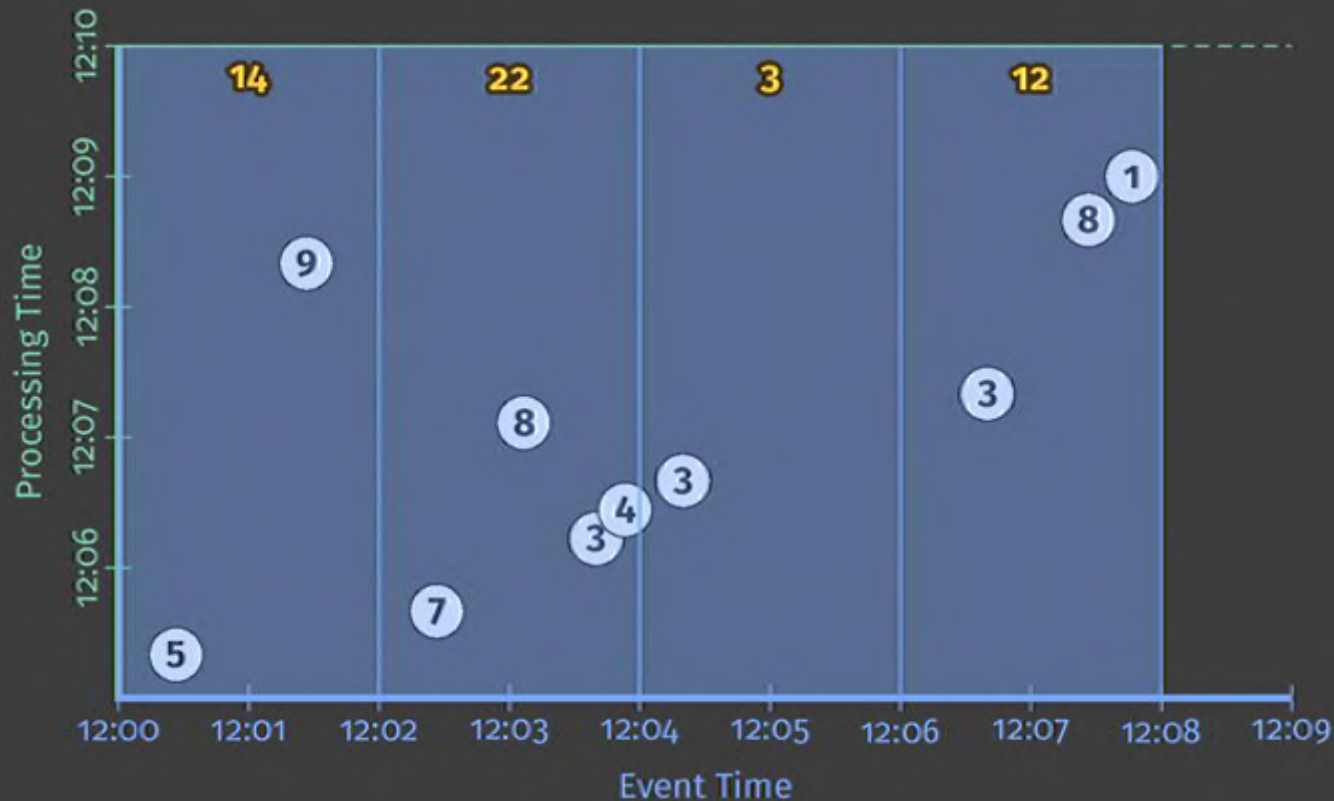
Windowing 可以将流数据划分为有边界的数据集



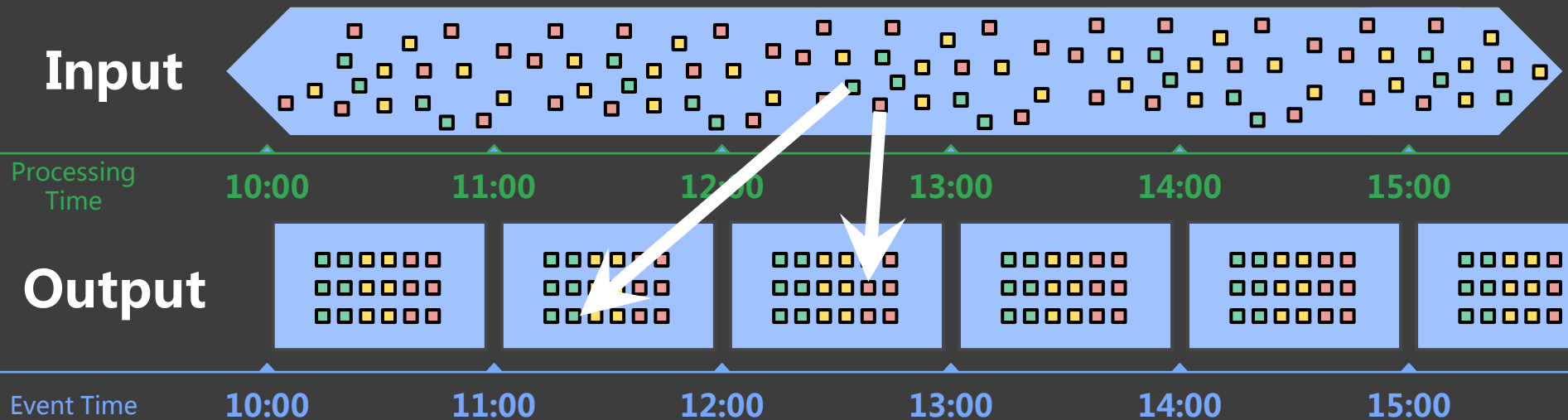
Where: 两分钟 固定窗口

```
PCollection<KV<String, Integer>> scores = input
    .apply(Window
        .into(FixedWindows.of(Duration.standardMinutes(2))))
    .apply(Sum.integersPerKey());
```


Where: 两分钟 固定窗口



流数据处理: 引入Watermark的概念



When: 何时输出计算结果?



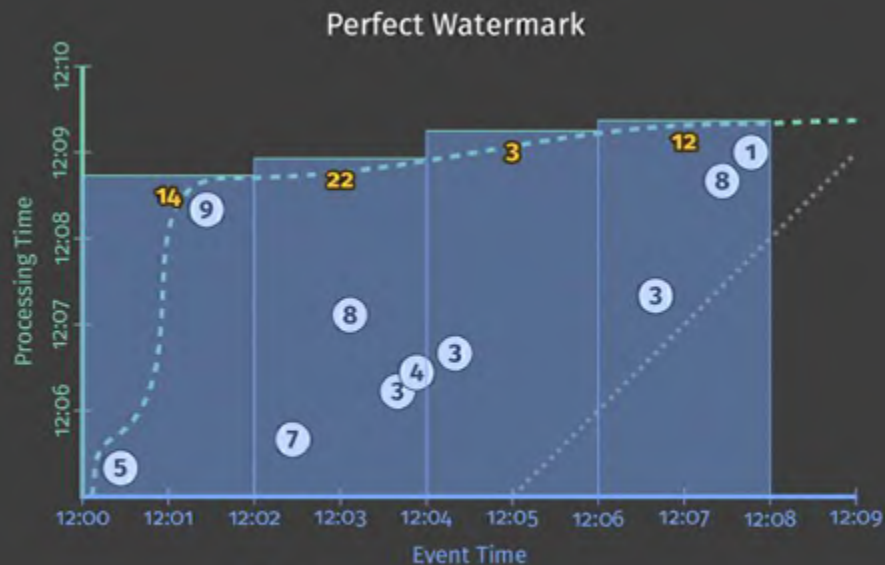
Triggers控制结果输出

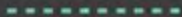

`triggering(AtWatermark())`

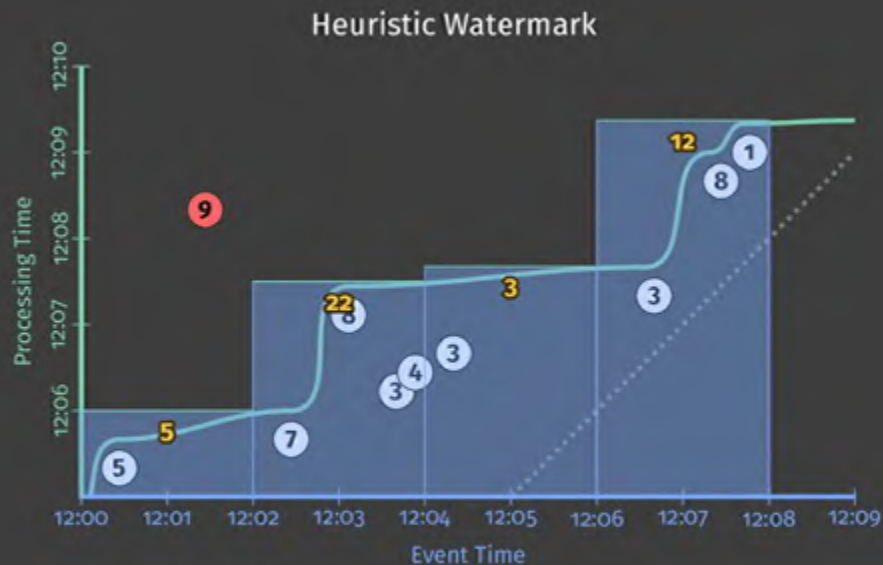
When: 按时输出 (Triggering at the Watermark)

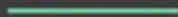

```
PCollection<KV<String, Integer>> scores = input
    .apply(Window
        .into(FixedWindows.of(Duration.standardMinutes(2))
            .triggering(AtWatermark())))
    .apply(Sum.integersPerKey());
```

When: 按时输出 (Triggering at the Watermark)



Perfect watermark: 
Ideal watermark: 

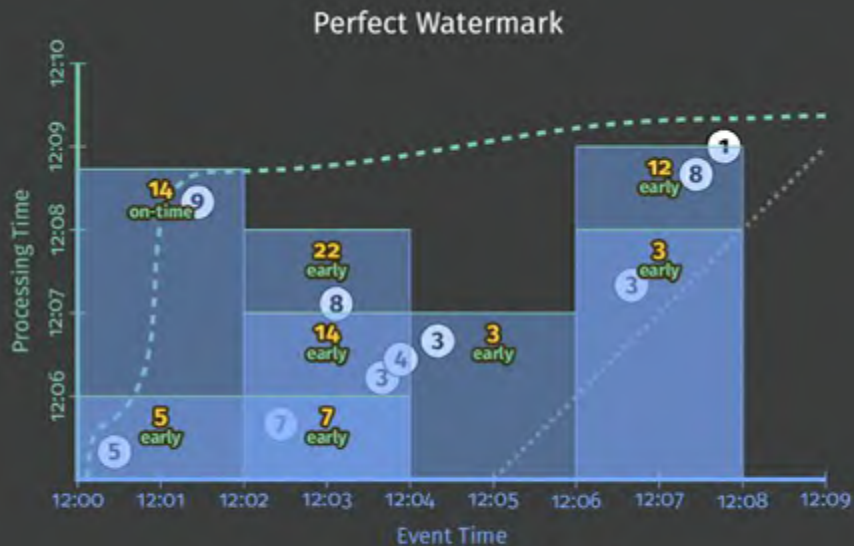


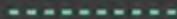

Heuristic watermark: 
Ideal watermark: 

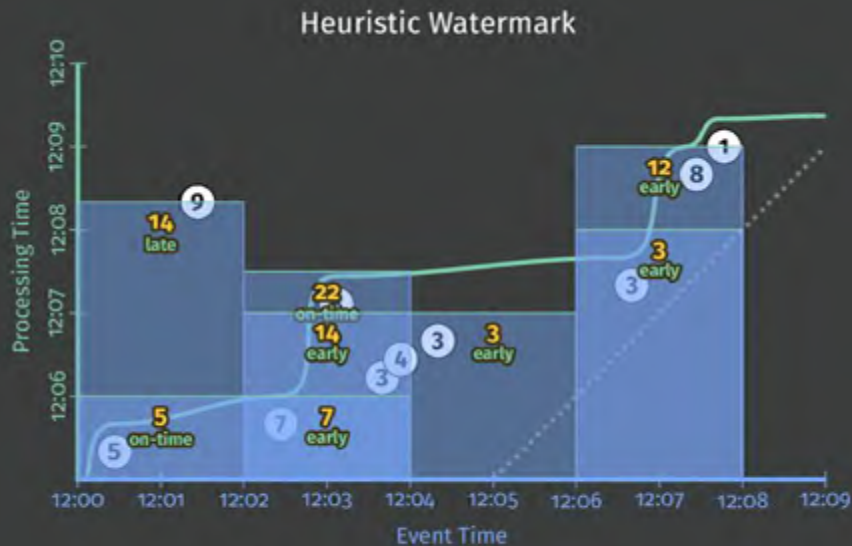
When: 提早输出(Early Firing) 和 迟到输出(Late Firing)



```
PCollection<KV<String, Integer>> scores = input
    .apply(Window
        .into(FixedWindows.of(Duration.standardMinutes(2)))
        .triggering(AtWatermark()
            .withEarlyFirings(AtPeriod(Duration.standardMinutes(1)))
            .withLateFirings(AtCount(1))))
    .apply(Sum.integersPerKey());
```

When: 提早输出(Early Firing) 和 迟到输出(Late Firing)



Perfect watermark: 
Ideal watermark: 



Heuristic watermark: 
Ideal watermark: 

How: 如何更新结果?

- 同一个窗口的多次输出是如何累计的?
- 取决于下游消费者

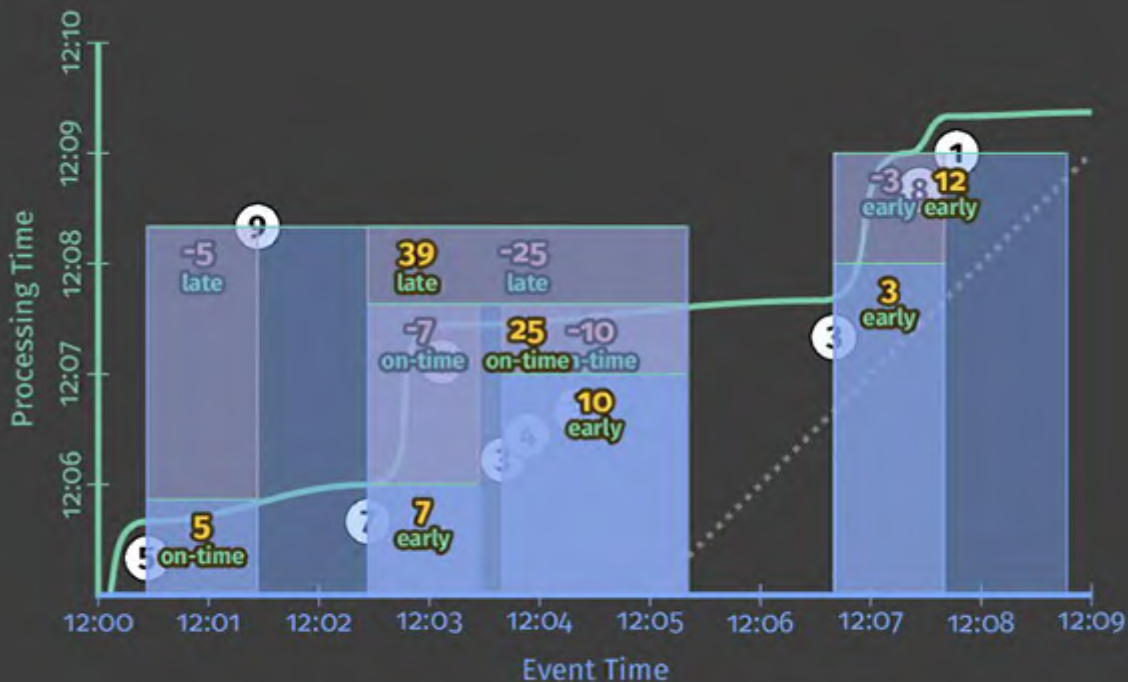
Firing	Elements	Discarding	Accumulating	Acc. & Retracting
提前输出	[3]	3	3	3
按时输出	[5, 1]	6	9	(9, -3)
迟到输出	[2]	2	11	(11, -9)
<i>Last Observed</i>		2	11	11
<i>Total Observed</i>		11	23	11

(Accumulating & Retracting not yet implemented.)

How: 加上新的，減去旧的

```
PCollection<KV<String, Integer>> scores = input
    .apply(Window
        .into(Sessions.withGapDuration(Duration.standardMinutes(1)))
            .triggering(AtWatermark()
                .withEarlyFirings(AtPeriod(Duration.standardMinutes(1)))
                .withLateFirings(AtCount(1)))
            .accumulatingAndRetractingFiredPanels())
        .apply(Sum.integersPerKey()));
```

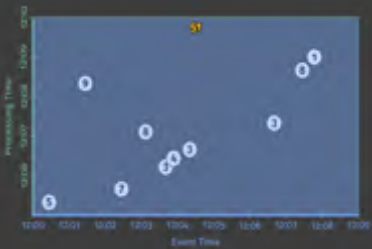
How: 加上新的，減去旧的



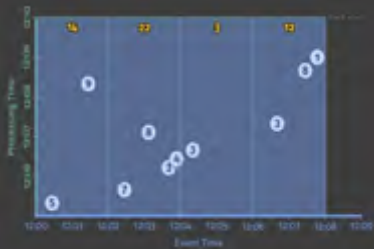
Heuristic watermark: ————

Ideal watermark: ······

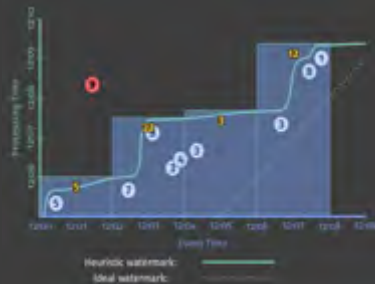
What When Where How



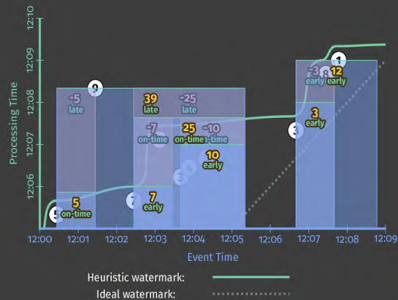
1. 传统批处理



2. 支持Window的批处理



3. 流数据处理



4. 支持Retractions的流处理



Apache Beam流水线的可移植性



数据流水线开发的权衡

1 + 1 = 2

完整性



实时性



运维成本



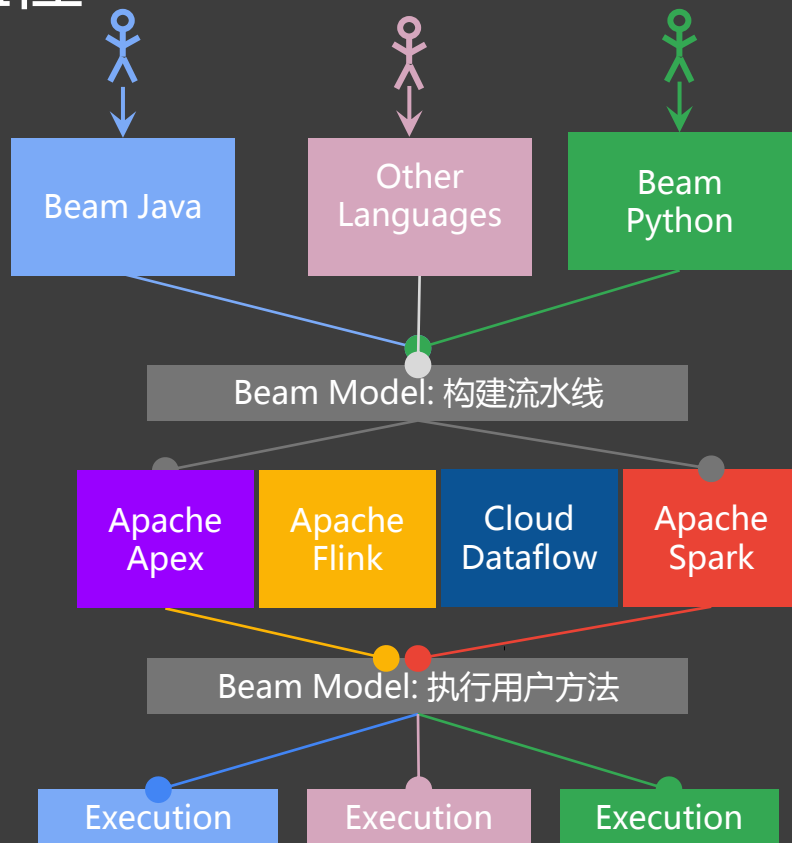
Apache Beam流水线的可移植性

1. 编写

- 选择最熟悉的语言*

2. 执行

- 选择最适合的引擎
- 统一的语义便于引擎的切换



执行引擎对Beam功能的支持表

What is being computed?

	Beam Model	Google Cloud Dataflow	Apache Flink	Apache Spark	Apache Apex
ParDo	✓	✓	✓	✓	✓
GroupByKey	✓	✓	✓	-	✓
Flatten	✓	✓	✓	✓	✓
Combine	✓	✓	✓	✓	✓
Composite Transforms	✓	-	-	-	-
Side Inputs	✓	✓	✓	✓	✓
Source API	✓	✓	✓	✓	✓
Aggregators	-	-	-	-	✗
Stateful Processing	✓	-	-	✗	✗

Where in event time?

	Beam Model	Google Cloud Dataflow	Apache Flink	Apache Spark	Apache Apex
Global windows	✓	✓	✓	✓	✓
Fixed windows	✓	✓	✓	✓	✓
Sliding windows	✓	✓	✓	✓	✓
Session windows	✓	✓	✓	✓	✓
Custom windows	✓	✓	✓	✓	✓
Custom merging windows	✓	✓	✓	✓	✓
Timestamp control	✓	✓	✓	✓	✓

When in processing time?

	Beam Model	Google Cloud Dataflow	Apache Flink	Apache Spark	Apache Apex
Configurable triggering	✓	✓	✓	✗	✓
Event-time triggers	✓	✓	✓	✗	✓
Processing-time triggers	✓	✓	✓	✓	✓
Count triggers	✓	✓	✓	✗	✓
[Meta]data driven triggers	✗ (BEAM-101)	✗	✗	✗	✗
Composite triggers	✓	✓	✓	✗	✓
Allowed lateness	✓	✓	✓	✗	✓
Timers	✓	-	-	✗	✗

How do refinements relate?

	Beam Model	Google Cloud Dataflow	Apache Flink	Apache Spark	Apache Apex
Discarding	✓	✓	✓	✓	✓
Accumulating	✓	✓	✓	✗	✓
Accumulating & Retracting	✗ (BEAM-91)	✗	✗	✗	✗

<https://beam.apache.org/learn/runners/capability-matrix/>



Apache Beam的高效性





Apache Beam与阿里巴巴



Growing the Community

Apache Beam交流群



该二维码7天内(3月31日前)有效, 重新进入将更新

- 如果你有大数据APIs, 写一个Beam **SDK** 或者 **DSL**.
- 如果你有一个存储系统、消息系统, 写一个Beam **IO connector**!
- 如果你有一个分布式引擎, 写一个Beam **runner**!

谢谢!



Learn More!

Apache Beam

<https://beam.apache.org>

Resources

<https://beam.apache.org/documentation/resources/>

Join the Beam mailing lists!

user-subscribe@beam.apache.org

dev-subscribe@beam.apache.org