

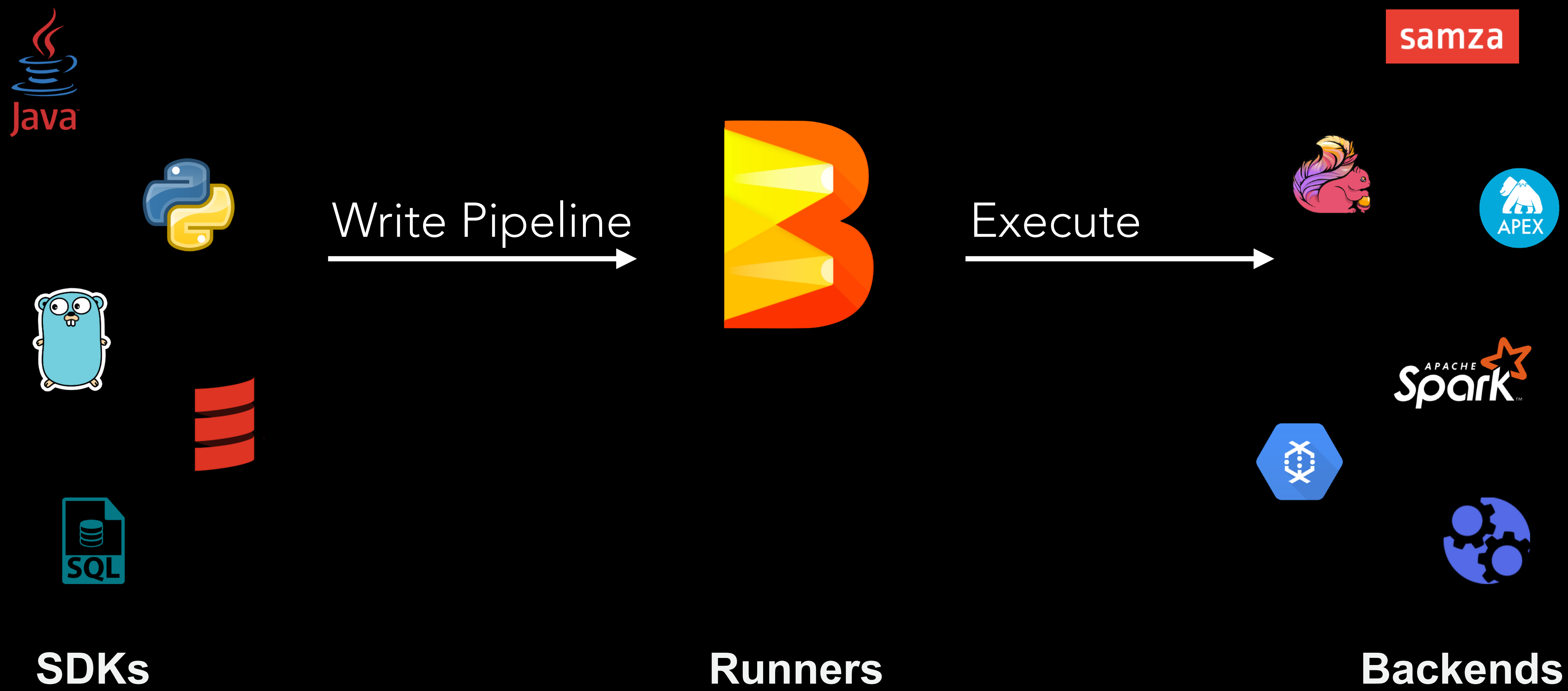
TOWARDS PORTABILITY AND BEYOND



DATA PROCESSING WITH
APACHE BEAM

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BEAM VISION



THE BEAM MODEL

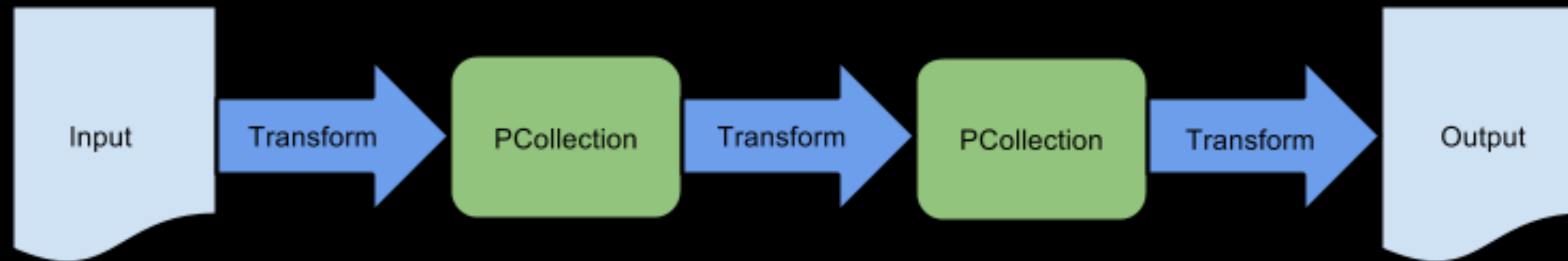
- Unified **batch** and **stream** programming model
- **Stream:** Batch is just a bounded stream
- Massively parallelizable **Transformations**
- **Event Time** as an explicit concept



2015 VLDB: "The Dataflow Model: A Practical Approach to Balancing Correctness, Latency, and Cost in Massive-Scale, Unbounded, Out-of-Order Data Processing"

BEAM API

THE BEAM API



- Pipeline: An (acyclic) graph of PCollections
- `[Output PCollection] = [Input PCollection].apply([Transform])`
 - `Pipeline p = Pipeline.create(options)`
 - `PCollection pCollection = p.apply(...).apply(...)...`
 - `p.run()`

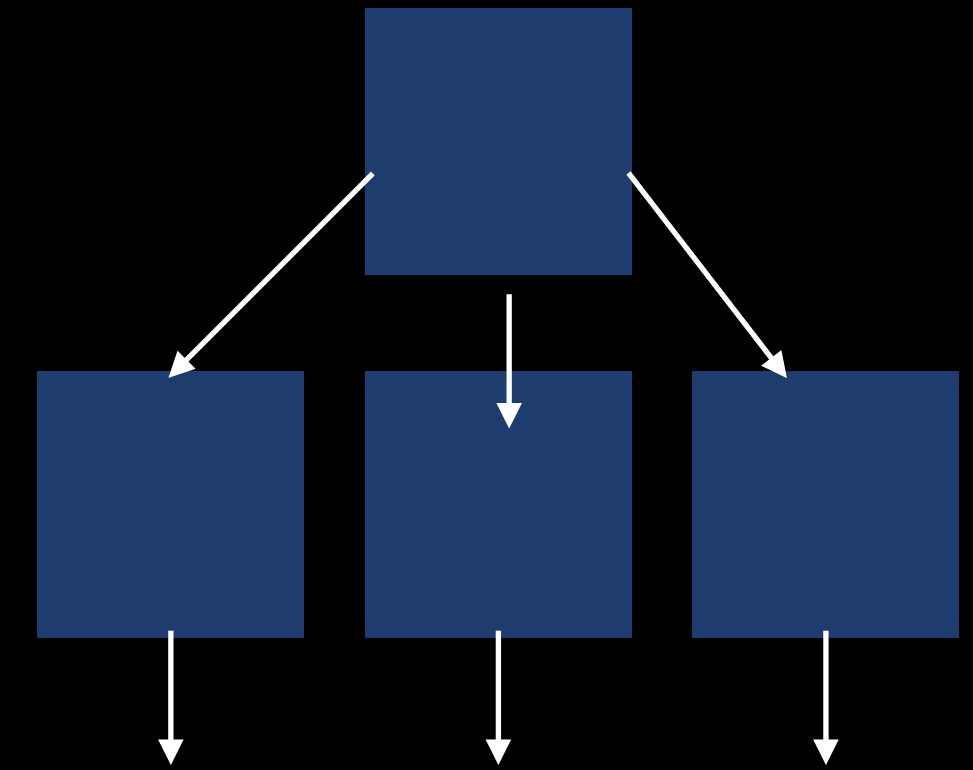
INPUT

```
Pipeline p = Pipeline.create(options);

PCollection<String> input1 = p.apply(
    "ReadMyFile",
    TextIO.read().from("protocol://path/to/some/inputData.txt"));

PCollection<String> input2 = p.apply(
    Create.of("DataEngConf", "is", "awesome"));

PCollection<KV<Long, String>> input3 = p.apply(
    KafkaIO.<Long, String>read()
        .withBootstrapServers("broker_1:9092,broker_2:9092")
        .withTopic("my_topic")
        .withKeyDeserializer(LongDeserializer.class)
        .withValueDeserializer(StringDeserializer.class))
```



CORE PRIMITIVES TRANSFORMS

ParDo

input -> output

“to” -> KV<“to”, 1>

“be” -> KV<“be”, 1>

“or” -> KV<“or”, 1>

“not” -> KV<“not”, 1>

“to” -> KV<“to”, 1>

“be” -> KV<“be”, 1>

Map/Reduce Phase

GroupByKey

KV<k, v>... -> KV<k, [v...]>

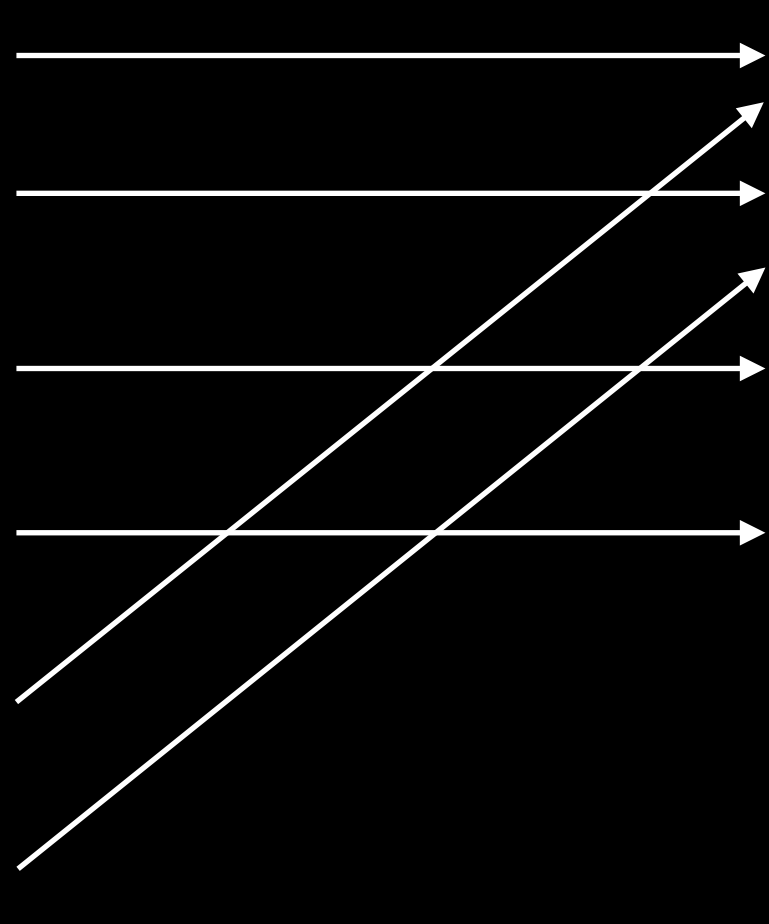
KV<“to”, [1, 1]>

KV<“be”, [1, 1]>

KV<“or”, [1]>

KV<“not”, [1]>

Shuffle Phase



CORE PRIMITIVES TRANSFORMS

ParDo

```
PCollection<String> words = ...;
```

```
PCollection<KV<String, Integer>> wordCounts = words.apply("AssignWordCounts",  
    ParDo.of(new DoFn<String, KV<String, Integer>>() {  
        @ProcessElement  
        public void processElement(@Element String word, OutputReceiver<Integer> out) {  
            out.output(KV.of(word, 1));  
        }  
    }  
));
```

GroupByKey

```
PCollection<KV<String, Iterable<Integer>>> groupByWordCounts =  
    wordCounts.apply("GroupByWords", GroupByKey.create());
```

COMPOSITE TRANSFORMS

Combine

```
PCollection<KV<String, Integer> wordCounts = ...
```

```
PCollection<KV<String, Integer> combinedWordCounts = wordCounts.apply(  
Combine.perKey(new SerializableFunction<Iterable<Integer>, Integer> {  
    @Override  
    public Integer apply(Iterable<Integer> input) {  
        int sum = 0;  
        for (int count : input) {  
            sum += count;  
        }  
        return sum;  
    }  
});
```

For more sophisticated
combing, you can define
a CombineFn.

(Map/) Shuffle/Reduce Phase

MORE TRANSFORMS

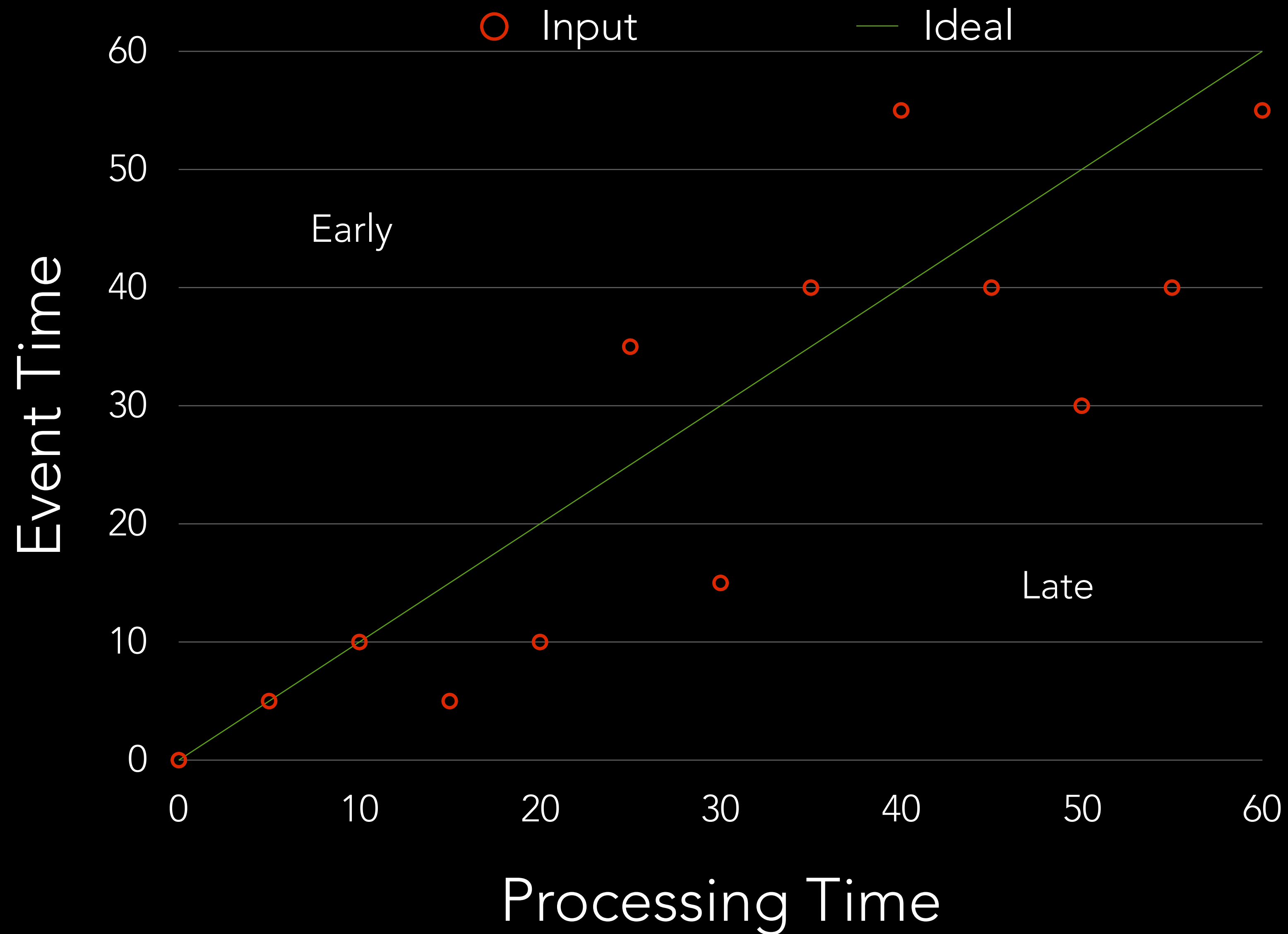
- CoGroupByKey (Join)
- Flatten
- Partition
- Define your own!
- Also: Side Inputs / Multiple Outputs / State / Timers

PROCESSING UNBOUNDED DATA

PROCESSING (UN)BOUNDED DATA

- Stream are unbounded by nature
- **Windows** group data according to a windowing function
- **Triggers** decide when to kick off execution of windows
- Event Time is the predominant domain for windowing
 - Every element has an event **timestamp**
 - **Watermark** indicates the current event time

EVENT VS PROCESSING TIME



EVENT VS PROCESSING TIME: STAR WARS

Year	1977	1980	1983	1999	2002	2005	2015	2017	2019
Episode	IV	V	VI	I	II	III	VII	VIII	IX

These episodes appeared out-of-order

Year	1999	2002	2005	1977	1980	1983	2015	2017	2019
Episode	I	II	III	IV	V	VI	VII	VIII	IX

That's much better!

PROCESSING TIME

EVENT TIME

THE BEAM MODEL: A SIMPLE PIPELINE

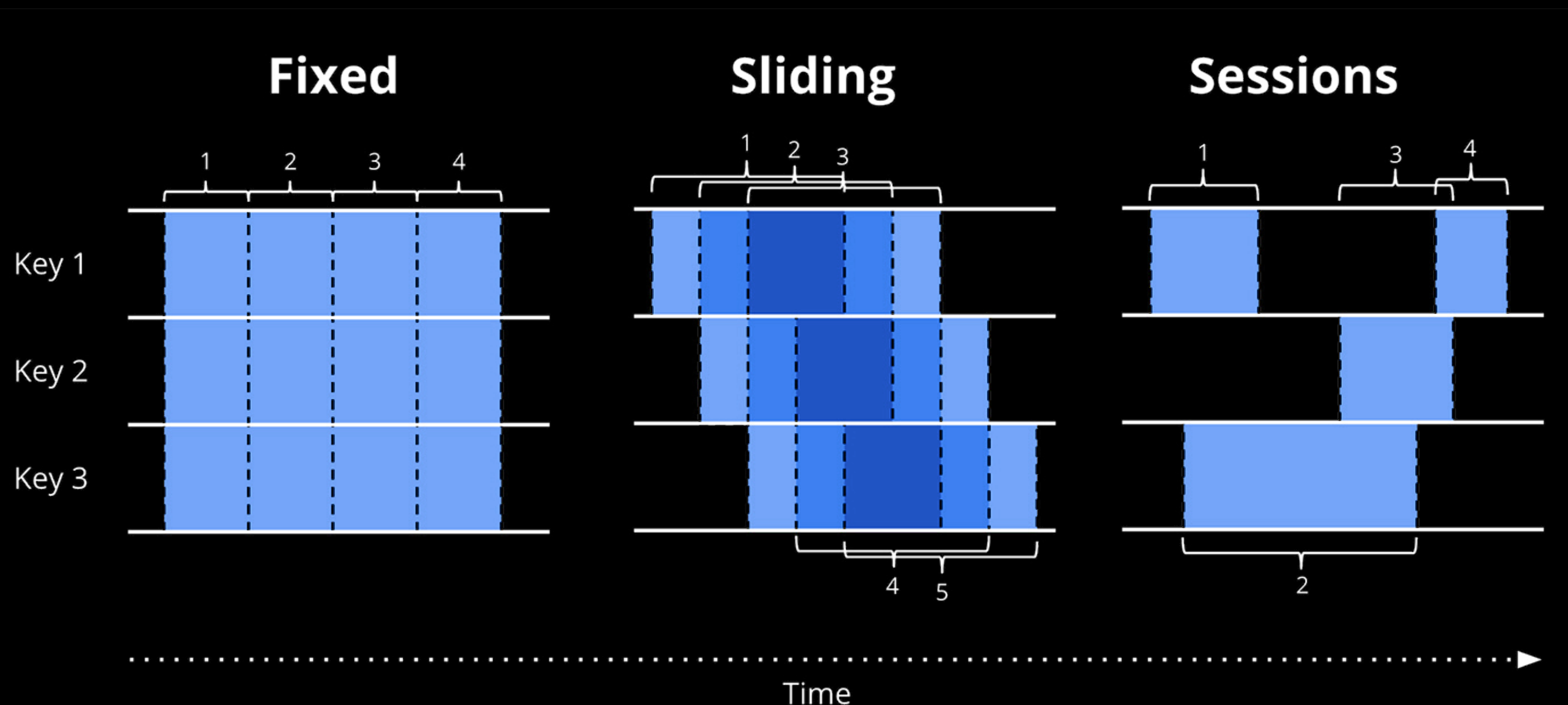
What, Where, When, How

```
PCollection<KV<String, Integer>> scores = input  
    .apply(Sum.integersPerKey());
```

A SIMPLE PIPELINE: WINDOWING

What, Where, When, How

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



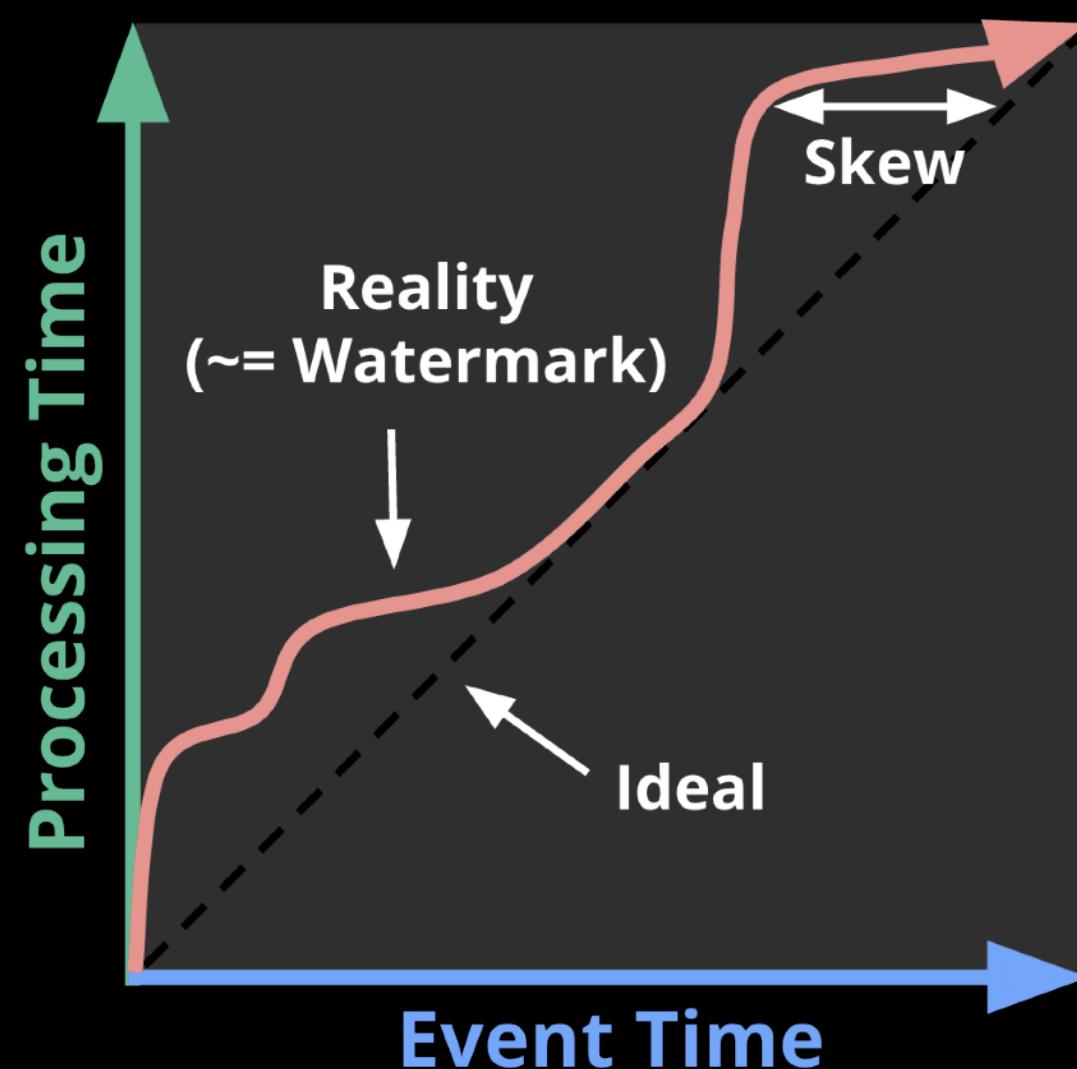
Window Types

- Global Window
- Fixed Time Windows
- Sliding Time Windows
- Per-Session Windows

A SIMPLE PIPELINE: TRIGGERING

What, Where, When, How

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



Triggers

- Event time
- Processing
- Data-driven
- Composite triggers

A SIMPLE PIPELINE: TRIGGERING

What, Where, When, How

```
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());
```

Triggers

- Event time
- Processing
- Data-driven
- Composite triggers

A SIMPLE PIPELINE: REFINEMENT

What, Where, When, How

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

JAVA VS PYTHON



```

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

```



```

scores = input
    | WindowInto(FixedWindows(120)
        trigger=AfterWatermark(
            early=AfterProcessingTime(60),
            late=AfterCount(1))
        accumulation_mode=ACCUMULATING)
    | CombinePerKey(sum)

```

EXECUTE WITH CHOICE OF RUNNER

```
input  = pipeline | ReadFromText("/path/to/text*") | Map(lambda line: ...)
scores = input
      | WindowInto(FixedWindows(120)
                  trigger=AfterWatermark(
                      early=AfterProcessingTime(60),
                      late=AfterCount(1))
                  accumulation_mode=ACCUMULATING)
      | CombinePerKey(sum)
      | WriteToText("/path/to/outputs")

MyRunner().run(pipeline)
```

RUNNERS

RUNNERS



Direct



Apache Flink



Apache Spark



Apache Apex



Apache Samza



Google Cloud
Dataflow



Apache Gearpump

WIP



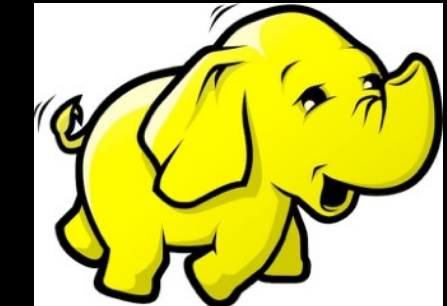
Ali Baba
JStorm



Apache Storm



IBM Streams



Hadoop
MapReduce

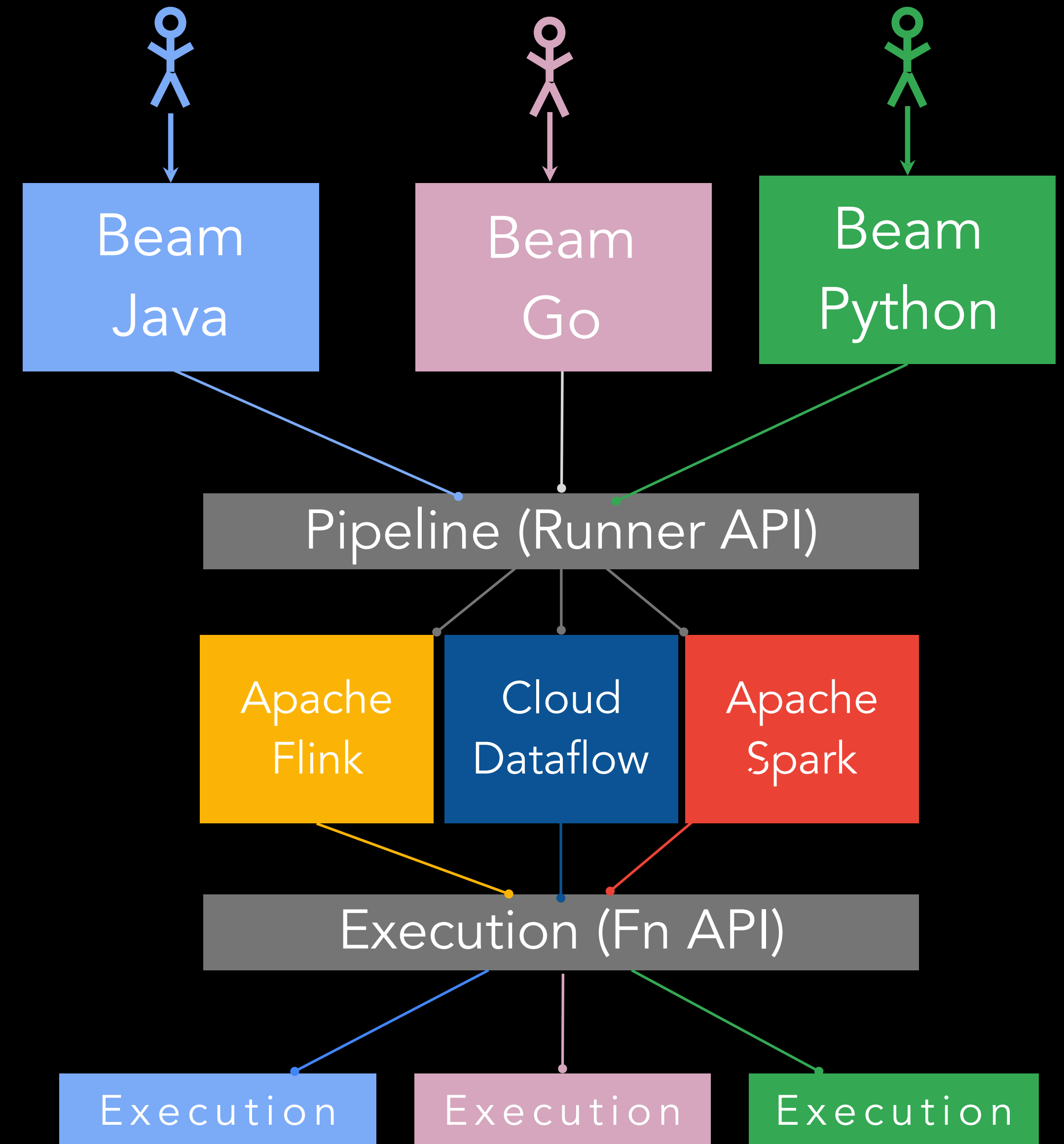
PROBLEMS WITH THIS APPROACH

- All execution backends written in Java
- Can Python run on top of Java?
- N Runners, M languages $\Rightarrow$ $N*M$ translation paths?
- Submission Flow

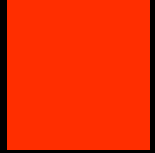
PORTABILITY

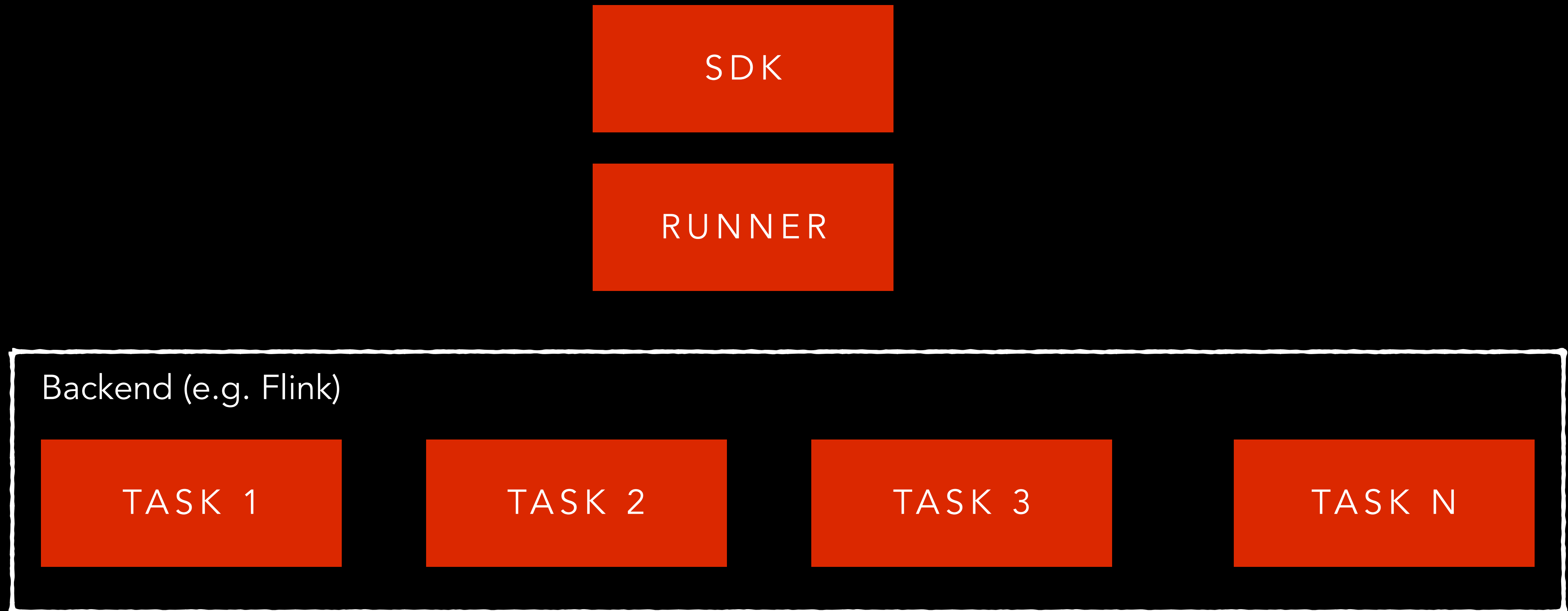
THE BEAM VISION

- Portability in Beam means Pipelines can be written and executed in any supported SDK (Java/Python/Go)
- Pipelines also contain language-specific code (e.g. map/reduce functions)
- Libraries of the language can be used (!)



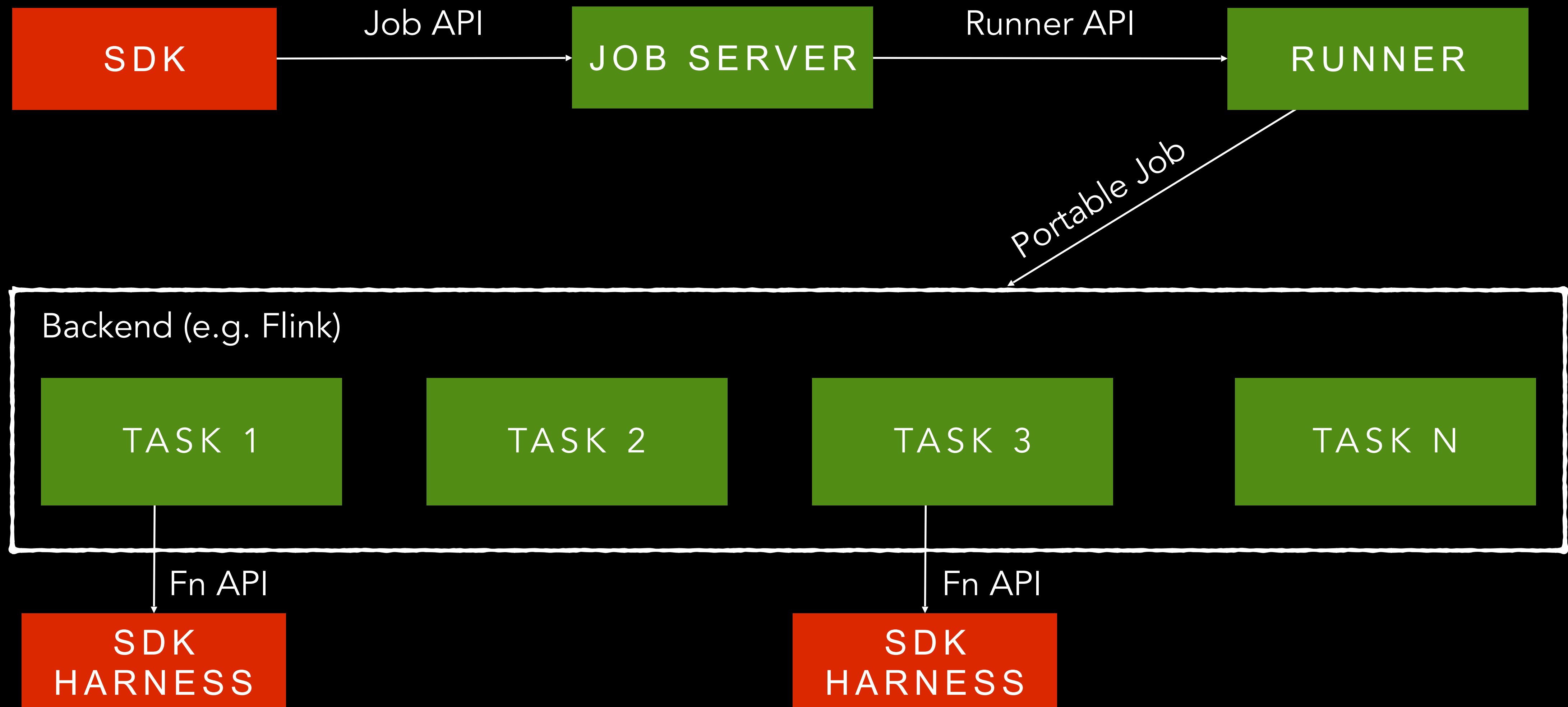
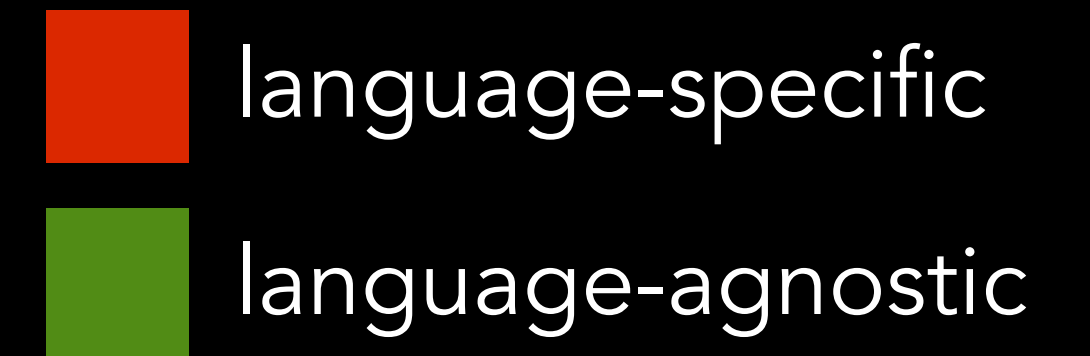
WITHOUT PORTABILITY

 language-specific



All components are tight to a single language

WITH PORTABILITY



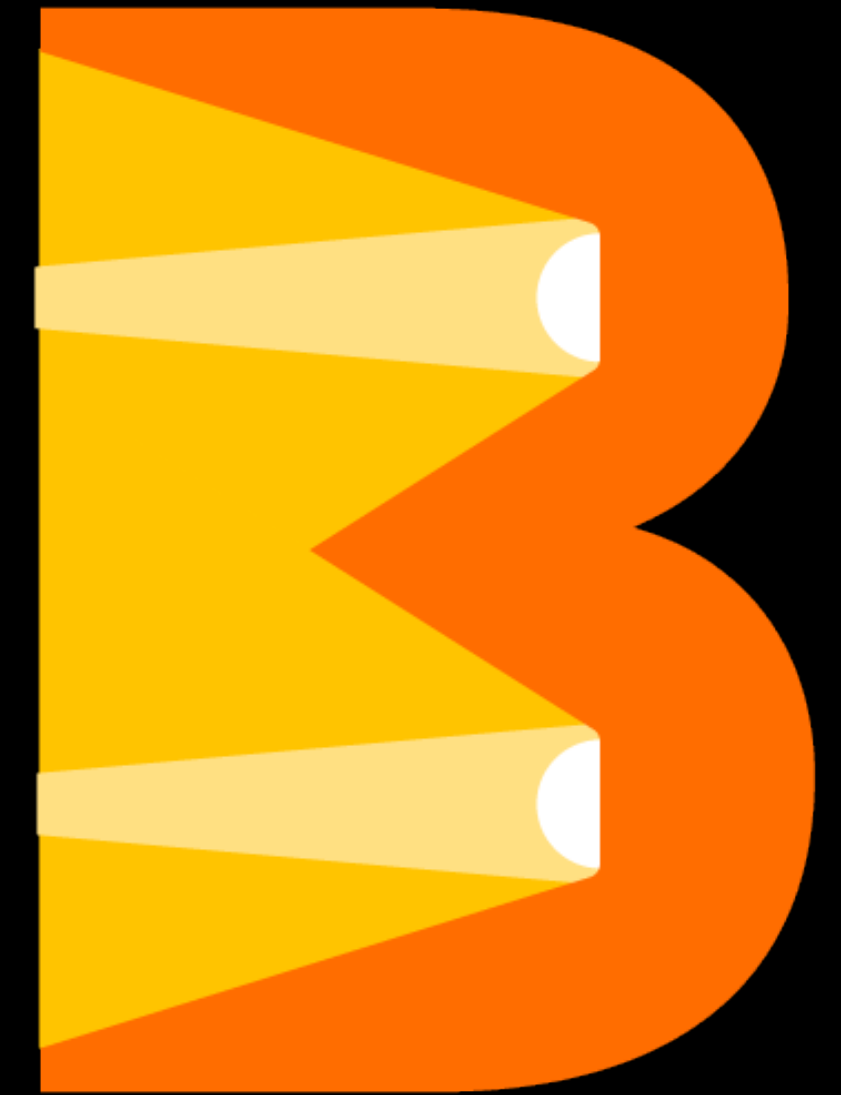
WHAT IS THE STATE OF THE BEAM VISION?

- $|\text{Runners}| * |\text{SDKs}|$ is not feasible
- Instead each SDK only implements a Portable Runner
- Flink Runner is the first OSS Runner to work with the Portable Runner
- Cross-language pipelines in the future
- Overhead has been measured to be 5-10%, could be less in real-world scenarios

DEMO TIME

HOW TO GET INVOLVED

- **Visit** beam.apache.org
 - Documentation
 - Examples
- **Subscribe** to the mailing lists:
 - user-subscribe@beam.apache.org
 - dev-subscribe@beam.apache.org
- **Join** the ASF Slack channel #beam
- **Follow** @ApacheBeam



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