

Beam at Talend: the long road together

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- 2006** ● Founded in 2006, Talend was the first company to market open-source data integration software;
- 2006** ● Released in October 2006, Talend Open Studio is the company's first product;
- 2007** ● July 2007, Talend launched its first commercial version, Talend Data Integration;
- 2015** ● March 2015, the company launched Talend Integration Cloud to enable developers to simplify and accelerate cloud and hybrid integration projects;
- 2016** ● January 2016, Talend joins Cloudera, Data Artisans, Google, Cask and Paypal on the Apache Foundation's Google's Cloud Dataflow project - Apache Beam;
- 2018** ● May 2018, Talend launched Talend Data Streams for AWS - a new free offering for self-service integration;
- 2019** ● April 2019, the company launched Talend Pipeline Designer (formerly Talend Data Streams), a next generation data integration design environment included in Talend Cloud.
- 2023** ● May 2023, Qlik acquires Talend

Talend and Open Source

- Talend has a rich Open Source culture from the very beginning ;
- Talend is a long-time partner of the ASF ;
- Open Source team at Talend is ASF contributor for many projects:
 - notably in the *Apache CXF, Camel, Karaf, ActiveMQ, Beam, Spark, Flink, Avro* and other projects;
- Help to mentor numerous projects through the ASF Incubator ;
 - Beam is a good example
- The company is also a member of other open source foundations:
 - *Java Community Process (JCP), Eclipse Foundation, OW226 and the Open Source School.*

Beam at Talend

🔍 The long and winding road...



(c) <https://parisandbeyond2012.com/>

Talend Open Studio / Data Integration

- Talend Open Studio is a free open source ETL tool for Data Integration and Big Data;
- Eclipse based developer tool and job designer;
- Drag&drop components and connect them to create and run ETL/ELT jobs;
- No need to write a single line of code.

The screenshot displays the Talend Open Studio interface for a data integration job. The main workspace shows a job design with the following components and connections:

- fact_ratings** (Input) connects to **Join**.
- dim_users** (Input) connects to **Join**.
- Join** connects to **Ag_Output (Main)**.
- Ag_Output (Main)** connects to **Avg_Rating**.
- Avg_Rating** connects to **row1 (Main)**.
- row1 (Main)** connects to **dim_stateratings** (Output).

The **Avg_Rating (AggregateRow_1)** component configuration is shown in the bottom right:

Output column	Function	Input column position	Ignore null values
Average_Rating	count	Movie_Id	<input type="checkbox"/>

The left sidebar shows the project structure, including a metadata tree with table schemas like **dim_movies**, **dim_stateratings**, **dim_users**, and **fact_ratings**. The bottom left pane shows the job outline with components like **tAggregateRow_1 (Avg_Rating)**, **tMap_1 (Join)**, **tMysqlInput_1 ("fact_ratings")**, **tMysqlInput_2 ("dim_users")**, and **tMysqlOutput_1 ("dim_stateratings")**.

Talend Pipeline Designer (TPD)

- Modern flexible integration tool to process data in easy and powerful manner;
- Provides a graphical interactive Web UI to create complex pipelines;
- Live preview of data changes;
- Schema-based data collections;
- Batch & Streaming;
- Portable & Scalable;
- Uses Beam under the hood!

The screenshot displays the Talend Pipeline Designer (TPD) interface. At the top, the pipeline is titled "Pipeline Example" and is set to "Batch" mode. The pipeline consists of three main stages: "Input", "Processors", and "Output".

- Input:** A "FIFA-21" dataset connector.
- Processors:** An "Aggregate By Country" processor followed by a "Filter By Total" processor.
- Output:** An "Output dataset" connector.

Below the pipeline diagram, a "Data preview" window shows the data for the "Aggregate By Country" processor. It displays two tables: "Input" (100 records) and "Output" (28 records).

Input				Output		
player_id*	name*	nationality*	position*	nationality*	total*	
FR Postal Codez (string)	First Name (string)	Country Code (string)	Country Code ISO2 (string)	Country (string)	Long	
158023	Lionel Messi	Argentina	ST CF RW	Uruguay	4	
20801	Cristiano Ronaldo	Portugal	ST LW	Italy	6	
190871	Neymar Jr	Brazil	CAM LW	Morocco	1	
203376	Virgil van Dijk	Netherlands	CB	Slovakia	1	
200389	Jan Oblak	Slovenia	GK	Poland	2	

On the right side, the "Aggregate By Country" processor configuration panel is visible. It shows the "Group by" field set to ".nationality" and the "Field path" set to ".nationality". The "Operation" is set to "Count".

Live preview
(before/after)

Processor configuration

Using Beam at Talend

- Started to use Beam in 2016 as ASF Incubator project for Talend DataStreams, then Talend Pipeline Designer ;
- Talend Open Source team helped Beam to become a top-level ASF project ;
- Beam is used in the Data Processing Platform for several Talend products :
 - Pipeline Designer : Batch & Streaming pipelines
 - Data Inventory : Sampling sources
 - Data Preparation : Running data pre-processing jobs

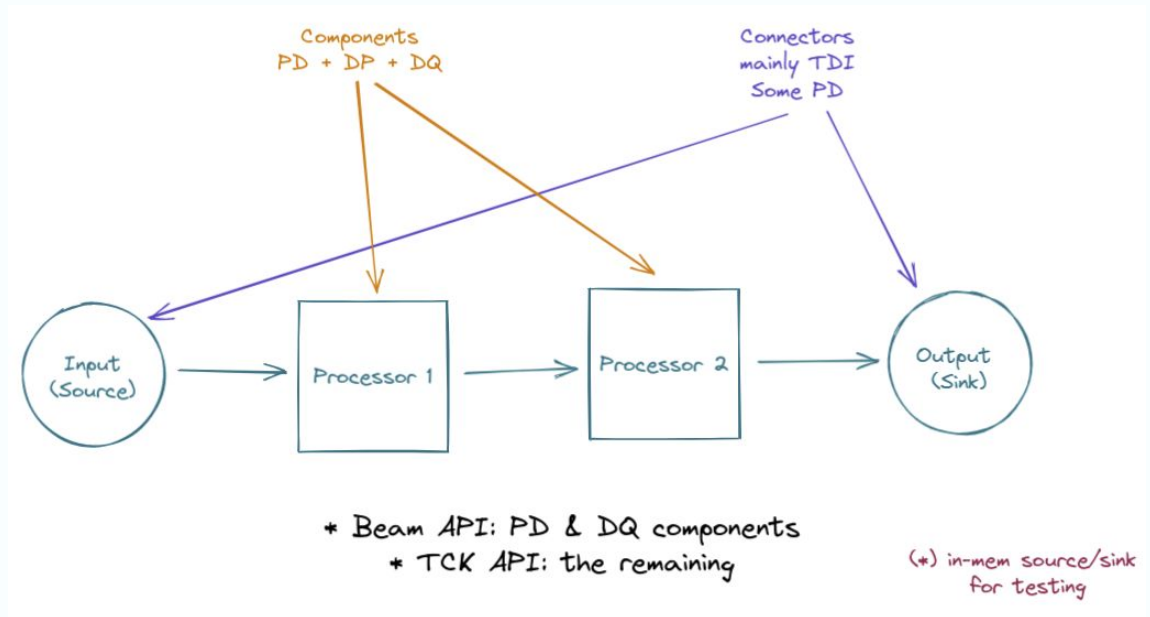
The image displays three screenshots of Talend software interfaces, illustrating the use of Apache Beam in different components:

- Pipeline Designer:** Shows a workflow titled "Pipeline Example" with steps: "FIFA-21", "Aggregate By Country", "Filter By Total", and "Output dataset". The "Aggregate By Country" step is highlighted.
- Data Inventory:** Shows a data source named "FIFA-21" with a "Talent Trust Score" of 2.86/5. It includes a radar chart showing "Validity: 84%", "Usage: 43%", "Discoverability: 25%", and "Completeness: 100%". A bar chart shows "Talent Trust Score" over time.
- Data Preparation:** Shows a table titled "FIFA-21 Preparation" with columns: player_id, name, nationality, position, overall, age, and hls. The table contains 16 rows of player data.

Engine Runtime: pipeline

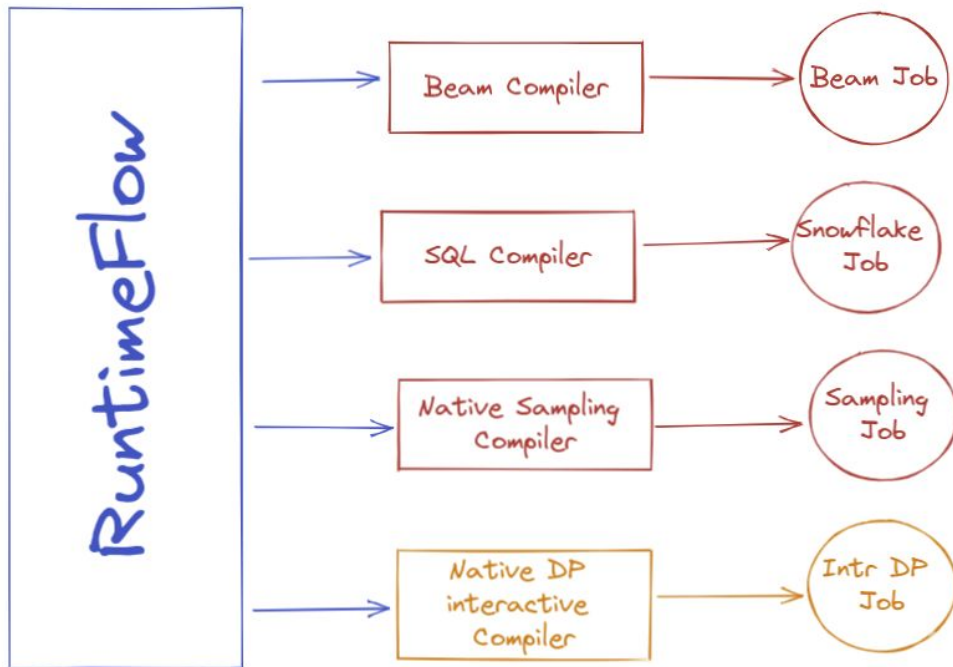
Connectors and components:

- A pipeline is essentially a DAG of components:
 - IO components: a.k.a. Connectors.
 - Intermediate components: a.k.a. Processors.
- To be used in a pipeline, connector or component have to be either:
 - Beam-based: implement Beam API (e.g, PTransform for processors)
 - TCK-based: internal components framework



Engine Runtime: compiler

A pipeline is represented as RuntimeFlow (RTF) object (JSON of components)



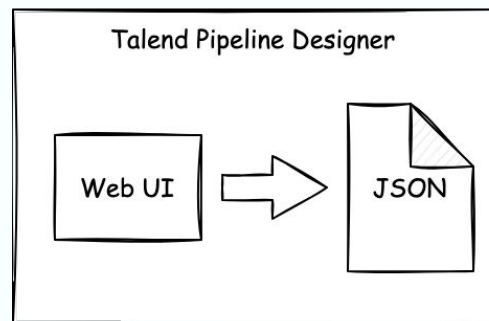
Beam Compiler (Translator):

- The first compiler that has been implemented;
- It translates an RTF to Beam pipeline;
- Then Beam pipeline is executed using either:
 - SparkRunner (Livy/FullRun job)
 - FlinkRunner (Interactive mode)
 - DirectRunner (Preview mode)

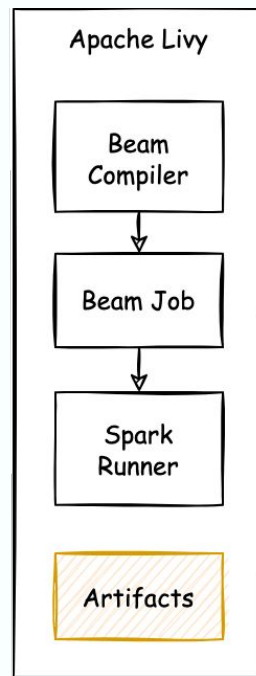
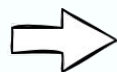
Full run Beam/Spark architecture

Example:

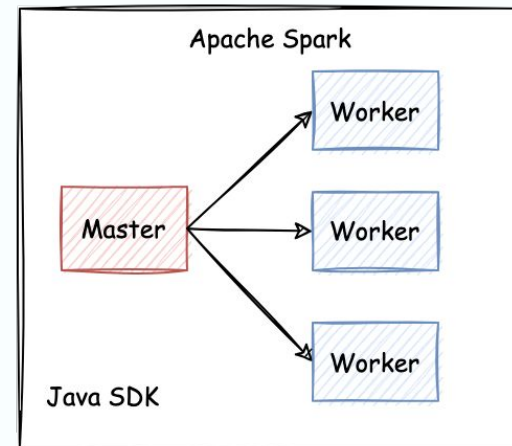
An architecture of full run job in Pipeline Designer



REST



spark-submit



Use cases: Python processor

Python processor

- TPD processor
- The Python processor executes user Python code to perform custom processing on user records.
- Originally, Python processor used Jython 2.7 as Python engine to process Python2 code

Python Component Pipeline Example

Run

Python 2 component

Python 2

Map type*

Map

Python code*

```
# Here you can define your custom MAP tra
# The input record is available as the "i
# The output record is available as the "
# The record columns are available as def
# The return statement is added automatic
# so there's no need to add it here

# Code Sample :

# 1. When choosing Map, output is a dicti

# output['col1'] = input['col1'] + 1234
# output['col2'] = "The " + input['col2']
# output['col3'] = CustomTransformationCl

# -----

# 2. When choosing FlatMap, output is a l

# recordOne = input
# recordOne['col1'] = 'newOne'
# output.append(recordOne)

output = input
output['name'] = input['name'].upper()
```

Data preview - Python 2 component

Display Both View Grid Runs on [AROMANENKO] Remote Engine Gen2

Input				Output			
player_id*	name*	nationality*	position*	player_id*	name*	nationality*	position*
FR Postal Codez (string)	First Name (string)	Country (string)	Country Co	FR Postal Codez (string)	First Name (string)	Country (string)	Country Co
158023	Lionel Messi	Argentina	ST CF	158023	LIONEL MESSI	Argentina	ST CF
28801	Cristiano Ronaldo	Portugal	ST LW	28801	CRISTIANO RONALDO	Portugal	ST LW
190871	Neymar Jr	Brazil	CAM LW	190871	NEYMAR JR	Brazil	CAM LW

Reset Save

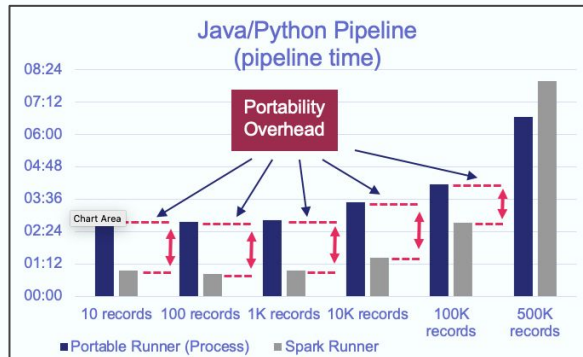
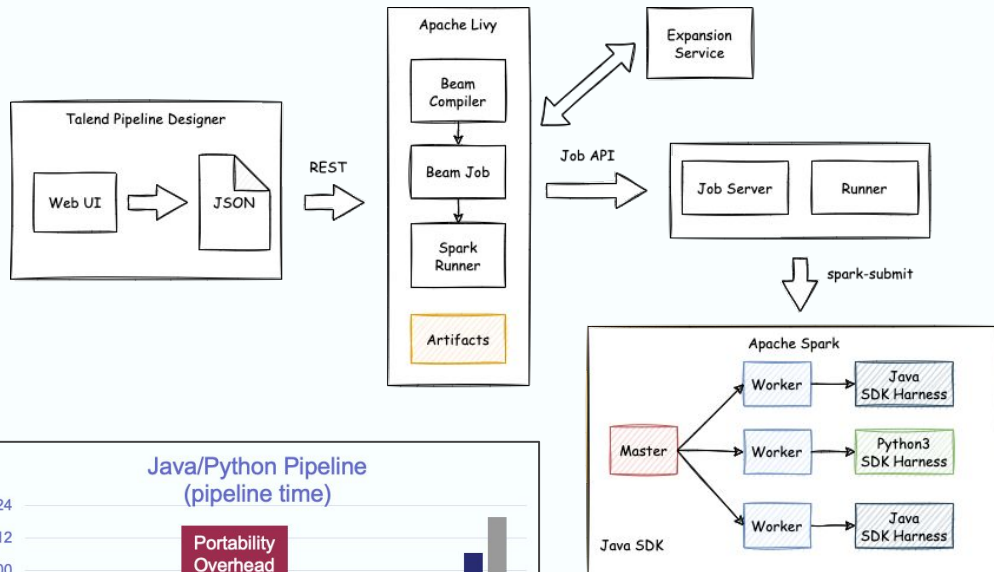
Problem:

- Python 2.7 reached EOL on 12/31/2019
- Pipeline Designer Python processor used Jython 2.7 as Python engine
- Jython didn't support Python3, no plans to support it in the future
- No easy way to install 3rd-party Python libraries

Potential solutions:

- Beam portability framework:
 - Run Python 3 code as a Beam cross-language transform with Beam Portable Runner
 - See my talk *"Using Cross-Language pipeline to run Python 3 code with Java SDK"* at Beam Summit 2020
- Python-as-Service:
 - Use a custom Python server and dedicated PTransform to execute Python code
 - Thanks to Ryan Scraba (@ryanskraba) who worked on this

Cross-language Beam/Spark



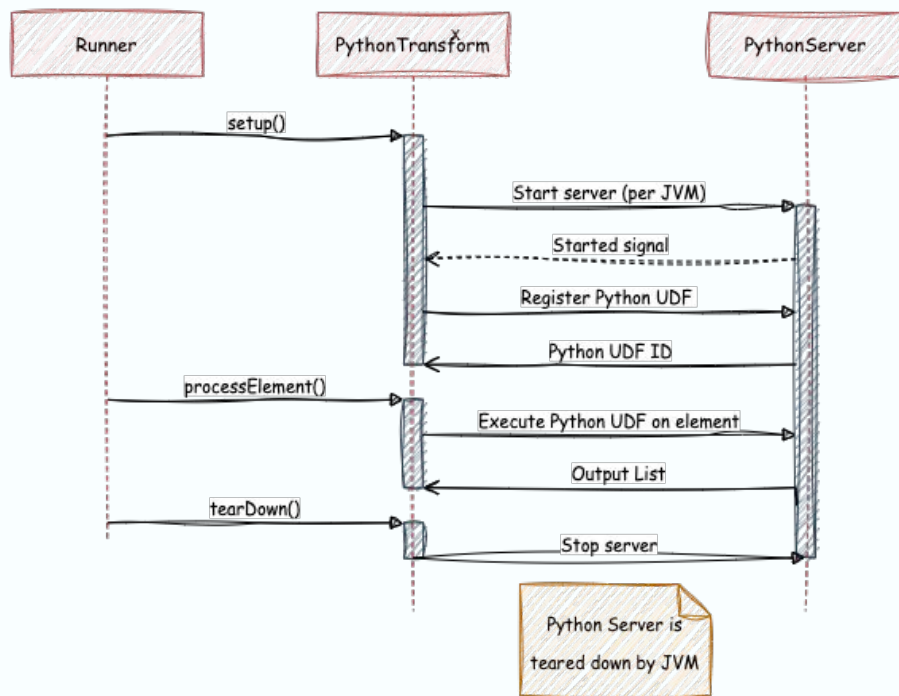
Advantages:

- Full support of Beam model and its features out-of-box;
- Tested and maintained by Beam community;
- Good performance for large data sets.

Drawbacks:

- Several times worse performance for small data;
- Required a complicated re-architecture of the TPD Runtime part
- High maintenance costs

Own Python server to execute python



Advantages:

- Simpler and configurable for our use case;
- No extra overhead/dependencies;
- Better performance for small data.

Drawbacks:

- Implementation/maintenance of the Python server;
- Only useful for specific use cases (no advanced Beam features - e.g. metrics, triggers, state, timers, etc);
- Requires a robust implementation of the Python server because of potential issues on startup/shutdown and resource leaks;
- Not tested/supported by a large community.

Use Cases: Small Data Performance

Problem:

- One pipeline (DAG/schema) → three sizes of input dataset
 - **Small dataset** (50-100 rows) for preview and interactive use;
 - **Average dataset** (~10K rows) for data sampling;
 - **Large dataset** (+10M rows) for full run pipeline.
- Fast (instant) results are critical for interactive mode
- Beam is supposed to run with large datasets and on distributed environments

Potential solutions:

- Use different runners for different use cases (current solution);
- Use native Java code compilation (PoC);
- Create Fast (In-Memory) runner for small/average datasets (PoC, WIP).

- Run a Beam pipeline (*MinimalWordCount*) locally as GraalVM native image
 - *GraalVM* is a high-performance JDK distribution designed to accelerate the execution of applications written in Java and other JVM languages along with support for a number of other popular languages.
- Use *DirectRunner* to simplify experiments
 - Other runners (*SparkRunner* & *FlinkRunner*) are in our ToDo list
- Our expectations:
 - Much lower memory usage for native images,
 - Faster startup times.

Benchmark results (*MinimalWordCount*):

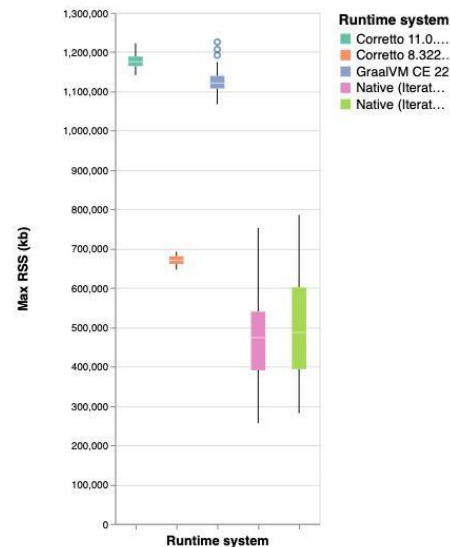
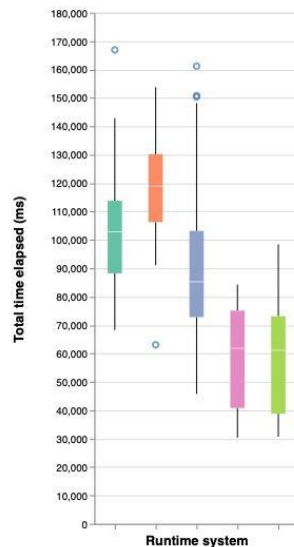
- Memory usage improved ~ 29% (median) compared to the best performing JVM
- Performance also improved ~ 27% (median) compared to the best performing JVM.

Next steps:

- Run with more performance-oriented runners, like Spark/Flink or new *Fast Local* runner

More details at Moritz Mack's blog post:

<https://github.com/mosche/blogposts/blob/main/beamnative/README.md>



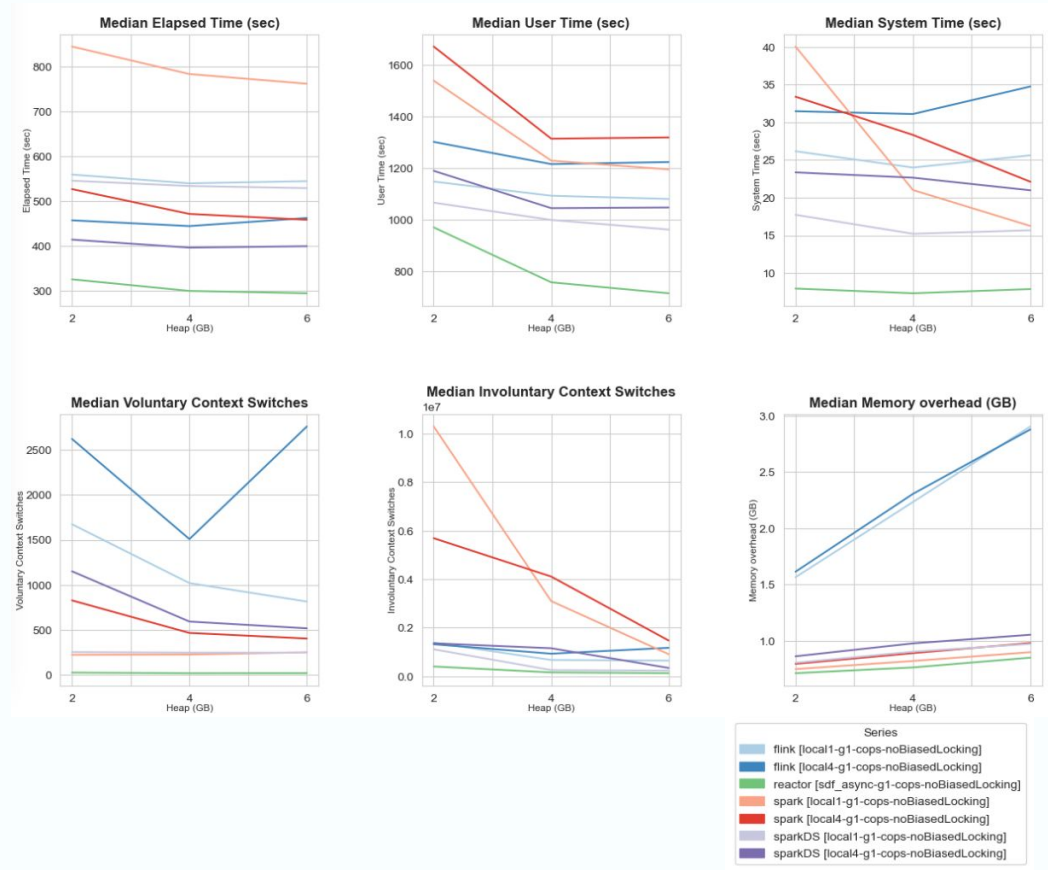
- Develop a local in-memory Beam runner from scratch;
- Replace *DirectRunner*, *FlinkRunner* and *SparkRunner* used in local mode;
- Limited Beam model implementation (at least, for PoC):
 - Batch only
 - No state / timer support
 - Global Windows only
- Use Reactive Streams (Project Reactor)
 - One JVM, keep all data in memory
 - Map Java stream operations to Beam transforms
- PoC implemented by **Moritz Mack**, early stage:
 - WIP: <https://github.com/mosche/beam/tree/reactor>

Reactor is a fourth-generation reactive library, based on the Reactive Streams specification, for building non-blocking applications on the JVM



Intermediate results

- Used Beam TPC-DS benchmarks, 10 runs for every configuration;
- No *DirectRunner*, it constantly fails with OOM errors for the TPC-DS dataset of 1GB;
- Significant performance improvements with *ReactorRunner*;
- Next steps:
 - Add Windowing support
 - Run ValidateRunner tests
 - Add Streaming support
 - Contribute back to Beam



Talend contributions to Beam

Our Beam code contributions

- Java IO connectors
 - AWS, Hadoop, Kafka, Elasticsearch, Hbase, Jdbc, Avro, Parquet, ...
- Nexmark benchmark improvements
- TPC-DS benchmark integration
- Spark Runner
 - RDD runner improvements
 - Dataset runner from scratch
- Security fixes

Other contributions

- Releases testing
- PRs reviews
- Documentation updates
- Project mailing lists discussions
- Beam users support
- Blogging and talks at conferences
 - Beam Summit, ApacheCon, OpenSource Summit, etc

Some takeaways

- It's very important to contribute back to the OS project that is a key component of your product;
- Knowledge sharing saves time and money;
- Be part of project community;
- Sometimes it's challenging to find a balance between your specific and common users requests;
- Don't wait until someone do what you need - do it yourself!

Many-many-many thanks to Talend all-time Beam contributors:

- JB Onofré (@jbonofre)
- Ismaël Mejía (@iemejia)
- Etienne Chauchot (@echauchot)
- Daniel Kulp (@dkulp)
- Ryan Skraba (@ryanskraba)
- Colm O'Heigeartaigh (@coheigea)
- Moritz Mack (@mosche)
- Romain Manni-Bucau (@rmannibucau)
- Alexey Romanenko (@aromanenko-dev)

Thank you!

QUESTIONS?

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