Unified Streaming And Batch Pipelines At LinkedIn Using Beam

Shangjin Zhang Yuhong Cheng

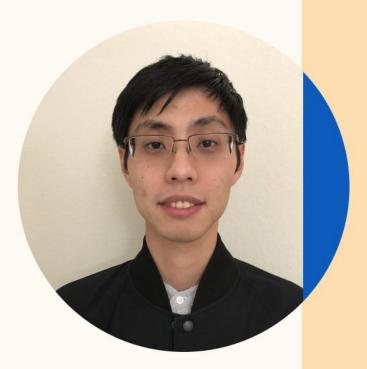
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Photo by Kari Shea on Unsplash

Hi, I am Shangjin

- CE Master @ Columbia University
- 4 Years @ Bloomberg
- 3 Years working in Al Infra @ Linkedin



Hi, I am Yuhong

- CS Master @ Rice University
- Intern @ IBM Austin
- 3 Years working in Streaming Infra @ LinkedIn



Agenda

Background: Standardization Pipelines & Backfilling

2 Problem: Backfilling Issues

1

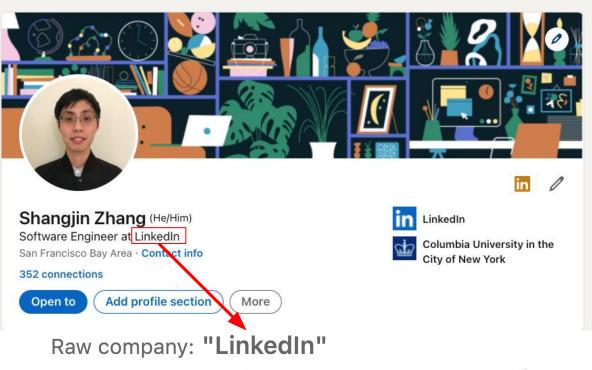
3 Solution: Unified Pipelines

4 Outcome: Performance Gains





Jobs Messagin



LinkedIn (0.96) urn:li:company:1234

Industry (name:category)

Internet:Tech (1) urn:li:industry:1 1

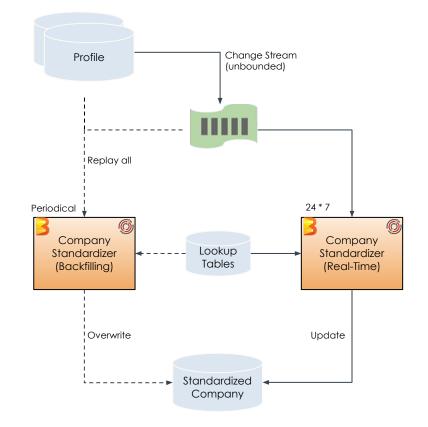
Standardization

Convert user input information into a set of pre-defined IDs

Widely used for search / model training and etc

A heavy process with NLP and deep learning models

Original Architecture



• Real-Time

- 100+ parallel streaming pipelines
- 200/sec throughput
- Apache Beam & Apache Samza
- Backfilling
 - Exactly same streaming pipelines
 - Deployed temporarily
 - 830 million member profiles
 - 40,000/sec throughput

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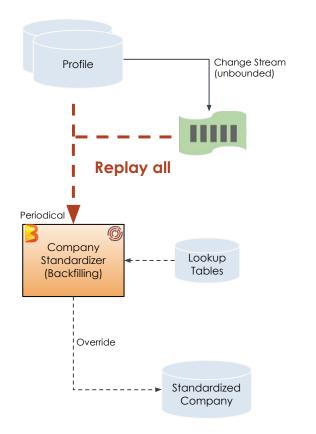
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Backfilling Issues



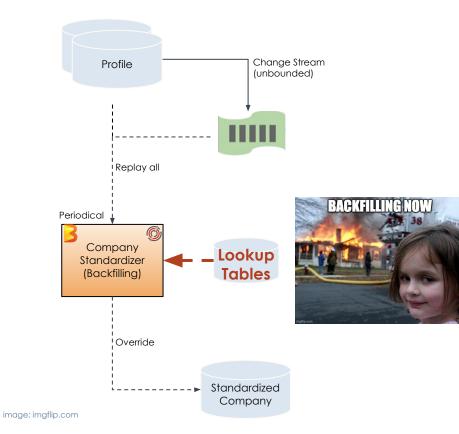
Heavy load leads to long backfilling time

- Hours to days turnaround time
- Complex model can't finish within reasonable time

Hard to scale

- Model iterates weekly instead of quarterly now
- Streaming cluster is not optimized for spiky resource footprint
- Can only host 3 concurrent backfillings

(More) Backfilling Issues



Impact on other systems

- Flood lookup tables
- Noisy neighbor to co-located streaming pipelines

Operational overhead

 Need to monitor and stop the backfilling manually

Motivation

• Run backfilling as batch job (lambda architecture)

Agenda

Background: Standardization Pipelines & Backfilling

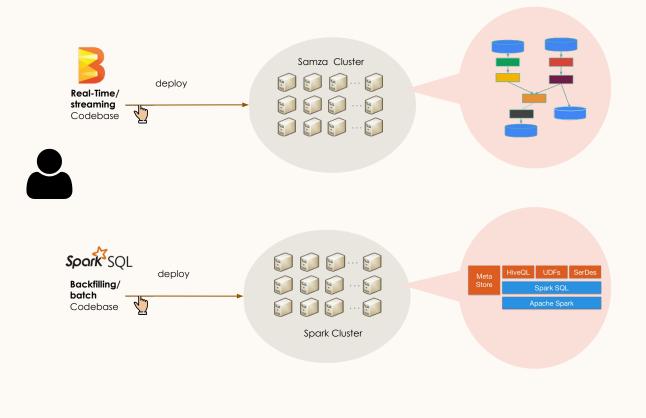
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3 Solution: Unified Pipelines

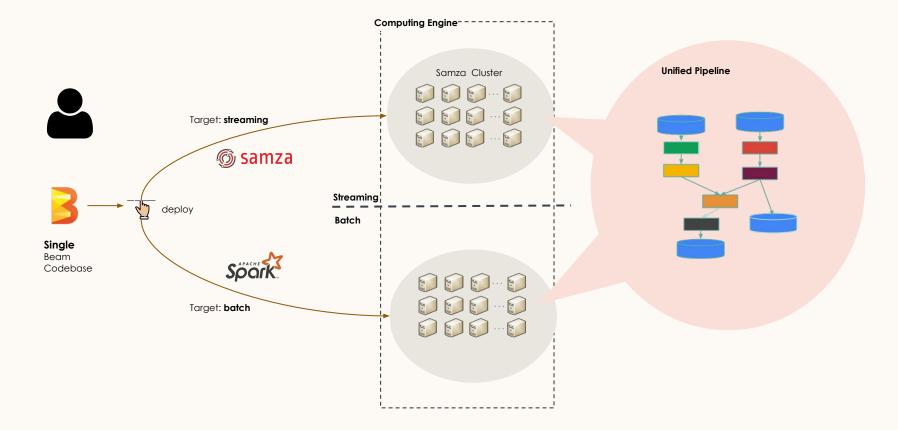
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Dropped Solution: Two Codebases





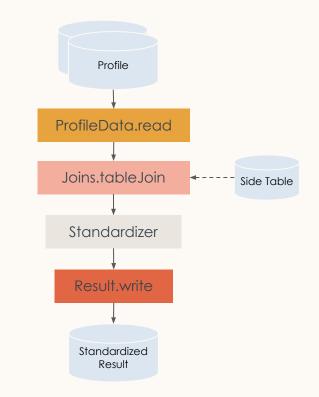
Unified Architecture



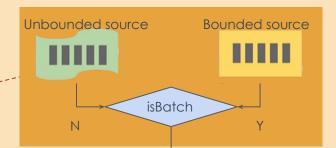
Unified Pipeline Example

PipelineOptions pipelineOpts = PipelineOptionsFactory.fromArgs(args).create(); Pipeline pipeline = Pipeline.create(pipelineOpts);

pipeline.apply(ProfileData.read())
.apply(Joins.tableJoin(sideTable))
.apply(Standardizer())
.apply(Result.write());

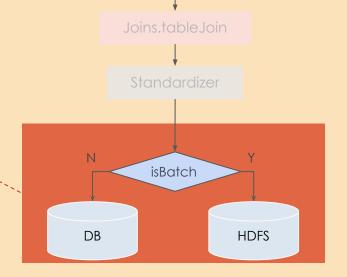


Unified IO



pipeline.apply(ProfileData.read())

.apply(Joins.tableJoin(sideTable))
.apply(Standardizer())
.apply(Result.write());



Unified PTransform

A special PTransform that provides a unified interface to users but allows different implementations according to pipeline type public static class Read extends UnifiedPTransform<PBegin,
PCollection<String>> {

```
@Override
protected PCollection<String> expandStreaming(PBegin
pBegin) {
    return pBegin.getPipeline()
        .apply(KafkalO.<String>read()
        .withTopic(getStreamingInput()))
        .apply(...);
}
```

```
@Override
```

```
protected PCollection<String> expandBatch(PBegin pBegin) {
  return pBegin.getPipeline()
    .apply(FileIO.match().filepattern(getBatchInput()))
    .apply(...);
```

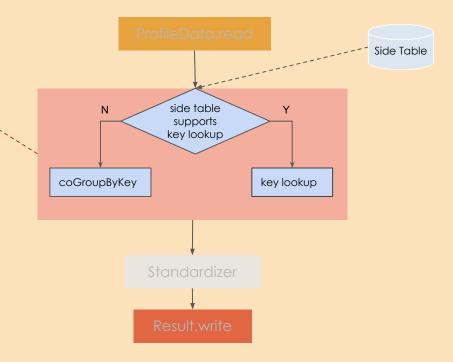
Unified Table Join

pipeline.apply(ProfileData.read())
.apply(Joins.tableJoin(sideTable))
.apply(Standardizer())
.apply(Result.write());

Provide options to do join based on the table type to avoid unnecessary data shuffling

Typically

- streaming => key lookup
- batch => coGroupByKey



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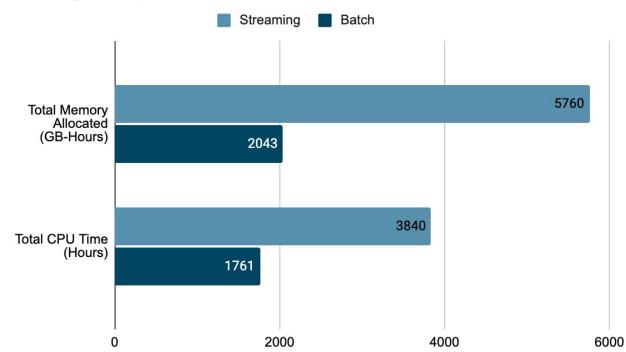
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3 Solution: Unified Pipelines

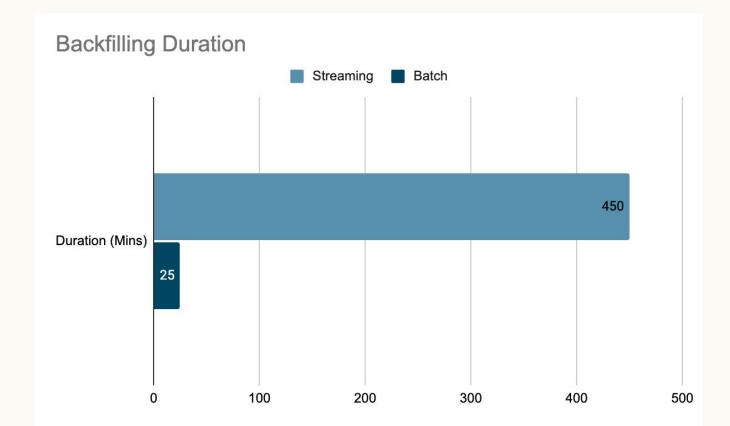
4 Outcome: Performance Gains

Benchmarks

Backfilling Computation Resources



(More) Benchmarks







Dev productivityFasterResourcesWrite code ONCESaved 94%Used ~50% lessrun everywhereprocessing timecpu time and memory

Cost to Serve Reduced ~11X cost

Future Works







More Use Cases Other Runners

Thank you

