CS 327E Class 10

Apr 30, 2021

Announcements

- Feedback on Test 3
- Testing your Beam set up
- Extra credit opportunity
- Remaining project milestones

Motivations for Dataflow

- A system for processing arbitrary computations on large amounts of data
- Can process batch data and streaming data using the same code
- Uses Apache Beam, an open-source programming model
- Designed to be very scalable, millions of QPS

Apache Beam Concepts

- A model for describing data and data processing operations:
 - Pipeline: a data processing task from start to finish
 - PCollection: a collection of data elements
 - PTransform: a data transformation operation
- Supported languages: Java, Python and Go
- Executed in the cloud on Dataflow, Spark, Flink, etc.
- Executed locally with Direct Runner for dev/testing

Beam Pipeline

- Pipeline = A directed acyclic graph where the nodes are
 PTransforms and the edges are PCollections
- General Structure of a Pipeline:
 - Reads one or more data sources as input PCollections
 - Applies one or more PTransforms on PCollections
 - Outputs resulting PCollection as one or more data sinks
- Executed as a single unit
- Runs in batch or streaming mode

PCollection

- A collection of data elements, either bounded or unbounded
- Elements can be made up primitive and complex types
- Distributed across machines
- PCollections are immutable
- Created from a data source or a PTransform
- Written to a data sink or passed to another PTransform

PTransforms

All operations on data are different kinds of PTransforms

- Element-wise:
 - o maps 1 input to (1, 0, many) outputs
 - o Examples: ParDo, Map, FlatMap
- Aggregation:
 - reduces many inputs to (1, fewer) outputs
 - o Examples: GroupByKey, CoGroupByKey, Flatten
- Composite: combines element-wise and aggregation
 - o GroupByKey -> ParDo

PTransform Properties

- Serializable
- Parallelizable
- Idempotent

ParDo Transform

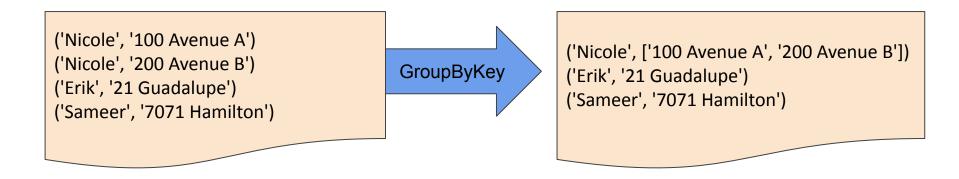
- ParDo = "Parallel Do"
- Maps 1 input to (0, 1, many) outputs
- Takes as input a PCollection
- Applies the user-defined ParDo to the input
- Outputs results as new PCollection
- Typical usage: filtering, formatting, extracting parts of data, performing computations on data elements

Hello World Example 1

```
class Multiply(beam.DoFn):
  def process(self, element):
     return [element * 10]
p = beam.Pipeline('DirectRunner', options=opts)
in pcoll = p \mid beam.Create([1, 2, 3, 4, 5])
out pcoll = in pcoll | 'Multiply' >> beam.ParDo(Multiply())
out pcoll | 'Write results' >> WriteToText('multiplied.txt')
```

GroupByKey Transform

- Input: PCollection where each element is a (key, value) pair
- Groups the values by unique key
- Output: PCollection where each element is a (key, list(value)) pair



Hello World Example 2

```
class SplitWords(beam.DoFn):
  def process(self, element):
     results = []
     words = element.split()
     for word in words:
        results.append((word, 1))
     return results
p = beam.Pipeline('DirectRunner', options=opts)
in pcoll = p | beam.Create(['Hello world', 'Hello beam users', 'beam is fun'])
split pcoll = in pcoll | 'Split Words' >> beam.ParDo(SplitWords())
out pcoll = split pcoll | 'Group Words' >> beam.GroupByKey()
```

Hands-on Exercises

- clone <u>snippets</u> repo
- step through <u>instructor beam.py</u>, <u>student beam.py</u>, and <u>class beam.py</u>
- run code samples from <u>beam_college.ipynb</u>

How to develop Beam pipelines

- 1. Start with a working code example and incrementally add to it.
- 2. Test and debug one transform at a time.
- 3. Write temporary and final PCollections to log files.
- 4. You may encounter some Jupyter notebook issues.
- 5. Start on the assignment **as early as possible**. The Beam Python documentation is **sparse** and learning Beam requires patience, perseverance, and experimentation.
- 6. Piazza won't be a good way to debug.
- 7. If you get stuck, you should go to OHs. If you can't make it to OHs, email the TAs to set up an appointment.

Milestone 3

http://www.cs.utexas.edu/~scohen/milestones/Milestone3.pdf