Pregel: A System for Large-Scale Graph Processing

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What is Pregel?

- A System for Large-Scale Graph Processing.
- An iterative and graph specific version of MapReduce.
- A distributed implementation of the Bulk Synchronous Parallel model (BSP).
- Efficient, scalable and fault-tolerant.

Graph Examples

- Web Graphs.
- Social networks.
- Transport networks.
- Similarity of newspaper articles.
- Paths of disease outbreaks (epidemiology)
- Citation relationships.

Algorithms

- Maximum Value.
- Shortest Path.
- Clustering.
- Variations of Page Rank.
- Minimum Cut.
- Connected Components.

Graph processing challenges

- Poor locality of memory access.
- Low compute to communication ratio.
- Changing degree of parallelism over the course of execution.

Previous Options

- Craft a custom distributed infrastructure.
 - Lots of effort.
 - Have to re-implement for each new algorithm or graph representation.
- Use existing distributed computing platform such as MapReduce.
 - Can lead to sub-optimal performance and usability issues.
 - Better fit would be a message passing model.
- Use graph algorithm libraries for use on a single machine.
 - Severely limits scale.
- Use existing parallel graph system.
 - No fault tolerance or support for other distributed system problems.

Pregel's solution

- Implement a scalable and fault-tolerant platform with an API that is sufficiently flexible to express arbitrary graph algorithms.
- Just like MapReduce, take care of all distributed problems behind the scenes.
- Present simple functions to be filled in by the programmer.
- Designed to be optimal for graphs.

Pregel Computation

- One Master <-> Many workers.
- Master synchronizes workers, each worker performing a computation in each
 Superstep.
- Worker's send messages between themselves.
- Iterates until all vertices vote to halt a and there are no messages in transit.





- Has a modifiable value and a list of its outgoing edges and their values.
- Only computes when active.
- All perform the same function.
 - Receives all messages sent to it in the previous superstep.
 - Performs computation.
 - Sends messages (most likely along outgoing edges).
 - Optionally vote to halt.
- Can request to add/remove vertices/edges.



Example: PageRank

```
template <typename VertexValue,
        typename EdgeValue,
        typename MessageValue>
class Vertex {
   public:
    virtual void Compute(MessageIterator* msgs) = 0;
   const string& vertex_id() const;
   int64 superstep() const;
   const VertexValue& GetValue();
   vertexValue* MutableValue();
   OutEdgeIterator GetOutEdgeIterator();
   void SendMessageTo(const string& dest_vertex,
```

```
class PageRankVertex
    : public Vertex<double, void, double> {
 public:
 virtual void Compute(MessageIterator* msgs) {
    if (superstep() >= 1) {
      double sum = 0;
      for (; !msgs->Done(); msgs->Next())
        sum += msgs->Value();
      *MutableValue() =
          0.15 / \text{NumVertices}() + 0.85 * \text{sum};
    }
    if (superstep() < 30) {</pre>
      const int64 n = GetOutEdgeIterator().size();
      SendMessageToAllNeighbors(GetValue() / n);
    } else {
      VoteToHalt();
    }
  }
};
```

Other Aspects

- Message Passing
 - Delivered in asynchronous batches using buffer .
 - No order guarantees.
- Combiners
 - Combines messages headed for destination.
 - No guarantee it will happen.
- Aggregators
 - Master can aggregate data passed to it by workers.
 - Statistics, coordination, leader assignment.
- Status Page

Other Aspects

- Graph Partitioning
 - Uses default hash on ID.
 - Can be replaced to get better locality.
- Fault tolerance
 - Check-pointing to persistent storage.
 - Failures detected using pings.
 - Frequency automatically calculated by mean time to failure model.
 - Confined recovery being looked into.

Performance

- Tested using Single Source Shortest Path Algorithm and default partitioning hash.
- Using binary tree and log-normal random graphs.
- Gives linear runtime increase for increasing graph size for both.
- Gives poorer performance for denser graphs.



Performance

• For binary tree on fixed number of machines.



Criticism

- Master is a single point of failure.
- A lot of network communication, especially for dense graphs.
- Still more limited (less expressive) than systems created later.
- Hard to partition the graph in a way that takes advantage of locality.
- Synchronicity slows all workers to the slowest worker.
- No way to redistribute load between workers.
- Performance not tested against any other systems or implementations.

Questions?