Class Website

CX4242:

Scaling Up Pig

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http://pig.apache.org

High-level language

instead of writing low-level map and reduce functions

Easier to program, understand and maintain

Created at Yahoo!

Produces sequences of Map-Reduce programs

(Lets you do "joins" much more easily)

Pig



http://pig.apache.org

Your data analysis task becomes a data flow sequence (i.e., data transformations)

Input → data flow → output

You specify data flow in Pig Latin (Pig's language). Then, Pig turns the data flow into a sequence of MapReduce jobs automatically!

Pig: 1st Benefit

Write only a few lines of Pig Latin

Typically, MapReduce development cycle is long

- Write mappers and reducers
- Compile code
- Submit jobs

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Pig: 2nd Benefit

Pig can perform a sample run on representative subset of your input data automatically!

Helps debug your code in smaller scale (much faster!), before applying on full data

What Pig is good for?

Batch processing

- Since it's built on top of MapReduce
- Not for random query/read/write

May be **slower** than MapReduce programs coded from scratch

 You trade ease of use + coding time for some execution speed

How to run Pig

Pig is a client-side application (run on your computer)

Nothing to install on Hadoop cluster

How to run Pig: 2 modes

Local Mode

- Run on your computer (e.g., laptop)
- Great for trying out Pig on small datasets

MapReduce Mode

- Pig translates your commands into MapReduce jobs
- Remember you can have a single-machine cluster set up on your computer

Pig program: 3 ways to write

Script

Grunt (interactive shell)

Great for debugging

Embedded (into Java program)

- Use PigServer class (like JDBC for SQL)
- Use PigRunner to access Grunt

Grunt (interactive shell)

Provides code completion

Press Tab key to complete Pig Latin keywords and functions

Let's see an example Pig program run with Grunt

```
records = LOAD 'input/ ncdc/ micro-tab/ sample.txt'
  AS (year:chararray, temperature:int, quality:int);
filtered records =
  FILTER records BY temperature != 9999
  AND (quality = = 0 OR quality = = 1 OR
       quality = 4 OR quality = 5 OR
       quality = = 9);
grouped records = GROUP filtered records BY year;
max temp = FOREACH grouped records GENERATE
  group, MAX (filtered records.temperature);
DUMP max temp;
```

grunt>

```
records = LOAD 'input/ncdc/micro-tab/sample.txt'
   AS (year:chararray, temperature:int, quality:int);
                            (1950, 0, 1)
 grunt> DUMP records;
                            (1950, 22, 1)
                                                 called a "tuple"
                            (1950, -11, 1)
                            (1949, 111, 1)
                             (1949, 78, 1)
 grunt> DESCRIBE records;
records: {year: chararray, temperature: int, quality: int}
```

Find highest temperature by year

```
grunt>
filtered records =
  FILTER records BY temperature != 9999
  AND (quality == 0 OR quality == 1 OR
       quality == 4 OR quality == 5 OR
       quality == 9);
grunt> DUMP filtered records;
                                      (1950, 0, 1)
                                      (1950, 22, 1)
                                      (1950, -11, 1)
                                      (1949, 111, 1)
                                      (1949, 78, 1)
```

In this example, no tuple is filtered out

Find highest temperature by year

= unordered collection of tuples

```
grunt> DESCRIBE grouped_records;
```

alias that Pig created

```
grouped_records: {group: chararray, filtered_records:
{year: chararray, temperature: int, quality: int}}
```

```
(1949, { (1949, 111, 1), (1949, 78, 1) })
(1950, { (1950, 0, 1), (1950, 22, 1), (1950, -11, 1) })
grouped_records: { group: chararray, filtered_records: { year: chararray, temperature: int, quality: int} }
```

```
grunt> max_temp = FOREACH grouped_records GENERATE
  group, MAX(filtered_records.temperature);

grunt> DUMP max_temp;

(1949,111)
  (1950,22)
```

Run Pig program on a subset of your data

You saw an example run on a tiny dataset

How to do that for a larger dataset?

 Use the ILLUSTRATE command to generate sample dataset

Run Pig program on a subset of your data

grunt> ILLUSTRATE max temp;

records	year:chararray	 	temperature:int	quality:int
	1949 1949 1949	 	78 111 9999	1 1 1
filtered_rec	ords year:chararra	 ау	temperature:int	quality:int
	1949 1949		78 111	1
grouped_records group:chararray filtered_records:bag{:tuple(year:chararray, temperature:int,quality:int)}				
	1949		{(1949, 78, 1), (1	949, 111, 1)}
max_temp	group:chararray		:int	
	1949		111	

How does Pig compare to SQL?

SQL: "fixed" schema

PIG: loosely defined schema, as in

```
records = LOAD 'input/ncdc/micro-tab/sample.txt'
AS (year:chararray, temperature:int, quality:int);
```

How does Pig compare to SQL?

SQL: supports fast, random access (e.g., <10ms, but of course depends on hardware, data size, and query complexity too)

PIG: batch processing

Pig vs SQL

- 1. Pig Latin is procedural, where SQL is declarative.
- 2. Pig Latin allows pipeline developers to decide where to checkpoint data in the pipeline.
- 3. Pig Latin allows the developer to select specific operator implementations directly **rather than relying on the optimizer**.
- 4. Pig Latin supports **splits** in the pipeline.
- 5. Pig Latin allows developers to insert their own code almost anywhere in the data pipeline.

Much more to learn about Pig

Relational Operators, Diagnostic Operators (e.g., describe, explain, illustrate), utility commands (cat, cd, kill, exec), etc.

Table 11-1. Pig Latin relational operators

Category	Operator	Description	
Loading and storing	LOAD	Loads data from the filesystem or other storage into a relation	
	STORE	Saves a relation to the filesystem or other storage	
	DUMP	Prints a relation to the console	
Filtering	FILTER	Removes unwanted rows from a relation	
	DISTINCT	NCT Removes duplicate rows from a relation	
	FOREACHGENERATE Adds or removes fields from a relation		
	MAPREDUCE	Runs a MapReduce job using a relation as input	
	STREAM	Transforms a relation using an external program	
	SAMPLE	Selects a random sample of a relation	
Grouping and joining	JOIN	Joins two or more relations	
	COGROUP	Groups the data in two or more relations	
	GROUP	Groups the data in a single relation	
	CROSS	Creates the cross-product of two or more relations	
Sorting	ORDER	Sorts a relation by one or more fields	
	LIMIT	Limits the size of a relation to a maximum number of tuples	
Combining and splitting	UNION	Combines two or more relations into one	
	SPLIT	Splits a relation into two or more relations	