L08: Just enough Scala for Spark

ANLY 502: Massive Data Fundamentals Simson Garfinkel & Marck Vaisman March 20, 2017



GEORGETOWN UNIVERSITY

Agenda

Administrivia

- A04 not graded yet
- A05 will be posted mid-week (hopefully)
- Virtual Machine tools (useful for learning and playing with new tools)
 - Vagrant
 - Docker

Brief overview Functional programming concepts

- Just Enough Scala for Spark
 - Basic Scala data types
 - Methods and Functions
 - RDD Operations
- Lab play with the Dean Wampler's Spark Notebook and look at non-trivial Scala code
 - Walk through steps involved in creating an inverted index
- Time Permitting Brief overview of PIG





https://codesachin.files.wordpress.com/2016/04/220px-function_machine2-svg.png

Technical functional code characteristics:

- immutable data
- first class functions
- tail call

Functional code is characterized by:

- the absence of side effects
- not relying on data outside the current function
- not changing the data outside of the current function

https://maryrosecook.com/blog/post/a-practical-introduction-to-functionalprogramming

This is an unfunctional function:

```
a = 0
def increment():
    global a
    a += 1
```

This is a functional function:

```
def increment(a):
    return a + 1
```

```
name_lengths = map(len, ["Mary", "Isla", "Sam"])
```

print name_lengths
=> [4, 4, 3]

This is a map that squares every number in the passed collection:

```
squares = map(lambda x: x * x, [0, 1, 2, 3, 4])
```

print squares
=> [0, 1, 4, 9, 16]

import random

```
names = ['Mary', 'Isla', 'Sam']
code_names = ['Mr. Pink', 'Mr. Orange', 'Mr. Blonde']
```

```
for i in range(len(names)):
    names[i] = random.choice(code_names)
```

```
print names
# => ['Mr. Blonde', 'Mr. Blonde', 'Mr. Blonde']
```

This code counts how often the word 'sam' appears in a list of strings:

```
sentences = ['Mary read a story to Sam and Isla.',
                'Isla cuddled Sam.',
                'Sam chortled.']
sam_count = 0
for sentence in sentences:
        sam_count += sentence.count('Sam')
print sam_count
```

=> 3

This is the same code written as a reduce:

```
sentences = ['Mary read a story to Sam and Isla.',
    'Isla cuddled Sam.',
    'Sam chortled.']
sam_count = reduce(lambda a, x: a + x.count('Sam'),
        sentences,
        0)
```

Write declaratively, not imperatively

Imperative

```
from random import random
```

```
time = 5
car_positions = [1, 1, 1]
```

```
while time:
```

```
# decrease time
time -= 1
```

```
print ''
```

```
for i in range(len(car_positions)):
    # move car
    if random() > 0.3:
        car positions[i] += 1
```

```
# draw car
print '-' * car_positions[i]
```

```
from random import random
```

```
def move_cars():
    for i, _ in enumerate(car_positions):
        if random() > 0.3:
            car_positions[i] += 1
```

```
def draw_car(car_position):
    print '-' * car_position
```

```
def run_step_of_race():
    global time
    time -= 1
    move cars()
```

```
def draw():
    print ''
    for car_position in car_positions:
        draw_car(car_position)
```

```
time = 5
car_positions = [1, 1, 1]
```

```
while time:
    run_step_of_race()
    draw()
```

This is a functional version of the car race code:

```
from random import random
def move_cars(car_positions):
    return map(lambda x: x + 1 if random() > 0.3 else x,
               car positions)
def output car(car position):
    return '-' * car position
def run_step_of_race(state):
    return {'time': state['time'] - 1,
            'car positions': move cars(state['car positions'])}
def draw(state):
    print ''
    print '\n'.join(map(output car, state['car positions']))
def race(state):
    draw(state)
    if state['time']:
        race(run step of race(state))
race({'time': 5,
      'car positions': [1, 1, 1]})
```

The code is still split into functions, but the functions are functional. There are three signs of this. First, there are no longer any shared variables. time and car_positions get passed straight into race(). Second, functions take parameters. Third, no variables are instantiated inside functions. All data changes are done with return values. race() recurses³ with the result of run_step_of_race(). Each time a step generates a new state, it is passed immediately into the next step.

Scala

Why Scala?

Non-Technical

Spark is written in Scala

Most new features will be available trough the Scala API first before Python or R

There is industry demand

Technical

Statically typed

Mixed paradigm - object oriented programming

Mixed paradigm - functional programming

Sophisticated type system

Succinct, elegant, flexible syntax

Scalable

References used for this lecture



https://www.youtube.com/watch?v=LBoSgiLV_NQ

https://maryrosecook.com/blog/post/a-practical-introduction-to-functionalprogramming

Everything in Scala is an object

- Objects have states and behaviors. An object is an instance of a class
 - -Objects have types, the type determines what we can do with the object.
 - -Example: numbers, strings, files, digital images
- A Class can be defined as a template/blueprint that describes the behavior/state of an object.
- A Method is a behavior. A Class can contain many methods.

Scala has all the same data types as Java, with the same memory footprint and precision. Following is the table giving details about all the data types available in Scala:

Data Type	Description
Byte	8 bit signed value. Range from -128 to 127
Short	16 bit signed value. Range -32768 to 32767
Int	32 bit signed value. Range -2147483648 to 2147483647
Long	64 bit signed value9223372036854775808 to 9223372036854775807
Float	32 bit IEEE 754 single-precision float
Double	64 bit IEEE 754 double-precision float
Char	16 bit unsigned Unicode character. Range from U+0000 to U+FFFF
String	A sequence of Chars
Boolean	Either the literal true or the literal false
Unit	Corresponds to no value
Null	null or empty reference
Nothing	The subtype of every other type; includes no values
Any	The supertype of any type; any object is of type Any
AnyRef	The supertype of any reference type

All the data types listed above are objects. There are no primitive types like in Java. This means that you can call methods on an Int, Long, etc.

Variables

val - immutable variable

val m = 17 m = 18 // error!

res1: Array[String] = Array(Hello, null, null, null, null)

var - mutable variable (but can only be reassigned with another value of same type)

```
// src/main/scala/progscala2/typelessdomore/person.sc
scala> class Person(val name: String, var age: Int)
defined class Person
scala> val p = new Person("Dean Wampler", 29)
p: Person = Person@165a128d
scala> p.name
res0: String = Dean Wampler
                                  // Show the value of firstName.
scala> p.age
                                  // Show the value of age.
res2: Int = 29
scala> p.name = "Buck Trends"
<console>:9: error: reassignment to val
                                           // Disallowed!
       p.name = "Buck Trends"
               Λ
scala> p.age = 30
                                  // Okay!
p.age: Int = 30
```



The var and val keywords only specify whether the reference can be changed to refer to a different object (var) or not (val). They don't specify whether or not the object they reference is mutable.

Ranges

```
scala> 1 to 10  // Int range inclusive, interval of 1, (1 to 10)
res0: scala.collection.immutable.Range.Inclusive =
    Range(1, 2, 3, 4, 5, 6, 7, 8, 9, 10)
```

```
scala> 1 until 10 // Int range exclusive, interval of 1, (1 to 9)
res1: scala.collection.immutable.Range = Range(1, 2, 3, 4, 5, 6, 7, 8, 9)
```

```
scala> 1 to 10 by 3 // Int range inclusive, every third.
res2: scala.collection.immutable.Range = Range(1, 4, 7, 10)
```

```
scala> 10 to 1 by -3 // Int range inclusive, every third, counting down.
res2: scala.collection.immutable.Range = Range(10, 7, 4, 1)
```

```
scala> 1L to 10L by 3 // Long
res3: scala.collection.immutable.NumericRange[Long] = NumericRange(1, 4, 7, 10)
```

```
scala> 1.1f to 10.3f by 3.1f // Float with an interval!= 1
res4: scala.collection.immutable.NumericRange[Float] =
    NumericRange(1.1, 4.2, 7.2999997)
```

```
scala> 1.1f to 10.3f by 0.5f // Float with an interval < 1
res5: scala.collection.immutable.NumericRange[Float] =
    NumericRange(1.1, 1.6, 2.1, 2.6, 3.1, 3.6, 4.1, 4.6, 5.1, 5.6, 6.1, 6.6,
        7.1, 7.6, 8.1, 8.6, 9.1, 9.6, 10.1)</pre>
```

```
scala> 1.1 to 10.3 by 3.1 // Double
res6: scala.collection.immutable.NumericRange[Double] =
```

```
NumericRange(1.1, 4.2, 7.30000000000000)
```

```
scala> 'a' to 'g' by 3 // Char
res7: scala.collection.immutable.NumericRange[Char] = NumericRange(a, d, g)
```

```
scala> BigInt(1) to BigInt(10) by 3
res8: scala.collection.immutable.NumericRange[BigInt] =
   NumericRange(1, 4, 7, 10)
```

```
scala> BigDecimal(1.1) to BigDecimal(10.3) by 3.1
res9: scala.collection.immutable.NumericRange.Inclusive[scala.math.BigDecimal]
= NumericRange(1.1, 4.2, 7.3)
```

Statically typed languages can be very verbose. Consider this typical declaration in Java, before Java 7:

import java.util.HashMap;

HashMap<Integer, String> intToStringMap = new HashMap<Integer, String>();

We have to specify the type parameters <Integer, String> twice. Scala uses the term type annotations for explicit type declarations like HashMap<Integer, String>.

Java 7 introduced the diamond operator to infer the generic types on the righthand side, reducing the verbosity a bit:

HashMap<Integer, String> intToStringMap = new HashMap<>();

We've already seen some examples of Scala's support for type inference. The compiler can discern quite a bit of type information from the context, without explicit type notations. Here's the same declaration rewritten in Scala, with inferred type information:

val intToStringMap: HashMap[Integer, String] = new HashMap

If we specify HashMap[Integer, String] on the righthand side of the equals sign, it's

even more concise:

val intToStringMap2 = new HashMap[Integer, String]

Some functional programming languages, like Haskell, can infer almost all types, be- cause they can perform global type inference. Scala can't do this, in part because Scala has to support *subtype polymorphism* (inheritance), which makes type inference much harder.

Here is a summary of the rules for when explicit t

Literal Values

Integer - type Int

Floating Point - type Float

0
035
21
ØxFFFFFF
0777L

0.0		
1e30f		
3.14159f		
1.0e100		
.1		

Character - single character, single quote

String - sequence of characters, double quote

'a'
'\u0041'
'\n'
'\t'

"Hello,\nWorld!"		
"This string contains	a \"	character."

Multi Line Strings

"""the present string	
spans three	
lines """	

Multiple assignments

Tuples combine a fixed number of items together so that they can be passed around as a whole. Unlike an array or list, a tuple can hold objects of different types but they are immutable.

The following is an example of a tuple holding an integer, a string, and the console.

val t = (1, "hello", Console)

Which is syntactic sugar (shortcut) for the following:

val t = new Tuple3(1, "hello", Console)

Accessing tuple elements

val t = (4,3,2,1)

To access elements of a tuple t, you can use method t._1 to access the first element, t._2 to access the second, and so on. For example, the following expression computes the sum of all elements of t.

val sum = $t._1 + t._2 + t._3 + t._4$

What is an RDD

An immutable, partitioned collection of elements that can be operated on in parallel. This class contains the basic operations available on all RDDs, such as map, filter, and persist. In addition, PairRDDFunctions contains operations

available only on RDDs of key-value pairs, such as groupByKey and join; DoubleRDDFunctions contains operations available only on RDDs of Doubles; and SequenceFileRDDFunctions contains operations available on RDDs that can be saved as SequenceFiles. All operations are automatically available on any RDD of the right type (e.g. RDD[(Int, Int)] through implicit.

Internally, each RDD is characterized by five main properties:

- A list of partitions - A function for computing each split - A list of dependencies on other RDDs - Optionally, a Partitioner for key-value RDDs (e.g. to say that the RDD is hash-partitioned) - Optionally, a list of preferred locations to compute each split on (e.g. block locations for an HDFS file)

All of the scheduling and execution in Spark is done based on these methods, allowing each RDD to implement its own way of computing itself. Indeed, users can implement custom RDDs (e.g. for reading data from a new storage system) by overriding these functions. Please refer to the Spark paper for more details on RDD internals.

RDDs

- RDDs support many of the operations supported by native Scala collections RDDs are immutable
 - Cannot change an RDD once created. All operations create new RDDs or other Scala objects

RDDs are lazy

 Unlike native Scala collections, RDD operations are only evaluated when needed. (In the REPL any operation on a collection prints the values of the new collection to screen.

Transformations on RDDs: create new RDD from current one. Lazy evaluation Actions on RDDs: force the evaluation of an RDD and normally return a Scala object rather than an RDD. Actions are evaluated immediately.

List of Transformations and Actions

Some Scala Examples

https://gist.github.com/mmakowski/379028 https://gist.github.com/mmakowski/379031 https://github.com/dcsobral/ConwayLife

```
6
    val iiFirstPass1 = sc.wholeTextFiles(shakespeare.toString).
7
        flatMap { location_contents_tuple2 => // two-element tuple: (filename, contents)
            val words = location_contents_tuple2._2.split("""\\\\\\\\\""")
 8
            val fileName = location_contents_tuple2._1.split(File.separator).last
 9
            words.map(word => ((word, fileName), 1)) // two-element tuple with two-elem. tuple key!
10
11
        3.
        // The previous and next expressions are effectively this query:
12
13
        // "SELECT word, COUNT(*) FROM ... GROUP BY word"
14
        reduceByKey((count1, count2) => count1 + count2).
        map { word_file_count_tup3 => // Setup the word as the key, not the (word,filename)
15
16
            (word_file_count_tup3._1._1, (word_file_count_tup3._1._2, word_file_count_tup3._2))
17
            // output: (word, (filename, count))
18
        3.
19
        groupByKey. // group by words
        sortByKey(ascending = true).
20
        mapValues { iterable =>
21
22
            val vect = iterable.toVector.sortBy { file_count_tup2 =>
23
                (-file_count_tup2._2, file_count_tup2._1) // sort by count descending, then file
24
25
            vect.mkString(",") // make a comma-separated string of the (file,count) pairs
26
```

Coming Up

- A05 Due Friday 3/31
- Q07 and Q08
- Project proposals due 3/22
- L09 Scalable Machine Learning



Apache Pig

Apache Pig

Started at Yahoo! Research

- Easier approach for MapReduce
- Procedural language
- PigLatin scripts *interpreted and run as* MapReduce jobs.

Pig Advantages:

- Easier to program than MapReduce.
- Declarative statements directly describe data transformations.
- Optimizer makes efficient decisions.
- Debugging operators:
 - *–DESCRIBE, EXPLAIN, ILLUSTRATE*
- Can run "locally" or on Hadoop.

Pig Disadvantages:

- Simple statements may generate many MapReduce jobs.
- Can be hard to debug.
- Keywords are case insensitive -LOAD, USING, AS, GROUP, BY, ...
- Functions, relations, fields are case sensitive:

-PigStorage, COUNT,

Christopher Olston* Benjamin Reect Yahool Research Ravi Kumar* Ravi Rumar* Yahool Research Au sa growing need for ad-boc analysis of extremely ta sets, especially at internet companies where innor- tically depends on being able to analyze terability collected every day. Parallel database products, e.g., a, offer a solution, but are usually prohibitivey at this scale. Besides, many of the people who ana- st ata are enterneed procentum programmers, and the internet stata seventum programmers, and the internet programmers, and the internet stata seventum programmers, and the internet programmers, and the programmers, and the internet programmers, and the internet programmers, and the programers, and the programmers, and the programmers, and the pr	t Utkarsh Srivastava [‡] vahoo! Research vahoo! Research NTRODUCTION growing number of organizations, innovation revolves	Alan F. Gates, Olga Natkovich, Sh Shravan M. Narayanamurthy, Ch Santhosh Srinivasar Yahool	ubham Chopra, Pradeep Kamath, istopher Olston, Benjamin Reed, , Utkarsh Srivastava	Pig Latin Reference Manual 2
RACT I. It is a growing need for ad-hoc analysis of extremely ta sets, especially at internet companies where inno- ritically depends on being able to analyze transforms such collected every day. Parallel database products, e.g., a, offer a solution, but are usually prohibitively com- at this scale. Besides, many of the people who ana- st ata are enterneched procedural programmers, we exam	NTRODUCTION growing number of organizations, innovation revolves	ranoo	laa '	
RACT 1. 1 s a growing need for ad-hoc analysis of extremely tasts, especially at internet companies where inno- ritically depends on being able to analyze terabytes collected every day. Parallel database products, e.g., a, offer a solution, but are usually prohibitively even at this scale. Besides, many of the people who ana- sida are enterneched procedurul programmers, we evan	NTRODUCTION a growing number of organizations, innovation revolves	A DOWN A CON	, inc.	
collected every day. Parallel database products, e.g., a, offer a solution, but are usually prohibitively ex- at this scale. Besides, many of the people who ana- the in s data are entrenched procedural programmers, who	d the collection and analysis of enormous data sets is web crawls, search logs, and click streams. Inter-	ABSTRACT Increasingly, organizations capture, transform and analyze enormous data sets. Prominent examples include internet companies and e-science. The Map-Reduce scalable dataflow paradium has become novules for these analyzistics. Its	records; extracted entity-relationship models). The process- ing combines generic relational-style operations (e.g., fil- ter; join; count) with specialized domain-specific operations (e.g., part-of-speech tagging; face detection). A similar sit- uation arises in e-science, national intelligence, and other	by Table of contents
s data are entrenched procedural programmers, who examp	mpanies such as Amazon, Google, Microsoft, and Ya- are prime examples. Analysis of this data constitutes nermost loop of the product improvement cycle. For	simple, explicit dataflow programming model is favored by some over the traditional high-level declarative approach:	domains. The popular Map-Reduce [8] scalable data processing frame-	1 Overview
algorithm SOL stule to be unnetwork The sussess algori	ble, the engineers who develop search engine ranking	SQL. On the other hand, the extreme simplicity of Map- Reduce leads to much low-level hacking to deal with the	work, and its open-source realization <i>Hadoop</i> [1], cater to these workloads and offer a simple dataflow programming	2 Data 1 ypes and More
re procedural map-reduce programming model, and	g for exploitable trends.	many-step, branching dataflows that arise in practice. More- over, users, must repeatedly code standard operations such	model that appeals to many users. However, in practice, the extreme simplicity of the Map-Reduce programming model	4 Palational Operators
evidence of the above. However, the map-reduce and p	rocessed on highly parallel systems, such as shared-	as join by hand. These practices waste time, introduce bugs,	leads to several problems. First, it does not directly sup- port complex N-step dataflows, which often arise in prac-	5 Diagnostic Operators
is too low-level and rigid, and leads to a great deal nothin user code that is hard to maintain, and reuse. Oracle	g clusters. Parallel database products, e.g., Teradata, e RAC, Netezza, offer a solution by providing a simple	Pig is a high-level dataflow system that aims at a sweet	tice. Map-Reduce also lacks explicit support for combined	6 UDF Statements 9
cribe a new language called Pig Latin that we have to fit in a sweet spot between the declarative style ical cl	uery interface and hiding the complexity of the phys- uster. These products however, can be prohibitively	spot between SQL and Map-Reduce. Pig offers SQL-style high-level data manipulation constructs, which can be as-	matching operations), a crucial aspect of knowledge discov-	7 Eval Functions
nd the low-level, procedural style of map-reduce. expen manying system Pig is fully implemented and away	sive at web scale. Besides, they wrench programmers rom their preferred method of analyzing data, namely	sembled in an explicit dataflow and interleaved with custom Map- and Reduce-style functions or executables. Pig pro-	ery. Lastly, irequently-needed data manipulation primitives like filtering, aggregation and top-k thresholding must be	8 Load/Store Functions
Pig Latin into physical plans that are executed writin	g imperative scripts or code, toward writing declara- ueries in SOL, which they often find unnatural and	grams are compiled into sequences of Map-Reduce jobs, and executed in the Hadoon Man-Reduce environment. Both Pig	coded by hand. Consequently, users end up stitching together Map-Reduce	9 Math Functions
few examples of how engineers at Yahoo! are using overly	restrictive.	and Hadoop are open-source projects administered by the	dataflows by hand, hacking multi-input flows, and repeat- edly implementing standard operations inside black-box func-	10 String Functions
matically reduce the time required for the develop- execution of their data analysis tasks, compared to ing to	the more procedural map-reduce [4] programming	Apache Software Foundation. This paper describes the challenges we faced in develop-	tions. These practices slow down data analysis, introduce	11 Bag and Tuple Functions
oop directly. We also report on a novel debugging model ent that comes integrated with Pig. that can lead by-ag	. A map-reduce program essentially performs a group- gregation in parallel over a cluster of machines. The	ing Pig, and reports performance comparisons between Pig execution and raw Map-Reduce execution.	and impede automated optimization.	12 File Commands
higher productivity gains. Pig is an open-source, progra	ammer provides a map function that dictates how the		Our Pig system [4] offers composable high-level data ma- nipulation constructs in the spirit of SQL, while at the same	13 Shell Commands
in and Subject, and available for general dec.	gregation. What is appealing to programmers about	1. INTRODUCTION	time retaining the properties of Map-Reduce systems that make them attractive for certain users, data types, and	14 Utility Commands
tes and Subject Descriptors: units in tabase Management: Languages primi Terms: Languages. can b	virtices in a three are only two ingraved accurative ives (map and reduce) to enable parallel processing, e rest of the code, i.e., the map and reduce functions, written in any programming language of choice, and	organizations increasingly rely on utra-large-scale data processing in their day-to-day operations. For example, modern internet companies routinely process petabytes of web content and usage logs to populate search indexes and	workloads. In particular, as with Map-Reduce, Pig pro- grams encode explicit dataflow graphs, as opposed to im- plicit dataflow as in SQL. As one user from Adobe put it:	
yahoo-inc.com Und	ortunately, the map-reduce model has its own set of	perform ad-hoc mining tasks for research purposes. The data includes unstructured elements (e.g., web page text;	"Pig seems to give the necessary parallel pro-	
Nyahoo-inc.com limita rigid.	To perform tasks having a different data flow, e.g.,	images) as well as structured elements (e.g., web page click	gramming constructs (FOREACH, FLATTEN, COGROUP etc) and also give sufficient control	
a@yahoo-inc.com joins Also,	or n stages, inelegant workarounds have to be devised. custom code has to be written for even the most com-	*Author email addresses: {gates, olgan, shubhamc, pradeepk shrawanm olston breed sms utkarsh}	back to the programmer (which a purely declara- tive approach like [SQL on top of Map-Reduce] ¹	
@yahoo-inc.com mon c lead t	perations, e.g., projection and filtering. These factors a code that is difficult to reuse and maintain, and in	@yahoo-inc.com.	doesn't)."	
which	the semantics of the analysis task are obscured. More-		Pig dataflows can interleave built-in relational-style op-	
to make digital or hard copies of all or part of this work for classroom use is granted without fee provided that copies are	es the ability of the system to perform optimizations.	Permission to copy without fee all or part of this material is granted provided that the copies are not made or distributed for direct commercial advantage,	erations like liter and join, with user-provided executables (scripts or pre-compiled binaries) that perform custom pro-	
e distributed for profit or commercial advantage and that copies are we to the full citation on the first page and that copies to combi	have developed a new language called <i>Pig Latin</i> that nes the best of both worlds: high-level declarative	the VLDB copyright notice and the title of the publication and its date appear, and notice is given that copying is by permission of the Very Large Data	cessing. Schemas for the relational-style operations can be supplied at the last minute, which is convenient when work-	
post on servers or to redistribute to lists, requires prior specific und/or a fee	ng in the spirit of SQL, and low-level, procedural pro- ning à la map-reduce.	Base Endowment. To copy otherwise, or to republish, to post on servers or to redistribute to lists, requires a fee and/or special permission from the	ing with temporary data for which system-managed meta- data is more of a burden than a bonefit. For data used	Convrint © 2007 The Anache Software Foundation. All rights received
8, June 9–12, 2008, Vancouver, BC, Canada. 008 ACM 978-1-60558-102-6/08/06\$5.00.	· ·	publisher, ACM. VLDB '09, August 24-28, 2009, Lyon, France	¹ Deference to expecting entrying project removed	obpyright o 2007 the Apache Sottware Foundation. Air rights reserved.
		Copyright 2009 VLDB Endowment, ACM 000-0-00000-000100.	nezenice to specific software project removed.	

MASSIVE DATA FUNDAMENTALS

Famous example: Pig program to find top 5 websites for Twitter users age 18-25

- Users = load 'users' as (name, age);
 Filtered = filter Users by age >= 18 and age <=
 25;</pre>
- Pages = load 'pages' as (user, url);
- Joined = join Filtered by name, Pages by
 user;
- Grouped = group Joined by url;
- Sorted = order Summed by clicks desc;
- Top5 = limit Sorted 5;

store Top5 into 'top5sites';

Equivalent MapReduce program (in Java)

import java.util.ArrayList; import java.util.Iterator; import java.util.List; import org.apache.hadoop.fs.Path; import org.apache.hadoop.io.LongWritable; import org.apache.hadoop.io.Text; import org.apache.hadoop.io.Writable; import org.apache.hadoop.io.WritableComparable; import org.apache.hadoop.mapred.FileInputFormat; import org.apache.hadoop.mapred.FileOutputFormat; import org.apache.hadoop.mapred.JobConf; import org.apache.hadoop.mapred.JobConf; import org.apache.hadoop.mapred.KeyValueTextInputFormat; import org.apache.hadoop.mapred.Mapper; import org.apache.hadoop.mapred.MapReduceBase; import org.apache.hadoop.mapred.MapReduceBase; import org.apache.hadoop.mapred.OutputCollector; import org.apache.hadoop.mapred.Reducer; import org.apache.hadoop.mapred.Reducer; import org.apache.hadoop.mapred.Reporter; import org.apache.hadoop.mapred.SequenceFileInputFormat; import org.apache.hadoop.mapred.SequenceFileOutputFormat; import org.apache.hadoop.mapred.TextInputFormat; import org.apache.hadoop.mapred.jobcontrol.job; import org.apache.hadoop.mapred.jobcontrol.jobControl; import org.apache.hadoop.mapred.jobcontrol.jobControl; import org.apache.hadoop.mapred.lib.IdentityMapper; public class MRExample { public static class LoadPages extends MapReduceBase implements Mapper<LongWritable, Text, Text, Text> { public void mapfit mapfit table k, Text val, OutputCollector<Text, Text> oc, Reporter reporter) throws IOException { // Pull the key out String line = val.toString(); int firstComma = line.indexOf(','); String value = line.substring(DirstComma); String value = line.substring(FirstComma + 1); Text outRey = new Text(key); // Prepend an index to the value so we know which file // it came from. Text outVal = new Text('1" + value); oc.collect(outKey, outVal); } public static class LoadAndFilterUsers extends MapReduceBase implements Mapper<LongWritable, Text, Text, Text> { public void map(LongWritable k, Text val, ble void map(Long%ritable k, Text val, OutputCollector<Text, Text> oc, Reporter reporter) throws IOException { // Pull the key out String line = val.toString(); Int firstComma = line.indexOf(','); String value = line.substring(firstComma + 1); int age = Integer.parseInt(value); if (age < 18 || age > 25) return; String key = line.substring(0, firstComma); Text outKey = new Text(key); Text outKey = new Text(key); // Prepend an index to the value so we know which file // it came from. Text outVal = new Text("2" + value): oc.collect(outKey, outVal); public static class Join extends MapReduceBase implements Reducer<Text, Text, Text, Text> { public void reduce(Text key, Iterator<Text> iter, OutputCollector<Text, Text> oc, Reporter reporter) throws IOException { // For each value, figure out which file it's from and store it // accordingly. List<String> first = new ArrayList<String>(); List<String> second = new ArrayList<String>(); while (iter.hasNext()) { Text t = iter.next(); String value = t.toString(); if (value.charAt(0) == '1') first.add(value.substring(1));

import java.io.IOException;

else second.add(value.substring(1));

// Do the cross product and collect the values // bo the cross product and collect the values
for (String s1 : first) {
 for (String s2 : second) {
 String outval = key + "," + s1 + "," + s2;
 oc.collect(null, new Text(outval));
 reporter.setStatus("OK");
 }
} > public static class LoadJoined extends MapReduceBase implements Mapper<Text, Text, Text, LongWritable> { public void map(lic void map(
 Text k,
 Text val,
 OutputCollector<Text, LongWritable> oc,
 Reporter reporter) throws IOException (
 // Find the url
 String line = val.toString();
 int firstComma = line.indexOf(',');
 int secondComma = line.indexOf(',', firstComma);
 String key = line.substring(firstComma, secondComma);
 // drop the rest of the record, I don't need it anymore,
 // just pass a 1 for the combiner/reducer to sum instead.
 Text outKey = new Text(key); Text outKey = new Text(key); oc.collect(outKey, new LongWritable(lL));) public static class ReduceUrls extends MapReduceBase implements Reducer<Text, LongWritable, WritableComparable, Writable> { public void reduce(Text key, Iterator<LongWritable> iter, OutputCollectronWritable> Iter, OutputCollectronWritableComparable, Writable> oc, Reporter reporter) throws IOException { // Add up all the values we see long sum = 0; while (iter.hasNext()) { sum += iter.next().get();
reporter.setStatus("OK"); 3 oc.collect(key, new LongWritable(sum)); , public static class LoadClicks extends MapReduceBase implements Mapper<WritableComparable, Writable, LongWritable, Text> { public void map(
 WritableComparable key,
 Writableval,
 OutputCollector<LongWritable, Text> oc, Reporter reporter) throws IOException { oc.collect((LongWritable)val, (Text)key); public static class LimitClicks extends MapReduceBase implements Reducer<LongWritable, Text, LongWritable, Text> { int count = 0;
public void reduce(LongWritable key, Iterator<Text> iter, OutputCollector<LongWritable, Text> oc, Reporter reporter) throws IOException { // Only output the first 100 records
while {count < 100 && iter.hasNext()) {
 oc.collect(key, iter.next());
 count++;</pre> }

reporter.setStatus("OK");

3

/
public static void main(String[] args) throws IOException {
 JobConf 1p = new JobConf(MRExample.class);
 lp.setJobName("Load Pages");
 lp.setInputFormat(TextInputFormat.class);
}

lp.setOutputKeyClass(Text.class); lp.setOutputValueClass(Text.class); lp.setMapperClass(LoadPages.class); FileInputFormat.addInputPath(lp, new JobConf lfu = new JobConf(MRExample.class); lfu.setJobName("Load and Filter Users"); lfu.setInputFormat(TextInputFormat.class); lfu.setOutputKeyClass(Text.class); lfu.setOutputValueClass(Text.class); lfu.setMapperClass(LoadAndFilterUsers.class); FileInputFormat.addInputPath(lfu, new JobConf join = new JobConf(MRExample.class); join.setJobName("Join Users and Pages"); join.setInputFormat(KeyValueTextInputFormat.class); join.setOutputKeyClass(Text.class); join.setOutputValueClass(Text.class); join.setOutputValueClass(Text.class); join.setNapperClass(IdentityMapper.class); join.setReducerClass(Join.class); FileInputFormat.addInputPath(join, new (user/gates/tmp/fildered_pages")); FileOutputFormat.setOutputPath(join, new (user/gates/tmp/filtered_users")); Path(" Path(Path(' /user/gates/tmp/joined")); join.setNumReduceTasks(50); Job joinJob = new Job(join); joinJob.addDependingJob(loadPages); joinJob.addDependingJob(loadUsers); JobConf group = new JobConf(MRExample.class); group.setJobName("Group URLs"); group.setJobName("Group URLs"); group.setOutputKeyClass(Text.class); group.setOutputValueClass(LongWritable.class); group.setOutputValueClass(LongWritable.class); group.setCutputFormat(SequenceFileOutputFormat.class); group.setCass(LongUnical); group.setCass(LongUnical); group.setCass(LongUnical); group.setCass(LongUnical); group.setReducerClass(ReduceUrls.class) group.setReducerClass(ReduceUrls.class); FileInputFormat.addInputPath(group, new Path("/user/gates/tmp/joined")); FileOutputFormat.setOutputPath(group, new Path("/user/gates/tmp/grouped")); group.setNumReduceTasks(50); Job groupJob = new Job(group); groupJob.addDependingJob(joinJob); JobConf top100 = new JobConf(MRExample.class); JobConf top100 = new JobConf(NRExample.class); top100.setJobName("Top 100 sites"); top100.setInputFormat(SequenceFileInputFormat.class); top100.setOutputEvClass(LongWitable.class); top100.setOutputFormat(SequenceFileOutputFormat.class); top100.setComputFormat(SequenceFileOutputFormat.class); top100.setComputFormat(SequenceFileOutputFormat.class); top100.setComputFormat(SequenceFileOutputFormat.class); top100.setReducerClass(LimitClicks.class); FileInputFormat.addInputPath(top100, new Path("/user/gates/top100sitesforuserIsRo25")); Path("/user/gates/top100sitesforusers18to25")); top100.setNumReduceTasks(1); Job limit = new Job(top100); limit.addDependingJob(groupJob); JobControl jc = new JobControl("Find top 100 sites for users JobConstraint 18 to 25"); jc.addJob(loadDages); jc.addJob(joinJob); jc.addJob(joinJob); ic.addJob(groupJob); jc.addJob(limit); jc.run();

3

3

MASSIVE DATA FUNDAMENTALS





Figure 2: Pig Latin to logical plan translation.



Pig Latin to logical plan translation.

Which are translated to an efficient Map Reduce Plan.



Map Reduce Plan



Pig Latin Program – Basic Program Design

Basic Pig Latin program:

- LOAD data from a file system (HDFS or S3)
- Transform the data.
- STORE to file system or DUMP to output.

A = LOAD *filename* [USING function] [AS schema];

e.g.:

- A = LOAD 'file';
- A = LOAD *filename* USING BinStorage();
- A = LOAD filename USING PigStorage(field_delimiter);
- A = LOAD filename USING PigStorage() AS (field_desc);

Pig Data Loading Functions:

Pig Latin Program — Basic Program Design

Basic Pig Latin program:

- LOAD data from a file system (HDFS or S3)
- Transform the data.
- STORE to file system or DUMP to output.

```
    FILTER

            B = FILTER A BY $1 == 1;
            B = FILTER A BY date == "1980-01-01";
            B = FILTER A BY $1 > 50;
```

- ORDER BY
 - C = ORDER B BY \$0;
 - C = ORDER B BY date;
- LIMIT
 - D = LIMIT B 30;
- JOIN
 D = JOIN C BY \$1, B BY \$1;
 D = JOIN C BY ipaddress, D BY ipaddress;

Pig transformation examples:

Pig Latin Program — Basic Program Design

Basic Pig Latin program:

- LOAD data from a file system (HDFS or S3)
- Transform the data.
- STORE to file system or DUMP to output.

• STORE STORE A INTO 'outputfile'; STORE A INTO 'outputfile.gz';

-- Store UTF-8:
STORE A INTO 'output' USING PigDump();

-- Store in Binary
STORE A INTO 'output' USING
BinStorage();

--- Store with delimiters:
STORE A INTO 'output' USING
PigStorage('*');

Pig Storage examples:

Pig can run locally or on MapReduce

Which version am I running?

\$ pig -help

Pig modes of operation:

Warning: EMR has problems with pig -x local

	Local Mode	MapReduce Mode
Interactive	\$ pig —x local	\$ pig -x mapreduce
Batch	\$ pig -x local <i>filename.pig</i>	\$ pig -x mapreduce filename.pig

Pig Latin statements work with relations.A = LOAD 'foo.txt'A is a relation.

A relation is a "bag."

- A bag is a collection of tuples.
- A tuple is an ordered set of fields
- A field is a piece of data.

Pig Data Types:

- Scalar types: int, long, double, chararray
- map An "associative array" (like a python dictionary) chararray : *anytype*

```
—e.g.
  "first" : "George"
  "last" : "Washington"
  "born" : 1732
```

• tuple

```
(v0, v1, v2, ...)
```

bag — a collection of tuples
 ((a, b, c),
 (d, e, f),
 ...
)

Example (from reference guide)

	First Field	SecondField	Third Field
Data Type	chararray	int	float
Positional notation	\$0	\$1	\$2
Possible name	name	age	gpa
Field value	John	18	4.0

It's best to use names!

Pig Latin FOREACH ... GENERATE

FOREACH ... GENERATE creates new relations from old ones.

Example (from reference guide):

```
A = LOAD 'student' USING
PigStorage()
AS (name:chararray, age:int,
gpa:float);
DUMP A;
(John,18,4.0F)
(Mary,19,3.8F)
(Bill,20,3.9F)
(Joe,18,3.8F)
```

```
X = FOREACH A GENERATE name,$2;
DUMP X;
(John,4.0F)
(Mary,3.8F)
(Bill,3.9F)
(Joe,3.8F)
```

Simple Data Types	Description	Example
Scalars		
int	Signed 32-bit integer	10
long	Signed 64-bit integer	Data: 10L or 101 Display: 10L
float	32-bit floating point	Data: 10.5F or 10.5f or 10.5e2f or 10.5E2F Display: 10.5F or 1050.0F
double	64-bit floating point	Data: 10.5 or 10.5e2 or 10.5E2 Display: 10.5 or 1050.0
Arrays		
chararray	Character array (string) in Unicode UTF-8 format	hello world
bytearray	Byte array (blob)	

Simple data types:

Pig is a complete data flow programming language

Functions:

• +, -, *, /, %,

NULL:

 Operations can return NULL; NULL is ignored by AVG(), MIN(), MAX(), SUM(), COUNT()

Conditions:

• ==, !=, >, <, >=, <=

Conditionals:

- NO IF STATEMENT!
- conditional ? if-true : if-false

Pig is a complete data flow programming language

Functions:

• +, -, *, /, %,

NULL:

 Operations can return NULL; 		Example from Pig	Latin Reference Manual:
NULL is ignored by AVG(), M MAX(), SUM(), COUNT()	A = LOZ f2:int	AD 'data' AS (f1: , :bag{T:tuple(t1	<pre>int, :int,t2:int)});</pre>
Conditions:	DUMP A;	; {(2 3) (1 6)))	
• ==, !=, >, <, >=, <=	(10,1,4)	{(2,3),(4,6)})	
Conditionals:	(10,6,	{(2,3),(4,6),(5,7	<pre>() })</pre>
NO IF STATEMENT!	X = FOR	REACH A GENERATE	f1, f2, f1%f2;
• conditional ? if-true : if-false	DUMP X; (10,1,0) (10,3,1) (10,6,4)	; 0) 1) 4)	
	X = FOF DUMP X; (1,1L) (3,2L) (6,3L)	REACH A GENERATE	f2, (f2==1?1:COUNT(B));

Word Count with Pig

lines = LOAD 's3://gu-anly502/ps02/tobe.txt' as (line:chararray); words = FOREACH lines generate flatten(TOKENIZE(line)) as word; grouped = GROUP words by word; wordcount = FOREACH grouped GENERATE group, COUNT(words); dump wordcount;

LOAD — Loads the data

FOREACH — TOKENIZEs each line. Creates a "words" alias where each tuple is a "word"

GROUP — combines words that have the same word

FOREACH — counts the number of words in each group.

DUMP — sends to standard output.

Note:

- Put spaces around the equals sign (=) !
- Most Pig words are case-sensitive. (Exception: built-in statements like LOAD, FOREACH, GROUP and GENERATE).

grunt> — the Pig command line

grunt> help Commands: <pig latin statement>; - See the PigLatin manual for details: http://hadoop.apache.org/pig File system commands: fs <fs arguments> - Equivalent to Hadoop dfs command: http://hadoop.apache.org/common/docs/current/hdfs shell.html Diagnostic commands: **describe** <alias>[::<alias] - Show the schema for the alias. Inner aliases can be described as A::B. explain [-script <pigscript>] [-out <path>] [-brief] [-dot|-xml] [-param <param_name>=<param_value>] [-param_file <file_name>] [<alias>] - Show the execution plan to compute the alias or for entire script. -script - Explain the entire script. -out - Store the output into directory rather than print to stdout. -brief - Don't expand nested plans (presenting a smaller graph for overview). -dot - Generate the output in .dot format. Default is text format. -xml - Generate the output in .xml format. Default is text format. -param <param name - See parameter substitution for details. -param_file <file_name> - See parameter substitution for details. alias - Alias to explain. dump <alias> - Compute the alias and writes the results to stdout. Always ask for "help" Utility Commands: exec [-param <param_name>=param_value] [-param_file <file_name>] <script> -Execute the script with access to grunt environment including aliases. -param <param name - See parameter substitution for details. Always read the documentation -param file <file name> - See parameter substitution for details. script - Script to be executed. run [-param <param_name>=param_value] [-param_file <file_name>] <script> -Execute the script with access to grunt environment. -param <param name - See parameter substitution for details. -param file <file name> - See parameter substitution for details. script - Script to be executed. sh <shell command> - Invoke a shell command. kill <job id> - Kill the hadoop job specified by the hadoop job id. set <key> <value> - Provide execution parameters to Pig. Keys and values are case sensitive. The following keys are supported: default_parallel - Script-level reduce parallelism. Basic input size heuristics used by default. debug - Set debug on or off. Default is off. job.name - Single-quoted name for jobs. Default is PigLatin:<script name> job.priority - Priority for jobs. Values: very_low, low, normal, high, very_high. Default is normal stream.skippath - String that contains the path. This is used by streaming. any hadoop property. **help** - Display this message. **history** [-n] - Display the list statements in cache. -n Hide line numbers. quit - Quit the grunt shell. grunt>

Grunt supports many Unix commands: ls, cat,

grunt> ls s3://gu-anly502/ 16/02/15 15:48:52 INFO s3n.S3NativeFileSystem: listStatus s3://gu-anly502/ with recursive false s3://gu-anly502/bootstrap.sh<r 1> 936 s3://gu-anly502/gutenberg <dir> s3://gu-anly502/ps02 <dir> s3://gu-anly502/ps03 <dir> s3://gu-anly502/ps04 <dir> grunt>

grunt> ls s3://gu-anly502/ps02/ 16/02/15 15:49:01 INFO s3n.S3NativeFileSystem: listStatus s3://gu-anly502/ ps02 with recursive false s3://gu-anly502/ps02/hamlet.txt<r 1> 1644 s3://gu-anly502/ps02/tobe.txt<r 1> 43 grunt>

grunt> cat s3://gu-anly502/ps02/tobe.txt 16/02/15 15:49:05 INFO s3n.S3NativeFileSystem: Opening 's3://gu-anly502/ ps02/tobe.txt' for reading To be, or not to be- that is the question: grunt>

To minimize Pig output — lower the warning level

Pig uses log4j to log. Make a copy of the existing log4j.properties file and edit it:

```
$ cp /etc/pig/conf.dist/log4j.properties log4j_WARN
```

When you run pig, type: \$ pig -4 log4j_WARN

Hadoop Word Count in Pig

```
$ pig -4 log4j_WARN
grunt> lines = load 's3://gu-anly502/ps02/tobe.txt' as (line:chararray);
. . .
grunt> dump lines;
. . .
(To be, or not to be-)
(that is the question:)
grunt>
grunt> words = FOREACH lines generate flatten(TOKENIZE(line)) as word;
grunt> grouped = GROUP words by word;
grunt> wordcount = FOREACH grouped GENERATE group, COUNT(words);
grunt> dump wordcount;
68560 [JobControl] WARN org.apache.hadoop.mapreduce.JobResourceUploader - No job jar file
set. User classes may not be found. See Job or Job#setJar(String).
68560 [JobControl] WARN org.apache.hadoop.mapreduce.JobResourceUploader - No job jar file
set. User classes may not be found. See Job or Job#setJar(String).
68934 [DataStreamer for file /tmp/hadoop-yarn/staging/hadoop/.staging/job_1455488005182_0020/
job.xml block BP-1229375385-172.31.42.104-1455487984302:blk_1073742532_7091] INFO
amazon.emr.metrics.MetricsSaver - 1 aggregated HDFSWriteDelay 113 raw values into 1
aggregated values, total 1
(To, 1)
(be,1)
(is,1)
(or, 1)
(to, 1)
(be-,1)
(not, 1)
(the, 1)
(that, 1)
(question:,1)
grunt>
```

Sorting the output...

```
grunt> dump wordcount;
68560 [JobControl] WARN org.apache.hadoop.mapreduce.JobResourceUploader
                                                                            - No job jar file
set. User classes may not be found. See Job or Job#setJar(String).
68560 [JobControl] WARN org.apache.hadoop.mapreduce.JobResourceUploader
                                                                            - No job jar file
set. User classes may not be found. See Job or Job#setJar(String).
68934 [DataStreamer for file /tmp/hadoop-yarn/staging/hadoop/.staging/
job_1455488005182_0020/job.xml block
BP-1229375385-172.31.42.104-1455487984302:blk 1073742532 7091] INFO
amazon.emr.metrics.MetricsSaver - 1 aggregated HDFSWriteDelay 113 raw values into 1
aggregated values, total 1
(To, 1)
(be,1)
(is, 1)
(or,1)
(to, 1)
(be-,1)
(not, 1)
(the, 1)
(that, 1)
(question:,1)
grunt> sorted wordcount = ORDER wordcount by $0;
grunt> dump sorted_wordcount;
(To, 1)
(be,1)
(be-,1)
(is, 1)
(not, 1)
(or,1)
(question:,1)
(that, 1)
(the,1)
(to, 1)
```

Working with a larger data set — use LIMIT to limit output.

```
grunt> hamlet = LOAD 's3://gu-anly502/ps02/hamlet.txt' AS (line:chararray);
grunt> words = foreach hamlet generate flatten(TOKENIZE(line)) as word;
grunt> grouped = GROUP words by word;
grunt> wordcount = FOREACH grouped GENERATE group, COUNT(words);
grunt> sorted_words = ORDER wordcount BY $1 DESC;
grunt> sorted words20 = limit sorted words 20;
grunt> dump sorted words20;
(of, 14)
(the,14)
(to,9)
(and,7)
(The,6)
(a,5)
(To,5)
(And, 5)
(that,4)
(we,4)
(bear,3)
(That,3)
(us,3)
(in,3)
(make,2)
(end,2)
(makes,2)
(all,2)
(For, 2)
(have,2)
grunt>
```

Pig Latin scripts can be put in files and run from the command line (like mrjob).



Pig Status — don't just ignore it. Use store lines into 'outputfile'; to write output to a file.

4064342 [main] INFO org.apache.pig.tools.pigstats.mapreduce.SimplePigStats - Script Statistics:

HadoopVersion PigVersion UserId StartedAt FinishedAt Features 2.7.1-amzn-0 0.14.0-amzn-0 hadoop 2016-02-15 17:10:13 2016-02-15 17:10:34 UNKNOWN Success! Job Stats (time in seconds): MaxMapTime MinMapTime AvgMapTime MedianMapTime JobId Reduces Maps MaxReduceTime MinReduceTime AvgReduceTime MedianReducetime Alias Outputs Feature job_1455488005182_0036 0 6 6 6 6 1 0 lines hdfs://ip-172-31-42-104.ec2.internal: MAP ONLY 0 0 0 8020/user/hadoop/outputfile, Input(s): Successfully read 2 records (356 bytes) from: "s3://gu-anly502/ps02/tobe.txt" Output(s): Successfully stored 2 records (44 bytes) in: "hdfs://ip-172-31-42-104.ec2.internal:8020/user/ hadoop/outputfile" Counters: Total records written : 2 Total bytes written : 44 Spillable Memory Manager spill count : 0 Total bags proactively spilled: 0 Total records proactively spilled: 0 Job DAG: job 1455488005182 0036

• • •

16/02/15 17:10:34 INFO mapreduce.SimplePigStats: Script Statistics:

HadoopVersion PigVersionUserId StartedAt FinishedAtFeatures 2.7.1-amzn-00.14.0-amzn-0 hadoop 2016-02-15 17:10:13 2016-02-15 17:10:34 UNKNOWN Success! Job Stats (time in seconds): Reduces JobId Maps MaxMapTimeMinMapTimeAvgMapTimeMedianMapTime MinReduceTime MaxReduceTime AvgReduceTime MedianReducetime Alias Feature Outputs job_1455488005182_0036 1 0 6 6 6 6 0 0 lines MAP ONLY hdfs:// 0 0 ip-172-31-42-104.ec2.internal:8020/user/hadoop/outputfile, Input(s): Successfully read 2 records (356 bytes) from: "s3://gu-anly502/ps02/tobe.txt" Output(s): Successfully stored 2 records (44 bytes) in: "hdfs://ip-172-31-42-104.ec2.internal:8020/ user/hadoop/outputfile" Counters: Total records written : 2 Total bytes written : 44 Spillable Memory Manager spill count : 0 Total bags proactively spilled: 0 Total records proactively spilled: 0 Job DAG: job 1455488005182 0036

16/02/15 17:10:34 INFO mapreduce.SimplePigStats: Script Statistics:

HadoopVersion PigVersionUserId StartedAt FinishedAtFeatures 2.7.1-amzn-00.14.0-amzn-0 hadoop 2016-02-15 17:10:13 2016-02-15 17:10:34 UNKNOWN Success! Job Stats (time in seconds): JobId Maps Reduces MaxMapTimeMinMapTimeAvgMapTimeMedianMapTime MinReduceTime MaxReduceTime AvgReduceTime MedianReducetime Alias Feature Outputs job_1455488005182_0036 1 0 6 6 6 6 0 0 lines MAP ONLY hdfs:// 0 0 ip-172-31-42-104.ec2.internal:8020/user/hadoop/outputfile, Input(s): Successfully read 2 records (356 bytes) from: "s3://gu-anly502/ps02/tobe.txt" Output(s): Successfully stored 2 records (44 bytes) in: "hdfs://ip-172-31-42-104.ec2.internal:8020/ user/hadoop/outputfile" Counters: grunt> cat hdfs:///user/hadoop/output Total records written : 2 cat hdfs:///user/hadoop/outputfile Total bytes written : 44 To be, or not to be-Spillable Memory Manager spill count : 0 that is the question: Total bags proactively spilled: 0 Total records proactively spilled: 0 grunt> Job DAG:

Massive Data Fundamentals

job 1455488005182 0036

Grunt built-in commands:

```
"\\i" ...
Was expecting one of:
                                                    "run" ...
    <EOF>
    "cat" ...
                                                    "exec" ...
                                                    "scriptDone" ...
    "clear" ...
    "fs" ...
                                                    11 11
                                                       . . .
    "sh" ...
                                                    11 11
                                                       . . .
    "cd" ...
                                                    <EOL> ...
    "ср" ...
                                                    ","
    "copyFromLocal" ...
    "copyToLocal" ...
                                                grunt> describe lines
                                                describe lines
    "dump" ...
    "\\d" ...
                                                16/02/15 17:14:10 INFO
                     Describe and Illustrate
    "describe" ...
                                                Configuration.deprecation:
                     show the structure of
    "\\de" ...
                                                fs.default.name is deprecated.
    "aliases" ...
                                                Instead, use fs.defaultFS
                   relations.
                                                lines: {line: chararray}
    "explain" ...
    "\\e" ...
    "help" ...
                                                grunt> illustrate lines;
    "history" ...
    "kill" ...
    "ls" ...
                                                  lines | line:chararray
    "m∨" ...
    "mkdir" ...
    "pwd" ...
    "quit" ...
                                                            that is the question:
    "\\q" ...
    "register" ...
                                                          -------
    "rm" ...
    "rmf" ...
                                                grunt>
    "set" ...
    "illustrate" ...
```

Pig User Defined Functions (UDFs)

UDFs expand Pig's functionality.

- Parse input lines
- Perform complex operations.
- Example a UDF could search the MaxMind IP address geolocation database provided that the database is on each node.

Coding Options:

- Write in Java import as registered jar files.
- Write in jython (Python that generates jar files) import as registered jar files.
- Write in python Access with "pig streaming API" (similar to Hadoop streaming)

Pig can process any tab-delimited data. How do you process data that aren't tab-delimited? (e.g. Apache log files)

Piggybank — a collection of algorithms for pig.

- CommonLogLoader
 - https://pig.apache.org/docs/r0.14.0/api/org/apache/pig/piggybank/storage/apachelog/CommonLogLoader.html
- CombinedLogLoader:
 - https://pig.apache.org/docs/r0.14.0/api/org/apache/pig/piggybank/storage/apachelog/CombinedLogLoader.html

```
raw = LOAD 'combined_log' USING
org.apache.pig.piggybank.storage.apachelog.CombinedLogLoader AS (remoteAddr,
remoteLogname, user, time, method, uri, proto, status, bytes, referer, userAgent);
```

Note: I was not able to get CombinedLogLoader to work with the ForensicsWiki logs!

I used REGEX_EXTRACT to extract the log file entries:

Pig program to produce hits-by-day

DEFINE EXTRACT org.apache.pig.piggybank.evaluation.string.EXTRACT();

raw_logs = load 's3://gu-anly502/ps03/forensicswiki.2012.txt' as (line:chararray);

```
logs_base =
FOREACH raw_logs GENERATE FLATTEN (
EXTRACT( line,
    '^(\\S+) (\\S+) \\[([\\w/]+):(\\d{2}:\\d{2}:\\d{2}) [+\\-]\\d{4}\\]
"(\\S+) (\\S+) \\S+" (\\S+) (\\S+) "([^"]*)" "([^"]*)"'
    ) AS (
    host: chararray, identity: chararray, user: chararray, date: chararray, time:
chararray, verb: chararray, url: chararray, request: chararray, status: int,
    size: chararray, referrer: chararray, agent: chararray
    );
by_date = GROUP logs_base BY (date);
date_counts = FOREACH by_date GENERATE
    group as date, -- the key you grouped on
    COUNT(logs_base); -- the number of log lines wiht this date
```

dump date_counts;

Pig output

```
$ pig parse apache.pig
16/02/21 20:18:47 INFO pig.ExecTypeProvider: Trying ExecType : LOCAL
16/02/21 20:18:47 INFO pig.ExecTypeProvider: Trying ExecType : MAPREDUCE
16/02/21 20:18:47 INFO pig.ExecTypeProvider: Picked MAPREDUCE as the ExecType
     [main] INFO org.apache.pig.Main - Apache Pig version 0.14.0-amzn-0 (r:
45
unknown) compiled Jan 14 2016, 02:55:53
16/02/21 20:18:47 INFO pig.Main: Apache Pig version 0.14.0-amzn-0 (r: unknown)
compiled Jan 14 2016, 02:55:53
16/02/21 20:23:09 INFO util.MapRedUtil: Total input paths to process : 5
(01/Jul/2012,35039)
(01/Sep/2012,33272)
(02/Jul/2012,46445)
(02/Sep/2012,36225)
(03/Jul/2012,43922)
(03/Sep/2012,40703)
(04/Jul/2012,38576)
(30/Jul/2012,45488)
(30/Sep/2012,37817)
                                         266 seconds to process 4GB file!
(31/Jul/2012,48353)
263298 [main] INFO org.apache.pig.Main - Pig script completed in 4 minutes, 23
seconds and 386 milliseconds (263386 ms)
16/02/21 20:23:10 INFO pig.Main: Pig script completed in 4 minutes, 23 seconds and
386 milliseconds (263386 ms)
[20:23:11 last: 266s][~/ANLY502/L05]
$
```

Parse the date as a "datetime" and create a new relation with just the desired fields.

```
Old regular expression:
                       logs_base =
                              FOREACH raw_logs GENERATE FLATTEN (
                                        EXTRACT( line,
                                         '^(\\S+) (\\S+) \\[([\\w/]+):(\\d{2}:\\d{2}) [+\\-]\\d{4}\\]
                       "(\\S+) (\\S+) \\S+" (\\S+) "([^{"}]^{*})" "([^{"}]^{*})" "
                                        ) ) AS (
                                        host: chararray, identity: chararray, user: chararray, date: chararray, time:
                       chararray, verb: chararray, url: chararray, request: chararray, status: int,
                                        size: chararray, referrer: chararray, agent: chararray);
New:
                       logs base =
                              FOREACH
                                 raw logs
                              GENERATE
                                 FLATTEN ( EXTRACT( line,
                                         ^(()S+) (()S+) (((S+) (()S+) (()S+)
                       "([^"]*)" "([^"]*)"'
                                        )) AS (
                                        host: chararray, identity: chararray, user: chararray, datetime_str:
                       chararray, verb: chararray, url: chararray, request: chararray, status: int,
                                        size: int, referrer: chararray, agent: chararray
                                        );
                                                                                                                                "schema"
```

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:

#
New Logical Plan:
logs: (Name: LOStore Schema: date#58:datetime,host#46:chararray,url#51:chararray,size#54:int)
logs: (Name: LOForEach Schema: date#58:datetime,host#46:chararray,url#51:chararray,size#54:int)
 (Name: LOGenerate[false,false,false,false] Schema: date#58:datetime,host#46:chararray,url#51:chararray,size#5 Uids=[51, 54, 58, 46]
(Name: UserFunc(org.apache.pig.builtin.ToDate2ARGS) Type: datetime Uid: 58)
datetime_str:(Name: Project Type: chararray Uid: 49 Input: 0 Column: (*))
(Name: Constant Type: chararray Uid: 57)
host:(Name: Project Type: chararray Uid: 46 Input: 1 Column: (*))
url:(Name: Project Type: chararray Uid: 51 Input: 2 Column: (*))
(Name: LOInnerLoad[3] Schema: datetime_str#49:chararray)
(Name: LOInnerLoad[0] Schema: host#46:chararray)
(Name: LOInnerLoad[5] Schema: url#51:chararray)
(Name: LOInnerLoad[8] Schema: size#54:int)
 logs_base: (Name: LOForEach Schema: host#74:chararray,identity#75:chararray,user#76:chararray,datetime_str#77 rray,status#81:int,size#82:int,referrer#83:chararray,agent#84:chararray)
 (Name: LOGenerate[true] Schema: host#74:chararray,identity#75:chararray,user#76:chararray,datetime_str#77 rray.status#81:int.size#82:int.referrer#83:chararray.agent#84:chararray)
 (Name: UserFunc(org.apache.pig.piggybank.evaluation.string.FXTRACT) Type: tuple Uid: 73)
(Name: Cast Type: chararray Uid: 17)
(Name: Constant Type: chararray Hid: 72)
 raw logs: (Name: 101.oad Schema: line#17:bytearray)RequiredEields:null
"

Explain "Logical Plan"

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:



Explain Physical Plan

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:

```
Map Reduce Plan
MapReduce node scope-19
Map Plan
logs: Store(fakefile:org.apache.pig.builtin.PigStorage) - scope-18
 ---logs: New For Each(false,false,false,false)[bag] - scope-17
       POUserFunc(org.apache.pig.builtin.ToDate2ARGS)[datetime] - scope-9
        |---Project[chararray][3] - scope-7
        |---Constant(dd/MMM/yyyy:HH:mm:ss Z) - scope-8
       Project[chararray][0] - scope-11
       Project[chararray][5] - scope-13
       Project[int][8] - scope-15
     ---logs_base: New For Each(true)[bag] - scope-6
            POUserFunc(org.apache.pig.piggybank.evaluation.string.EXTRACT)[tuple] - scope-4
            [---Cast[chararray] - scope-2
                |---Project[bytearray][0] - scope-1
             ---Constant(^(\S+) (\S+) \[([^\]]+)\] "(\S+) (\S+) \S+" (\S+) (\S+) "([^"]*)" "([^"]*)") - scope-3
        |---raw_logs: Load(s3://gu-anly502/ps03/forensicswiki.2012-01.unzipped/access.log.2012-01-01:org.apache.pig.builtin.PigStorage) - scope-0--
```

Explain Map Reduce Plan

logs = FOREACH logs_base GENERATE ToDate(datetime_str,'dd/MMM/yyyy:HH:mm:ss Z') AS
date, host, url, size;

Describe logs:

logs: {date: datetime,host: chararray,url: chararray,size: int}

Explain logs:

Final demo: list of forensicswiki hits by date:

Program:

```
raw logs = load 's3://gu-anly502/ps03/forensicswiki.2012.txt' as (line:chararray);
       logs base =
         FOREACH
          raw logs
         GENERATE
          FLATTEN ( EXTRACT( line,
            "([^"]*)"'
            )) AS (
            host: chararray, identity: chararray, user: chararray, datetime str: chararray, verb:
       chararray, url: chararray, request: chararray, status: int,
            size: int, referrer: chararray, agent: chararray
            );
       by date = GROUP logs BY (date);
       date counts = FOREACH by date GENERATE
           group as date, -- the key you grouped on
COUNT(logs_base); -- the number of log lines wiht this date
       dump date counts;
Output:
       (,0)
       (2012-01-01T00:00:00.000Z,29116)
       (2012-01-02T00:00:00.000Z,38188)
       (2012-12-31T00:00:00.000Z,36631)
       (2013-01-01T00:00:00.000Z,1283)
       329255 [main] INFO org.apach pig.Main - Pig script completed in 5 minutes, 29 seconds and 337
                                                331 seconds! (4x faster that mrjob)
       milliseconds (329337 ms)
       16/02/22 00:43:57 INFO pig.Main: Pig script completed in 5 minutes, 29 seconds and 337
       milliseconds (329337 ms)
       [00:43:58 last: 331s][~/ANLY502/L05]
```

Add a second GENERATE:

logs = FOREACH logs_base GENERATE ToDate(SUBSTRING(datetime_str,0,11),'dd/MMM/
yyyy') AS date, host, url, size;
logs2 = FOREACH logs GENERATE SUBSTRING(ToString(date),0,10) AS date, host,
url, size;

```
by_date = GROUP logs2 BY (date);
date_counts = FOREACH by_date GENERATE
  group AS date, -- the key you grouped on
  COUNT(logs2); -- the number of log lines wiht this date
```

```
date_counts_sorted = ORDER date_counts BY date;
dump date_counts_sorted;
```

And run...

```
(2012-12-28,39090)
(2012-12-29,54360)
(2012-12-30,40828)
(2012-12-31,36631)
(2013-01-01,1283)
368896 [main] INFO org.apache.pig.Main - Pig script completed in 6 minutes, 8
seconds and 977 milliseconds (368977 ms)
16/02/22 01:21:35 INFO pig.Main: Pig script completed in 6 minutes, 8 seconds and
977 milliseconds (368977 ms)
[hadoop@ip-172-31-37-188 L05]$ %
```

368 seconds (up from 331)

MaxMind Join with the Forensicswiki Data

DEFINE EXTRACT org.apache.pig.piggybank.evaluation.string.EXTRACT();

raw_logs = load 's3://gu-anly502/ps03/forensicswiki.2012.txt' as (line:chararray);

maxmind = load 's3://gu-anly502/ps03/maxmind' as (ipaddr:chararray, country:chararray);

```
logs_base =
 FOREACH
  raw logs
 GENERATE
  FLATTEN ( EXTRACT( line,
    "([^"]*)"'
    )) AS (
    host: chararray, identity: chararray, user: chararray, datetime_str: chararray, verb:
chararray, url: chararray, request: chararray, status: int,
    size: int, referrer: chararray, agent: chararray
    );
geolocated_logs = JOIN logs_base BY host, maxmind BY ipaddr;
geolocated 50 = LIMIT geolocated logs 50;
dump geolocated_50;
(180.76.5.67,-,-,01/Jan/2012:13:02:39 -0800,GET,/wiki/Special:WhatLinksHere/User talk:Marc Yu,
200,3799,-,Mozilla/5.0 (compatible; Baiduspider/2.0; +http://www.baidu.com/search/
spider.html),180.76.5.67,China)
(180.76.5.89,-,-,01/Jan/2012:02:27:53 -0800,GET,/wiki/Special:RecentChangesLinked/Libvshadow,
200,4391,-,Mozilla/5.0 (compatible; Baiduspider/2.0; +http://www.baidu.com/search/
spider.html),180.76.5.89,China)
(180.76.5.89,-,-,01/Jan/2012:21:47:55 -0800,GET,/images/7/79/?C=S;0=D,200,553,-,Mozilla/5.0
(compatible; Baiduspider/2.0; +http://www.baidu.com/search/spider.html),180.76.5.89,China)
```