BigData and Machine Learning with Hadoop and Spark Frameworks

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Outline I

- Introduction
- 2 Hadoop
 - Introduction
 - Architecture
 - HDFS
 - Yarn
 - MapReduce
- Spark
 - Introduction
 - Architecture and Ecosystem
 - Spark Modules : Core, SQL, Mlib, Streaming and GraphX
 - Spark Core
 - Spark SQL
 - Spark Mlib
 - Spark Streaming
 - Spark GraphX



Outline II

- Dask
 - Introduction
 - Architecture
 - Dask API
 - Dask Ecosystem
- 5 TP
 - TP0 : Hadoop Installation
 - TP1 : Hadoop World count
 - TP2: Hadoop DataBase request
 - TP3 : Start with Spark
 - TP4 : Spark ML, Data processing
 - TP5 : Spark ML, Machine learning
 - TP6 : Spark ML, Image processing
 - TP7: Start with Dask



Objectifs

Objectifs

- General Overview on Hadoop and Spark
- Introduce to Hadoop
- Introduction to Spark Framework

Audience and Prerequisites

- Audience : computer science and data scientist students
- Prerequisites :
 - sequential programming in java and python
 - elementary of machine learning, data analytics
 - image processing
- Material(Slide+TPs) available at: git clone https://github.com/jgratien/BigDataHadoopSparkDaskCourse.git

Motivation

Introduction to Bigdata

BigData

- What is Bigdata?
- What are the BigData issues?

Outline





Hadoop

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- Architecture
- HDFS
- Yarn
- MapReduce



- Architecture and Ecosystem
- Spark Modules: Core, SQL, Mlib, Streaming and GraphX
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 - Spark GraphX





Introduction to Hadoop

- Hadoop definition
 - Java opensource software framework
 - Data storage management
 - Parallel data analysis
 - part of Apache project supported by the Apache Software Foundation

Introduction to Hadoop

- Hadoop History
 - 1990 2000 : World Wide Web
 - Yahoo, AltaVisa,...: first search engines
 - Nutch open source project created by Doug Cutting and Mike Cafarella
 - 2006: Nutch project is split: the distributed storage and computing framework -> Hadoop
 - 2008 : Hadoop 1.0 (Open Source Project proposed by Yahoo)
 - 2012 : Hadoop 2.0 release
 - 2017 : Hadoop 3.0

Introduction Architecture HDFS Yarn MapReduce

Hadoop Framework

Introduction to Hadoop

Why Hadoop?

- BigData issues :
 - · increasing amount of data amount
 - distributed storage facilities
 - parallel data processing management
 - fault tolerance management

Outline





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Hadoop Framework Architecture

the basic Hadoop Framework based of 4 main modules :

- HDFS : Hadoop Distributed File System
- YARN : Yest Another Ressource Negotiator

- Map : parallel data processing
- Reduce : collecting data and producing results





Hadoop Ecosystem

Hadoop ecosystem:

- Ambari: Hadoop component and services web interface management
- Cassandra : Distributed Data Base system
- Flume : Data Stream management layer
- HBase : NoSql distributed Data Base
- HCatalog : data storage management
- Hive : data storage with a SQL API
- Oozie : task framework
- Pig: HDFS data processing framework
- Solr : data indexing framework
- Sqoop : SQL DB and Hadoop data transfer framework
- Zookeeper: distributed data processing management



Introduction Architecture HDFS Yarn MapReduce

Hadoop Framework

Hadoop distributions

Hadoop Distributions:



- Hortonworks
- Cloudera
- MAPR







Outline



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Spark

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 - Spark GraphX
- 4

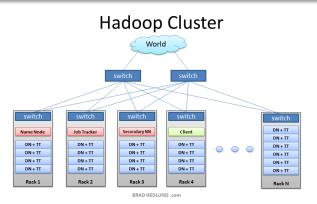
ask



Introduction Architecture HDFS Yarn MapReduce

Hadoop Framework

Hadoop Cluster



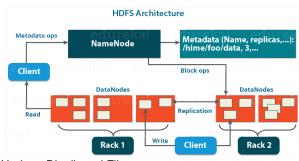
Purpose: Scalability, Fault tolerance and Reliality management



Hadoop Cluster

Concepts

- NameNode
- DataNode
- Replication
- Block, Blocksize



HDFS: Hadoop Distributed Filesystem

HDFS commands

HDFS commands:

```
Starting HDFS
```

```
1  # Format nodes
2 > hadoop namenode -format
3
4  # Starting HDFS services
5 > start-dfs.sh
```

Shutting down HDFS

```
1 # Stopping HDFS services
2 > stop-dfs.sh
```

HDFS commands

HDFS commands:

Inserting Data into HDFS

```
1 # Step 1 : Create input directory
2 > $HADOOP_HOME/bin/hadoop fs -mkdir /usr/input
3
4 # Step 2 : copy data from local filesystem to hdfs
    filesystem
5 > $HADOOP_HOME/bin/hadoop fs -put /home/file.txt /
    user/input
6
7 # Step 3 : check results with ls cmd
8 > $HADOOP_HOME/bin/hadoop fs -ls /usr/input
```

HDFS commands

HDFS commands:

Retreiving Data from HDFS

```
1 # Step 1 : view data
2 > $HADOOP_HOME/bin/hadoop fs -cat /user/outputfile
3
4 # Step 2 : get data from hdfs filesystem to local
    filesystem
5 > $HADOOP_HOME/bin/hadoop fs -get /user/output/ /
    home/hadoop_out
6
7 # Step 3 : check results with ls cmd
8 > $HADOOP_HOME/bin/hadoop fs -mkdir /usr/input
```

HDFS commands list

Commande name	Description
fs -help <cmd-name></cmd-name>	return cmd usage
fs -ls <path></path>	list <path> directory contents</path>
fs -lsr <path></path>	Is ,recursively with sub dirs
fs -du <path></path>	show disk usage in bytes
fs -dus <path></path>	show disk usage in bytes and summary
fs -test [ezd] <path></path>	return 1 if path exists;
	has 0 length; or is a directory,
	otherwize 0
fs -cat <filename></filename>	
fs -tail [-f] <filename></filename>	

HDFS commands list

Commande name	Description
fs -mv <src><dest></dest></src>	move file or directory within HDFS
fs -cp <src> <dest></dest></src>	copy file or directory within HDFS
fs -rm <path></path>	remove file or directory within HDFS
fs -rmr <path></path>	rm recursively
fs -put <localsrc> <dest></dest></localsrc>	copy files or dirs from local FS to HDFS

HDFS commands list

Commande name	Description
fs -copyFromLocal <localsrc> <dest></dest></localsrc>	identical to put
fs -moveFromLocal <localsrc> <dest></dest></localsrc>	move file or dirs
	from local FS to HDFS
fs -get [-crc] <src> <localdest></localdest></src>	copy file or dirs
	from HDFS to local FS
fs -getmerge [-crc] <src> <localdest></localdest></src>	copy all files from HDFS
	and merge
	to a single file in FS
fs -copyToLocal <localsrc> <dest></dest></localsrc>	copy file or dirs
	from HDFS to local FS
fs -moveToLocal <localsrc> <dest></dest></localsrc>	move file or dirs
	from HDFS to local FS
fs -mkdir <path></path>	create directory in HDFS



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Spark

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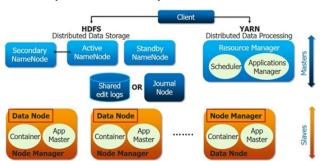


ask



YARN

Apache Hadoop 2.0 and YARN



YARN

Starting YARN

- # Starting YARN services
- 2 > start-yarn.sh

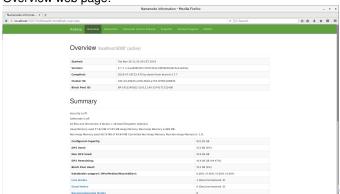
Shutting down YARN

- 1 # Stopping YARN services
- 2 > stop-yarn.sh

Hadoop Web tools

Hadoop Web Tools

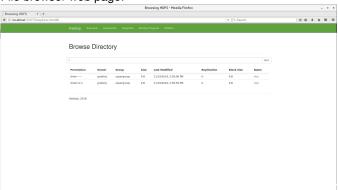
Web tools on: http://<hostname>:<port> <hostname>:<port> (default localhost:50070) are defined in hdfs-site.xml Overview web page:



Hadoop Web tools

Hadoop Web Tools

File browser web page:



Introduction Architecture HDFS Yarn MapReduce

Outline



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Spark

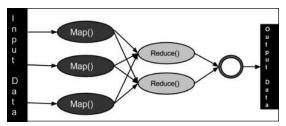
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)ask



MapReduce Framework



MapReduce Framework

MapReduce Framework

MapReduce Algorithm:

- Programming model;
- Two stages:
 - Map stage :
 - Mapper jobs;
 - data are processed in parallel by mapper jobs;
 - Reduce Stage :
 - Reducer jobs;
 - mapper output data are processed Reducer jobs;
 - Reducer jobs produce new set of output stored in HDFS.

MapReduce Framework

Input and Output:

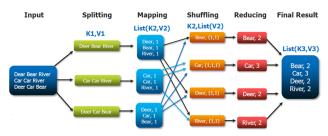
- Input : <key,value> pairs
 - key and values classes must implement Writable Interface;
 - key class have to implemente WritableComparable Interface;
- $\bullet \ \, \text{Job}: (\text{Input}) \rightarrow \text{map} \rightarrow \text{<k2,v2>} \rightarrow \text{reduce} \rightarrow \text{<k3,v3>}(\text{Output})$

	Input	Output
Мар	<k1,v1></k1,v1>	list(<k2,v2>)</k2,v2>
Reduce	<k2,v2></k2,v2>	list(<k3,v3>)</k3,v3>



MapReduce Framework

The Overall MapReduce Word Count Process



Java exemple

4 5

6 7

8 9

10

11

12

13 14 15

16

17 18

WordCount Java class

```
import org.apache.hadoop.*;
public class WordCount
  public static class TokenizerMapper extends Mapper<Object, Text, Text, IntWritable> {
    public void map(Object key, Text value, Context context)
                    throws IOException, InterruptedException {
        . . . ;
  public static class IntSumReducer extends Reducer<Text.IntWritable.Text.IntWritable> {
    public void reduce(Text key, Iterable<IntWritable> values,Context context)
                      throws IOException, InterruptedException {
         ...;
  public static void main(String[] args) throws Exception {
  . . . ;
```

Java exemple

2

4

10

11

Word Count Mapper Java class

Java exemple

2

6 7

8

10

11

12 13

14

15 16 17

WordCount Reducer Java class

```
public class WordCount
   public static class IntSumReducer extends Reducer<Text, IntWritable, Text, IntWritable>
        private IntWritable result = new IntWritable():
        void reduce(Text key, Iterable<IntWritable> values,Context context)
                   throws IOException, InterruptedException
           int sum = 0:
           for (IntWritable val : values) {
              sum += val.get();
           result.set(sum);
           context.write(key, result);
```

Java exemple

2

4 5

6 7

8

9

10

11

12

13

14

15

16 17

Main test function

```
public class WordCount
  public static void main(String[] args) throws Exception
    Configuration conf = new Configuration():
    Job job = Job.getInstance(conf, "word count");
    job.setJarByClass(WordCount.class);
    job.setMapperClass(TokenizerMapper.class);
    job.setCombinerClass(IntSumReducer.class);
    job.setReducerClass(IntSumReducer.class);
    job.setOutputKeyClass(Text.class);
    job.setOutputValueClass(IntWritable.class);
    FileInputFormat.addInputPath(job, new Path(args[0]));
    FileOutputFormat.setOutputPath(job, new Path(args[1]));
    System.exit(job.waitForCompletion(true) ? 0 : 1);
```

Java exemple

5 6 7

8

10

Suppose we have two test files file01 and file02 in current directory

Prepare Test Data

```
// Two test file file01 file02 in current directory
$ ls
    file01
    file02
$ hdfs dfs -put file01 /user/gratienj/input
$ hdfs dfs -cat /user/gratienj/input/file01
    Hello World Bye World
$ hdfs dfs -put file02 /user/gratienj/input
$ hdfs dfs -cat /user/gratienj/input/file02
    Hello Hadoop Goodbye Hadoop
```

Java exemple

Suppose the Java Project is compiled and generates the jar file BigDataTP1.jar

Run application

```
$ hadoop jar BigDataTP1.jar hadoop.WordCount /user/gratienj/input /user/gratienj/output
```

Check results

```
1 $ hdfs dfs -cat /user/gratienj/output/part-r-00000
2 Bye 1
3 Goodbye 1
4 Hadoop 2
5 Hello 2
6 World 2
```

Python example

Python example with Hadoop Streaming

Mapper python script

8

```
import sys
for line in sys.stdin:
    line = line.strip()
    words = line.split()
    for word in words:
        print ('%s\t%s' % (word, 1))
```

Python example

Reducer python script Part 1

```
from operator import itemgetter
import sys
current_word = None
current_count = 0
word = None
```

Python example

3

10

11

12

13

14

15

16

Reducer python script Part 2

```
for line in sys.stdin: # input comes from STDIN
    line = line.strip()
    word, count = line.split('\t', 1)
    try:
        count = int (count)
    except ValueError:
        continue
    if current_word == word:
        current count += count
    else.
        if current word:
            print ('%s\t%s' % (current word, current count)) # write result to STDOUT
        current count = count
        current word = word
if current word == word: # do not forget to output the last word if needed!
    print('%s\t%s' % (current_word, current_count))
```

Python example

Python example with Hadoop Streaming: Part 1

Copy test files on HDFS

\$ hdfs dfs -copyFromLocal /home/gratienj/test/books /user/gratienj/input/books

Run application

```
$ hadoop jar $HADOOP_HOME/share/hadoop/tools/lib/hadoop-streaming-2.7.7.jar \
-file /home/hduser/mapper.py -mapper /home/hduser/mapper.py \
-file /home/hduser/reducer.py -reducer /home/hduser/reducer.py \
-input /user/gratienj/input/books/* -output /user/gratienj/books-output
```

Introduction Architecture HDFS Yarn MapReduce

Hadoop Framework

Python example

Python example with Hadoop Streaming: Part 2

Check results

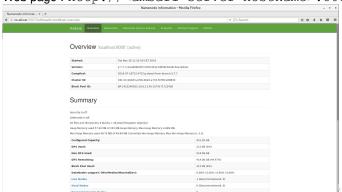
- \$ hdfs dfs -ls /user/gratienj/books-ouput
- 2 Found 1 items
 - /user/gratienj/books-output/part-00000
 - \$ hdfs dfs -cat /user/gratienj/books-output/part-00000

Hadoop Web tools

Ambari Server Tools

Ambari Server : tools to manage and monitor applications for Apache Hadoop

Web page: http://<ambari-server-hostname>:8080



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Introduction to Spark

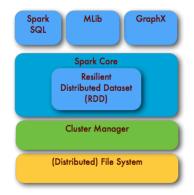
- Spark : Big Data framework for data processing
- History
 - 2009 : AMPLab, UC Berkeley University
 - 2010 : Open source as an Apache project
- Complete and Unified framework
 - Hadoop (MapReduce)
 - Storm (Streaming)
 - Languages : Java, Scala, Python
 - SQL

Outline

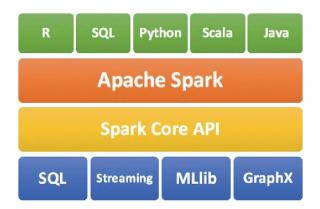
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Apache Spark Architecture



Apache Spark Ecosystem



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- Dasl



Spark Core: Spark configuration

Spark Cluster Configurations:

- Local mode
- Cluster mode
- Client mode

Spark parallel concepts:

- multiple executors (private JVM)
- multiple cores per executor

Spark Core: Spark configuration

Configuring a SparkContext

```
import pyspark
from pyspark import SparkConf
sc_conf = SparkConf()
sc_conf.setAppName(app_name)
sc_conf.setMaster('local[*]')
sc_conf.set('spark.executor.memory', '4g')
sc_conf.set('spark.executor.cores', nb_cores)
sc_conf.set('spark.driver.memory', '16G')
sc_conf.set('spark.cores.max', '32')
sc_conf.set('spark.driver.maxResultSize', '10G')
sc_conf.set('spark.logConf', True)
```

ntroduction Architecture and Ecosystem

Spark Modules: Core, SQL, Mlib, Streaming and GraphX

Spark Framework

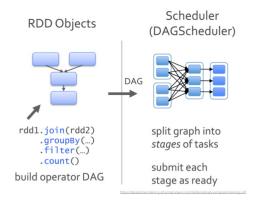
Spark Core: Spark context

Create a SparkContext

```
1 import pyspark
```

- 2 from pyspark import SparkContext
- 3 sc =SparkContext()

Spark Core: Data concepts



Spark Core: Data concepts

Spark Data concepts :

- RDD : Resilient Distributed Data, list of <key,value>
- Transformations: apply lambda to creating new RDDs
- DAG : pipeline of transformation
- Actions : operations on the RDD producing results
- Scheduler : perform actions on DAG
- Stage: parallel operations
- Pipeline : sequence of stages

Spark Core: RDD

```
Create a Spark RDD
```

```
import pyspark
from pyspark import SparkContext
sc = SparkContext()

nums= sc.parallelize([1,2,3,4])
nums.take(1)
```

Output

```
1 [1]
```

Spark Core: RDD Transformationq and Actions

```
Spark RDD transformations and Actions
```

```
sc =SparkContext()
nums= sc.parallelize([1,2,3,4])
squared = nums.map(lambda x: x*x).collect()
for num in squared:
    print('%i_' % (num))
```

Output

```
1 1
2 4
3 9
4 16
```

Spark Core: RDD Transformations and Actions

Transformation:

- apply lambda function to RDD
- create a new RDD
- lazy evaluation
- create a DAG

Spark Core: RDD Transformations and Actions

	Commande name	Description
	map()	apply to each RDD line
	flatMap()	apply to all RDD elements
	mapPartition	apply per partition
	filter()	apply to a selection of lines
Examples:	groupBy()	create new set of (key,value)
	groupByKey()	
	reduceByKey()	
	sample()	selection of lines
	union()	fusion of two RDDs
	join()	union without duplicate keys

Spark Core: RDD Transformations and Actions

Actions:

- get results on a pipeline of transformations
- perform all the transformation
- real evaluation

Examples:

Examples:			
Commande name	Description		
getNumPartition()			
reduce()	apply lambda		
	to all elements		
collect()	create a collection		
count()	count elements		
max(), min()	stats		
sum()			

Spark SQL: DataFrame

Unified Data Abstraction



Image credit: http://barrymieny.deviantart.com/





Spark Core: RDD

Create a Spark SQL context

```
1 from pyspark.sql import Row
2 from pyspark.sql import SQLContext
3 sqlContext = SQLContext(sc)
```

Create a DataFrame

```
1 list_p=[('John',19),('Smith',29),('Adam',35)]
2 rdd = sc.parallelize(list_p)
3 ppl_rdd=rdd.map(lambda x: Row(name=x[0], age=int(x [1])))
4 ppl df rdd = sglContext.createDataFrame(ppl rdd)
```

Spark SQL: DataFrame

Print DataFrame Schema

```
1 DF_ppl.printSchema()
2 root
3 |-- age: long (nullable = true)
4 |-- name: string (nullable = true)
```

Spark SQL: DataFrame

Print DataFrame Schema

```
df = sglContext.read.csy(SparkFiles.get("adult data.csy"), header=True, inferSchema= True)
     df string.printSchema()
      root
 4
       |-- age: string (nullable = true)
 5
       |-- workclass: string (nullable = true)
6
       |-- fnlwgt: string (nullable = true)
7
       |-- education: string (nullable = true)
8
       |-- education num: string (nullable = true)
9
       |-- marital: string (nullable = true)
10
       |-- occupation: string (nullable = true)
11
       |-- relationship: string (nullable = true)
12
       |-- race: string (nullable = true)
13
       |-- sex: string (nullable = true)
14
       |-- capital gain: string (nullable = true)
15
       |-- capital_loss: string (nullable = true)
16
       |-- hours week: string (nullable = true)
17
       |-- native_country: string (nullable = true)
18
       |-- label: string (nullable = true)
```

Spark SQL: DataFrame

```
Select columns

df.select('age','fnlwgt')

.show(5)

10
```

```
Select columns

+--+----+
|age|fnlwgt|
+--+----+
| 39| 77516|
| 50| 83311|
| 38|215646|
| 53|234721|
| 28|338409|
+--+----+
only showing top 5 rows
```

2

Spark Framework

Spark SQL: DataFrame

Select columns df.groupBy("education"). count().sort("count", 10 ascending=True).show() 11 15 16

```
Select columns
```

```
education|count|
   Preschool
                 511
     1st-4th| 168|
     5th-6th| 333|
    Doctorate | 413|
         12th| 433|
          9th|
                5141
 Prof-school |
                5761
     7th-8th|
                6461
         10th| 933|
  Assoc-acdm| 1067|
         11th| 1175|
   Assoc-voc| 1382|
     Masters| 1723|
    Bachelors| 5355|
|Some-college| 7291|
     HS-grad|10501|
```

18

19

Spark SQL: DataFrame

Describe data: describe() functions give a summary of statistics :

- count,
- mean,min,max
- standarddeviation

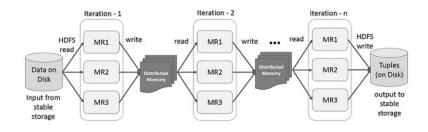
```
Describe

df.describe('capital_gain
    ').show()
```

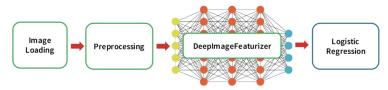
```
Describe

+----+
|summary| capital_gain|
+-----+
| count| 32561|
| mean|1077.6488437087312|
| stddev| 7385.292084840354|
| min| 0|
| max| 99999|
+-----+
```

Spark Mlib



Spark Mlib



MLlib Pipeline

Spark Mlib

Mlib provides tools for Machine learning

- set of classifier and regression algorithms
- create models from Spark Dataframe
- set of tools to evaluate the predicting models
- concept of pipeline to process Data

Spark Mlib Pipeline

Spark Pipeline : a sequence of stages (Transformer, Estimator)

- String Indexer : convert Categorical Data to numerics
- Standard Scaler on Continuous Values
- VectorAssembler: features must be a dense vector

Spark Mlib: Data processing

StringIndexer

Spark Mlib: Data processing

OneHotEncoder

```
1 from pyspark.ml.feature import OneHotEncoder
2 encoder = OneHotEncoder(dropLast=False, inputCo
```

- encoder = OneHotEncoder(dropLast=False, inputCol="
 encoded_key", outputCol="vec_key")
- 3 vec_df_rdd = encoder.transform(encoded_df_rdd)

VectorAssembler

```
1 from pyspark.ml.feature import VectorAssembler
```

- 3 ass_df_rdd = assembler.transform(df_rdd)

Spark Mlib: Data processing

Pipeline

Spark ML: ML pipeline Part 1

```
create DataFrame
```

features"])

```
rdd.map(lambda x: (x["newlabel"], DenseVector(x["
    features"])))
sqlContext.createDataFrame(input_data, ["label", "
```

Split data

```
randomSplit([.8,.2], seed=1234)
```

Spark ML: ML pipeline Part 2

Train model

2 lr.fit()

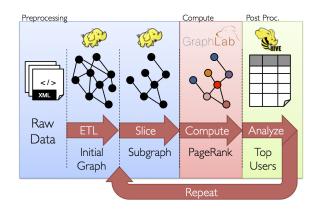
Make prediction

1 lr.transform()

Spark Streaming



Spark GraphX



Outline

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 - Introduction
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 - Spark GraphX
- Dask



Introduction to Dask

- Dask: Open source python framework for data processing
- developped with community projects like: Numpy, Pandas, and Scikit-Learn
- supported by: Anaconda, CapitalOne, NSF, Nvidia,...
- High-level collections:
 - Array, Bag, and DataFrame collections
 - mimic NumPy, lists, and Pandas
 - operate datasets out of core memory
- Low-Level schedulers :
 - dynamic task schedulers
 - execute task graphs in parallel



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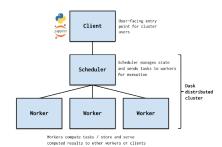




Dask Architecture

Dask architecture:

- Dask Cluster
- Dask Scheduler
- Dask collections



Dask Cluster

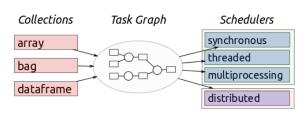
Various Dask cluster types:

- Hadoop/Spark clusters running YARN
- HPC clusters running job managers like SLURM, SGE, PBS, LSF, or others common in academic and scientific labs
- Kubernetes clusters

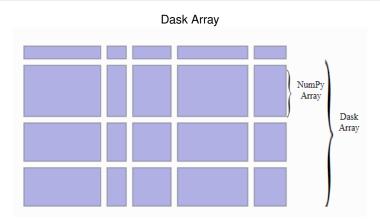
Dask Scheduler

Dask Scheduler:

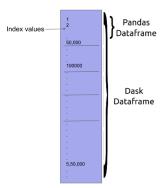
- Single machine scheduler
 Optimized for
 larger-than-memory use.
 Simple, easy and cheap
 to use, but does not scale
 as it only runs on a single
 machine.
- Distributed scheduler : More sophisticated, fully asynchronous



Dask collections



Dask DataFrame



Dask DataFrame



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Spark

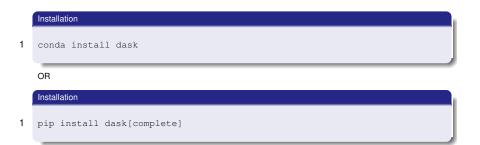
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Dask



Dask installation



Dask starting cluster

Lauching Dask cluster

```
1  from dask.distributed import LocalCluster, Client
2  cluster = LocalCluster()
3  client = Client(cluster)
```

Dashboard usually on http://localhost:8787/status

Dashboard

```
#To see where the port of the dashboard is, use this command
print(client.scheduler_info()['services'])
# {'dashboard': 8787} --> means you can access it at localhost:8787
```

Dask Collections

```
Dask Bag
```

```
1 import dask.bag as db
2 b = db.from_sequence([1, 2, 3, 4, 5, 6, 7, 8, 9, 10], npartitions=2)
```

Dask Array

```
1  import dask.array as da
2  x = da.random.random((10000, 10000), chunks=(1000, 1000))
```

Dask DataFrame

```
from dask import datasets
import dask.dataframe as dd
df = datasets.timeseries()
```

Dask Distributed

5

3

4

```
pask Delayed function

from dask import delayed
@delayed

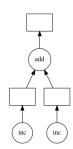
def inc(x):
    return x + 1
@delayed

def add(x, y):
    return x + y
```

```
Dask Lazy evaluation

x = inc(15)
y = inc(30)
total = add(x, y)

# execute all tasks
total.compute()
```



Dask Distributed

Dask Lazy evaluation

2

```
from dask.distributed import Client
c = Client(n_workers=4)

x = c.submit(inc, 1)
y = c.submit(dec, 2)
total = c.submit(add, x, y)
```

Dask Distributed

Dask evaluation # execute all tasks total.compute() Dask progess from dask.distributed import progress # to show progress bar progress(f)

Dask Distributed

```
Dask get results

# get result.
c.gather(f)

Dask persist
total.persist()
```

Dask Distributed

```
from dask.distributed import as_completed
def func(x):
    ...
    return y
futures = [c.submit(func, x) for x in range(n)]
iterator = as_completed(futures)
for res in iterator:
    print("RES_Y_:", res.result())
```

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Dask ecosystem

Ecosystem overview

Dask ecosystem overview:

- Dask tutorial :
 - https://github.com/dask/dask-tutorial.git
- Collection : Bag, Array, DataFrame
- Data Storage : CSV, HDF5, . . .
- Machine Learning : Scikit-learn, XGBoost,...
- Cluster : Local, SSH, YARN, ...

Spark

TPs

- TP 0 : Hadoop Installation
- TP 1 : WorldCount
- TP 2 : DataBase request
- TP 3 : Spark Installation
- TP 4 : Spark Compute PI
- TP 5 : Spark Image Processing
- TP 6 : Spark ML
- TP 7 : Dask

Introduction Hadoop Spark Dask TP

TP0: Hadoop Installation

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TP5: Spark ML, Machine learnin
TP6: Spark ML, Image processin
TP7: Start with Dask

TPs

TP0: Installation Hadoop

Hadoop Installation: hadoop-2.7.7.tar.gz

Installation

- 1 > cd /home/hduser
- 2 > mkdir local ; cd local
- 3 > wget https://www.apache.org/dist/hadoop/core/ hadoop-2.7.7/hadoop-2.7.7.tar.gz
- 4 > tar xvfz hadoop-2.7.7.tar.gz
- 5 > mv hadoop-2.7.7 hadoop
- 6 > chown -R hduser:hadoop hadoop

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TPs

TP0: Installation Hadoop

Env parameter settings

```
# Set JAVA_HOME (we will also configure JAVA_HOME directly for Hadoop later on)
export JAVA_HOME=/usr/local/Java/1.8.0-xxx

# Set Hadoop-related environment variables
export HADOOP_HOME=/home/hduser/local/hadoop

export HADOOP_CONF_DIR=${HADOOP_HOME}/etc/hadoop
export HADOOP_MAPRED_HOME=$HADOOP_HOME
export HADOOP_COMMON_HOME=$HADOOP_HOME
export HADOOP_COMMON_HOME=$HADOOP_HOME
export HADOOP_HOME=$HADOOP_HOME
export YARN_HOME=$HADOOP_HOME
export YARN_HOME=$HADOOP_HOME
```

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TP0: Installation Hadoop

Configuration files settings

```
1 # CREATE HADOOP TMP DIR
2 > mkdir -p /home/hduser/app/hadoop/tmp
3 > chown hduser:hadoop /home/hduser/app/hadoop/tmp
4 > chmod 750 /home/hduser/app/hadoop/tmp
5
6 # CREATE HDFS WORKINGDIR TO MNG HDFS File System
7 > mkdir -p /home/hduser/var/local/hadoop/hdfs/data
8 > chmod -R 777 /home/hduser/var/local/hadoop/hdfs
```

TPs

TP0: Installation Hadoop

Configuration files in /home/hduser/local/hadoop/etc/hadoop

hadoop-env.sh modification

```
JAVA_HOME="true_java_JOME_path"
export JAVA_HOME=${JAVA_HOME}
```

core-site.xml settings

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TP0: Installation Hadoop

Configuration files in /home/hduser/local/hadoop/etc/hadoop

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TP0: Installation Hadoop

Configuration files in /home/hduser/local/hadoop/etc/hadoop Copy mapred-site.xml.template mapred-site.xml

TPs

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TP0: Installation Hadoop

Configuration files in /home/hduser/local/hadoop/etc/hadoop

Introduction Hadoop Spark Dask **TP**

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TP1 : Hadoop World count

: Hadoop World count

P3 : Start with Spark

TP4: Spark ML, Data processing

TP6 : Spark ML, Image processing

TP7 : Start with Dask

TPs

TP0: Installation Hadoop

Lauch all services

> \$HADOOP_HOME/sbin/start-hdfs.sh
> \$HADOOP_HOME/sbin/start-yarn.sh

Check lauched services

1 > jps 2 3 26867

26867 DataNode

28228 Jps

27285 ResourceManager

26695 NameNode

27082 SecondaryNameNode

27420 NodeManager

- TP1: Hadoop World count





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TP1: MapReduce with Hadoop

Project MapReduce:

/home/hduser/BigDataHadoopSpark/TPs/TP1/MapReduce Two projects, A java Project and a python project

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TPs

TP1: MapReduce with Hadoop

Java project:

- create directory in hdfs /user/hduser/input
- copy the files of MapReduce/test/wordcount in /user/hduser/input
- generate Java project BigDataTP1
- 10 cd BigDataHadoopSpark/TPs/TP1/MapReduce
- 2 mvn package
- apply Java WordCount application
- check results



TP0: Hadoop Installation
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TP1: MapReduce with Hadoop

Python project

- create directory in hdfs /user/hduser/input/book
- copy the files of MapReduce/test/book in /user/hduser/input
- apply Python WordCount application
- check results

P0 : Hadoop Installation P1 : Hadoop World cour

TP2 : Hadoop DataBase request

P3: Start with Spark

P4 : Spark ML, Data processing

TP6 : Spark ML, Image processing

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TP3: Start with Spark





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TP0: Hadoop Installation TP1: Hadoop DataBase request TP2: Hadoop DataBase request TP3: Start with Spark TP4: Spark ML, Data processin TP5: Spark ML, Machine learnin TP6: Spark ML, Image processin TP7: Start with Dask

TPs

TP3: Installation Spark

Spark Installation: spark-2.4.0-bin-hadoop-2.7.tgz

Installation

```
1 > cd /home/hduser
2 > mkdir local; cd local
3 > wget https://www.apache.org/dyn/closer.lua/spark/spark-2.4.4/spark-2.4.4-bin-hadoop2.7.tgz
4 > tar xvfz spark-2.4.0-bin-hadoop-2.7.7.tar.gz
5 > mv spark-2.4.0-bin-hadoop-2.7.7 spark
6 > export HADOOP_CONF_DIR=$HADOOP_HOME/etc/hadoop
7 > export SPARK_HOME=/home/hduser/local/spark
8 > export PATH=$SPARK_HOME/bin:$PATH
9 > export LD_LIBRARY_PATH=$HADOOP_HOME/lib/native:$LD_LIBRARY_PATH
```

TP0: Hadoop Installation TP1: Hadoop World count TP2: Hadoop DataBase request TP3: Start with Spark TP4: Spark ML, Data processing TP5: Spark ML, Machine learnin TP6: Spark ML, Image processin TP7: Start with Dask

TPs

TP3: Installation PySpark

Installation

```
1 > cd /home/hduser
2 > export SPARK_HOME=/home/hduser/local/spark
3 > export PATH=$SPARK_HOME/bin:$PATH
4 > export LD_LIBRARY_PATH=$RADOOP_HOME/lib/native:$LD_LIBRARY_PATH
5 > export PYSPARK_PYTHON="path_to_python"
6 > pip install pyspark
7 > pip install findspark
8 > sbin/start-master.sh
9 > sbin/start-slave.sh spark://localhost:7077
```

TP0: Hadoop Installation TP1: Hadoop World count TP2: Hadoop DataBase request TP3: Start with Spark TP4: Spark ML, Data processing TP5: Spark ML, Machine learning TP6: Spark ML, Image processin TP7: Start with Dask

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3

8

9 10 11

12

13

14 15

16

17

18 19

TP3: Installation PySpark

test Spark shell

```
> spark-shell
Setting default log level to "WARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
Spark context Web UI available at http://localhost:4040
Spark context available as 'sc' (master = local[\star], app id = local-1578948405576).
Spark session available as 'spark'.
Welcome to
   \\/\\/\'\/\'/
____/____/\___/\__/\_\_\_version_3.0.0-preview2
___/_/
Using Scala version 2.12.10 (Java HotSpot (TM) 64-Bit Server VM, Java 1.8.0 92)
Type in expressions to have them evaluated.
Type :help for more information.
scala>
```

TPs

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11 12 13

14

15

16 17

TP3: Installation PySpark

test Spark shell

```
> pyspark
Python 2.7.5 (default, Apr 11 2018, 07:36:10)
[GCC 4.8.5 20150623 (Red Hat 4.8.5-28)] on linux2
Type "help", "copyright", "credits" or "license" for more information.
Setting default log level to "WARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
/work/irlin355 1/gratieni/BigData/local/spark/spark-3.0.0-preview2-bin-hadoop2.7/pvthon/pvspark/
      context.py:219: DeprecationWarning: Support for Python 2 and Python 3 prior to version 3.6 is
      deprecated as of Spark 3.0. See also the plan for dropping Python 2 support at https://spark.
      apache.org/news/plan-for-dropping-python-2-support.html.
 DeprecationWarning)
Welcome to
   \\/\\/\'\/\'\/\/
____/___/\___/\___version_3.0.0-preview2
___/_/
Using Python version 2.7.5 (default, Apr 11 2018 07:36:10)
```

TP0: Hadoop Installation
TP1: Hadoop World count
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TP3: Start with Spark
TP4: Spark ML, Data processing
TP5: Spark ML Machine learning

TPs

TP3: Test0 Test1 Test2

Test0:

- create spark context
- create liste of integer
- partition list with spark
- print num of partions

Test1:

- compute square of integer list
- print square list

Test2:

compute PI



- TP4: Spark ML. Data processing





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TPs

TP4: Spark ML, Data processing

Test0:

- load TPs/data/iris.csv file in Panda DataFrame
- create Spark DataFrame
- show 5 first lines
- select two columns
- print some statistics on Spark Data frame

- TP5: Spark ML, Machine learning





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TP5: Spark ML, Data processing

Spark ML:

- load TPs/data/iris.csv file in Panda DataFrame
- create Spark DataFrame
- create Pipeline to prepare date for machine learning
- compute a predicting model
- evaluate the predicting model

TP6: Spark ML. Image processing





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TPO: Hadoop Installation
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Project: Spark ML, Image processing

project:

- load Lena.jpg file
- develop a parallel median Filter in python with Spark

TP7: Start with Dask





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Test0 Test1 Test2 with Dask

Test0:

- create Dask client
- create liste of integer
- partition list with dask
- print num of partions

Test1:

- compute square of integer list
- print square list

Test2:

compute PI

