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BIG DATA AND GEOSPATIAL DATA MINING

3 DISTRIBUTED DATA PROCESSING

- Popularized by Google from the publication :
J. Dean and, S. Ghemawat, "MapReduce: Simplified Data Processing on Large Clusters", Communications of the ACM, Jan 2008, VI 51 No. 1.

3.1 MapReduce

A. Martín

NODE 04

```

error_mod.use = true
error_mod.use = true
error_mod.use = false
operation
  
```

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0. (**Splitting**). Data splitting (if necessary and possible)
1. (**Map**). Design a set of simple "Map" tasks that generate a set of intermediate results on an input data partition that are pairs of the type: (key, value)
2. (**Shuffle and Sort**). The intermediate results (key, value) are grouped and sorted (by key).
3. (**Reduce**). The pairs ordered by key are processed by another simple set of tasks "Reduce" to produce the result.

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Text file where each line contains meteorological measurements of a station in a country
(id_country, year, temp_mean, temp_max, temp_min)

File	Map	Shuffle & Sort & Group	Reduce	Output
21 1904 25 37 -24 12 1915 13 42 -11 14 1902 17 44 -46 13 1904 21 39 -54 45 1902 57 41 -93 13 1907 21 37 -31 ...	(1904, 37) (1915, 42) (1902, 44) (1904, 39) (1902, 41) ...	(1902, [44, 41]) (1904, [37, 39]) (1915, 42) ...	(1902, 44) (1904, 39) (1915, 42) ...	1902, 44 1904, 39 1915, 42 ...

cat data.txt | mapper.py | sort | reducer.py > salida.txt

(1902, 44)
(1902, 41)
(1904, 37)
(1904, 39)
(1915, 42)

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- Specially designed to work on huge amounts of **distributed data**

Suppose we want to process data from a database, where would we execute its operations?

Desktop system Client-Server System Distributed system

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Splitting

Nodo 1

21	1904	25	37	-24
12	1915	13	42	-11
14	1902	17	44	-46

Map

(1904, 37)
(1915, 42)
(1902, 44)

Nodo 2

13	1904	21	39	-54
45	1902	57	41	-93

Map

(1904, 39)
(1902, 41)

Shuffling

Sort & Group

(1902, [44, 41])
(1904, [37, 39])
(1915, 42)
...

Reduce

(1902, 44)
(1904, 39)
(1915, 42)
...

Output

1902, 44
1904, 39
1915, 42
...

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- MapReduce is a distributed programming environment on distributed data
- Reduces the natural complexity of a distributed system
- Take advantage of the local storage of the data
- It allows to process large amount of data efficiently

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EXERCISE: Shopping data set, purchases.txt file

```
angel@angel-VirtualBox:~/Documents/BigData/mapreduce/purchases$ head -n 5 purchases.txt
date, time, store, item, cost, payment
2012-01-01,09:00,San Jose,Men's Clothing,214.05,Amex
2012-01-01,09:00,Fort Worth,Women's Clothing,153.57,Visa
2012-01-01,09:00,San Diego,Music,66.08,Cash
2012-01-01,09:00,Pittsburgh,Pet Supplies,493.51,Discover
angel@angel-VirtualBox:~/Documents/BigData/mapreduce/purchases$
```

Calculate total values by store (by city)

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MAPPER.PY

```

1  #!/usr/bin/python
2
3  # Format of each line is:
4  # date,time,store name,item description,cost,method of payment
5
6  import sys
7
8  for line in sys.stdin:
9      data = line.strip().split(",")
10     if len(data) == 6:
11         date, time, store, item, cost, payment = data
12         try:
13             print ("{}},{1}".format(str(store), float(cost)))
14         except:
15             pass

```

Ln: 15 Col: 16

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REDUCER.PY

```

1  #!/usr/bin/python
2
3  import sys
4
5  salesTotal = 0
6  oldKey = None
7
8  for line in sys.stdin:
9      data_mapped = line.strip().split(",")
10     if len(data_mapped) != 2:
11         # Confusión, no lee esta línea.
12         continue
13
14     thisKey, thisSale = data_mapped
15
16     if oldKey != None and oldKey != thisKey:
17         print (oldKey, salesTotal)
18         salesTotal = 0
19
20     oldKey = thisKey
21     salesTotal += float(thisSale)
22
23 if oldKey != None:
24     #El ultimo se queda fuera del bucle al no haber nueva línea para comprobar el oldKey
25     print (oldKey, salesTotal)

```

Ln: 23 Col: 0

cat purchases.txt | python3 mapper.py | sort | python3 reducer.py > salida.txt

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