# **Warehouse Scale Computing**

### 1. Amdahl's Law

1) You are going to train the image classifier with 50,000 images on a WSC having more than 50,000 servers. You notice that 99% of the execution can be parallelized. What is the speedup?

$$1/(0.01 + 0.99 / 50,000) \approx 1/0.01 = 100$$

#### 2. Failure in a WSC

1) In this example, a WSC has 55,000 servers, and each server has four disks whose annual failure rate is 4%. How many disks will fail per hour?

$$(55,000 \times 4 \times 0.04) / (365 \times 24) = 1.00 \rightarrow MTTF = 1 \text{ hour}$$

2) What is the availability of the system if it does not tolerate the failure? Assume that the time to repair a disk is 30 minutes.

MTTF = 1, MTTR = 
$$0.5 \rightarrow \text{Availability} = 1 / (1 + 0.5) = 2/3 = 66.6\%$$

#### 3. Performance of a WSC

	Local	Rack	Array
DRAM latency (us)	0.1	100	300
Global hit rate	90%	9%	1%
DRAM bandwidth (MiB/sec)	20,000	100	10
Disk bandwidth (MiB/ sec)	200	100	10

1) Calculate the AMAT of this WSC. What is vital for WSC performance?

$$AMAT = 0.9 \times 0.1 + 0.09 \times 100 + 0.01 \times 300 = 0.09 + 9 + 3 = 12.09 \text{ us}$$

Locality of access within a server is vital for WSC performance

- 2) How long does it take to transfer 1,000 MiB a) between disks within the server, and b) between DRAM within the rack? What can you conclude from this example?
- a) 1,000 / 200 = 5 sec, b) 1,000 / 100 = 10 sec. Data transfer outside a single server is detrimental to WSC performance. Network switches are the bottlenecks
- **4. Power Usage Effectiveness (PUE) = (**Total Building Power) / (IT Equipment Power)

Sources speculate Google has over 1 million servers. Assume each of the 1 million servers draw an average of 200W, the PUE is 1.5, and that Google pays an average of 6 cents per kilowatt-hour for datacenter electricity.

- 1) Estimate Google's annual power bill for its datacenters.
- $1.5 \times 1,000,000 \text{ servers } \times 0.2 \text{kW/sever } \times \$0.06/\text{kW-hr} \times 8760 \text{ hrs/yr} = \$157.68 \text{ M/yr}$
- 2) Google reduced the PUE of a 50,000 machine datacenter from 1.5 to 1.25 without decreasing the power supplied to the servers. What's the cost savings per year?

```
(1.5 - 1.25) \times 50,000 \text{ servers } \times 0.2 \text{kW/server } \times \$0.06/\text{kW-hr} \times 8760 \text{ hrs/yr} = \$1.314 \text{M/yr}
```

## Map Reduce

Use pseudocode to write MapReduce functions necessary to solve the problems below. Also, make sure to fill out the correct data types. Some tips:

- The input to each MapReduce job is given by the signature of the map() function.
- The function emit(key k, value v) outputs the key-value pair (k, v).
- The **for(var in list)** syntax can be used to iterate through **Iterable**s or you can call the **hasNext()** and **next()** functions.
- Usable data types: **int**, **float**, **String**. You may also use lists and custom data types composed of the aforementioned types.
- The method **intersection(list1, list2)** returns a list that is the intersection of list1 and list2.

1. Given the student's name and the course taken, output each student's name and total GPA.

```
Declare any custom data types here:

CourseData:
    int courseID
    float studentGrade // a number from 0-4

map(String student, CourseData value):
    emit(student, value.studentGrade)

reduce( String key,
    Iterable< float > values):
    totalPts = 0
    totalClasses = 0
    for ( grade in values ):
        totalPts += grade
        totalClasses++
    emit(key, totalPts / totalClasses)
```

2. Given a person's unique int ID and a list of the IDs of their friends, compute the list of mutual friends between each pair of friends in a social network.

```
Declare any custom data types here:
FriendPair:
 int friendOne
 int friendTwo
map(int personID, list<int> friendIDs):
                                          reduce( FriendPair key,
  for ( fID in friendIDs ):
                                               Iterable< list<int> > values):
   if ( personID < fID ):</pre>
                                            mutualFriends =
     friendPair = ( personID, fID )
                                              intersection(
                                                values.next(), values.next())
   else:
                                            emit(key, mutualFriends)
     friendPair = ( fID, personID )
   emit(friendPair, friendIDs)
```

3. a) Given a set of coins and each coin's owner, compute the number of coins of each denomination that a person has.

b) Using the output of the first MapReduce, compute the amount of money each person has. The function valueOfCoin(String coinType) returns a float corresponding to the dollar value of the coin.

```
map(CoinPair key, int amount):
    emit(coinPair.person,
    valueOfCoin(coinPair.coinType)*amount)

for ( amount in values ):
    total += amount
    emit(key, total)
```

• RDD: primary abstraction of a distributed collection of items

• Transforms: RDD → RDD

<pre>map(func)</pre>	Return a new distributed dataset formed by passing each element of the source through a function <i>func</i> .
flatMap(func)	Similar to map, but each input item can be mapped to 0 or more output items (so <i>func</i> should return a Seq rather than a single item).
<pre>reduceByKey(func )</pre>	When called on a dataset of (K,V) pairs, returns a dataset of (K,V) pairs where the values for each key are aggregated using the given reduce function func, which must be of type (V,V) => V.

• Actions: RDD → Value

reduce(func)	Aggregate the elements of the dataset <i>regardless of keys</i> using a function <i>func</i>
--------------	--

1. Implement Problem 1 of MapReduce with Spark

2. Implement Problem 2 of MapReduce with Spark

3. Implement Problem 3 of MapReduce with Spark