## CS 61C WSC, MapReduce, Spark Summer 2019 <br> Discussion 12: August 5, 2019

## 1 MapReduce

For each problem below, write pseudocode to complete the implementations. Tips:

- The input to each MapReduce job is given by the signature of map().
- emit (key k, value v) outputs the key-value pair (k, v).
- for var in list can be used to iterate through Iterables 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.
- intersection(list1, list2) returns a list of the intersection of list1, list2.
1.1 Given a set of coins and each coin's owner, compute the number of coins of each denomination that a person has.

Declare any custom data types here:
CoinPair:
String person
String coinType

1 $\qquad$
$\qquad$ ):

1 reduce( $\qquad$ , $\qquad$

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

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## 2 Spark

Resilient Distributed Datasets (RDD) are the primary abstraction of a distributed collection of items

Transforms $R D D \rightarrow R D D$
$\operatorname{map}(f)$ Return a new dataset formed by calling $f$ on each source element.
flatMap $(f)$ Similar to map, but each input item can be mapped to 0 or more output items (so $f$ should return a sequence rather than a single item).
reduceByKey $(f)$ 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 $f$, which must be of type $(V, V) \rightarrow V$.

## Actions $R D D \rightarrow$ Value

reduce $(f)$ Aggregate the elements of the dataset regardless of keys using a function $f$.

Call sc. parallelize(data) to parallelize a Python collection, data.
Given a set of coins and each coin's owner, compute the number of coins of each denomination that a person has. Then, using the output of the first result, compute each person's amount of money. Assume valueOfCoin(coinType) is defined and returns the dollar value of the coin.

The type of coinPairs is a list of (person, coinType) pairs.

```
coinData = sc.parallelize(coinPairs)
```


## 3 Warehouse-Scale Computing

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

Estimate Google's annual power bill for its datacenters.

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?

## 4 MapReduce/Spark Practice: Optimize Your GPA

Given the student's name and course taken, output their name and total GPA.
Declare any custom data types here:

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

$1 \operatorname{map}($
$\qquad$ , ------------------ ) ): 1 reduce ( $\qquad$ , $\qquad$ _):

Solve the problem above using Spark.
The type of students is a list of (studentName, courseData) pairs.
1 studentsData = sc.parallelize(students)
2 $\qquad$ )))

## 5 MapReduce/Spark Practice: Optimize the Friend Zone

5.1 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. You have access to the intersection function, which takes in two lists finds the set of elements that appear in both lists.

Declare any custom data types here:
FriendPair:
int friendOne
int friendTwo

1 map(int personID, list<int> friendIDs):


Solve the problem above using Spark.
The type of persons is a list of (personID, list(friendID) pairs.

```
def genFriendPairAndValue(pID, fIDs):
    return [((pID, fID), fIDs) if pID < fID else (fID, pID) for fID in fIDs]
def intersection(l1, l2):
    return [x for x in b1 if x in b2]
personsData = sc.parallelize(persons)
```

