# Class 6: Pair RDDs

**New York University** 

**Summer 2017** 

- Key-Value Pair RDDs
- Map-Reduce
- Other Pair RDD Operations

#### What are Pair RDDs?

- Each RDD element must be a key-value pair (a two-element tuple)
- Keys and values can be any type
- Use with map-reduce algorithms
- Many additional functions are available for common data processing needs
  - -e.g., sorting, joining, grouping, counting, etc.

# Pair RDD (key1, value1) (key2, value2) (key3, value3) ...

# Common functions for creating Pair RDDs

- -map
- -flatMap / flatMapValues
- -keyBy

## Example: Create a Pair RDD from a tab-separated file

```
> val users = sc.textFile(file) \
   .map(line => line.split('\t')) \
   .map(fields => (fields(0), fields(1)))
```

```
user001\tFred Flintstone
user090\tBugs Bunny
user111\tHarry Potter
```

```
(user001,Fred Flintstone)
(user090,Bugs Bunny)
(user111,Harry Potter)
...
```

```
> sc.textFile(logfile) \
   .keyBy(line => line.split(' ')(2))
```

#### User ID

```
56.38.234.188 - 99788 "GET /KBDOC-00157.html HTTP/1.0" ...
56.38.234.188 - 99788 "GET /theme.css HTTP/1.0" ...
203.146.17.59 - 25254 "GET /KBDOC-00230.html HTTP/1.0" ...
...
```

```
(99788,56.38.234.188 - 99788 "GET /KBDOC-00157.html...)
(99788,56.38.234.188 - 99788 "GET /theme.css...)
(25254,203.146.17.59 - 25254 "GET /KBDOC-00230.html...)
...
```

### How would you do this?

- Input: a list of postal codes with latitude and longitude
- Output: postal code (key) and lat/long pair (value)

```
00210 43.005895 -71.013202
00211 43.005895 -71.013202
00212 43.005895 -71.013202
00213 43.005895 -71.013202
00214 43.005895 -71.013202
```

```
(00210, (43.005895, -71.013202))
(00211, (43.005895, -71.013202))
(00212, (43.005895, -71.013202))
(00213, (43.005895, -71.013202))
...
```

```
> sc.textFile("file").
  map(line => line.split('\t')).
  map(fields => (fields(0),(fields(1),fields(2))))
```

- Key-Value Pair RDDs
- Map-Reduce
- Other Pair RDD Operations

- Map-reduce is a common programming model
  - Easily applicable to distributed processing of large data sets
- Hadoop MapReduce is a well known implementation
  - Somewhat limited
    - Each job has one Map phase, one Reduce phase
    - Job output is saved to files
- Spark implements map-reduce with much greater flexibility
  - Map and reduce functions can be interspersed
  - Results can be stored in memory
    - Operations can easily be chained

#### Map-reduce in Spark works on Pair RDDs

#### Map phase

- Operates on one record at a time
- Maps each record to one or more new records
- -e.g. map, flatMap, filter, keyBy

#### Reduce phase

- Works on map output, or output from other tools in same keyvalue format
- Consolidates multiple records
- -e.g. reduceByKey, sortByKey, mean

## How would you do this?

Input: Text lines

- Output: Word count

#### Input Data

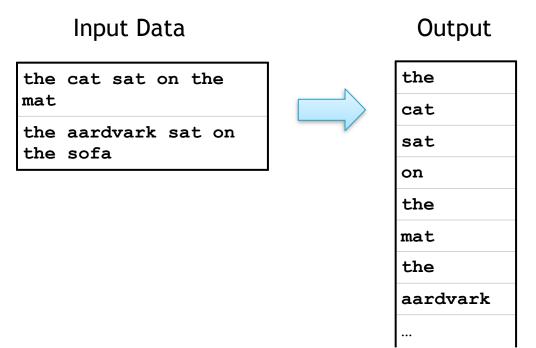
the cat sat on the mat the aardvark sat on the sofa



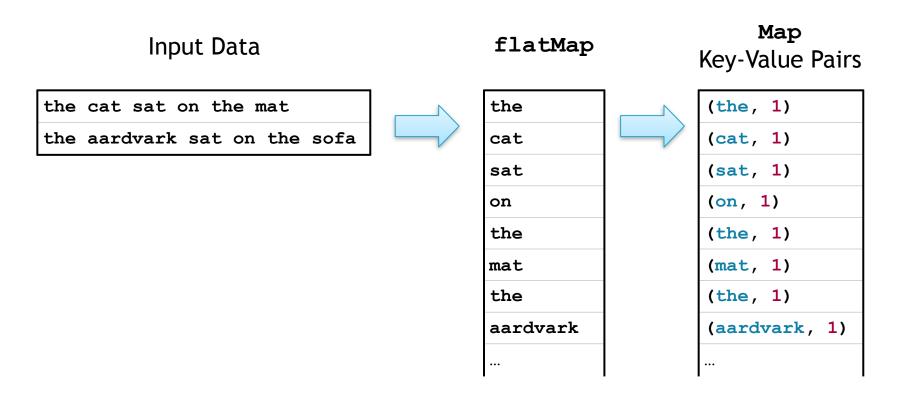
#### Output

aardvark	1
cat	1
mat	1
on	2
sat	2
sofa	1
the	4

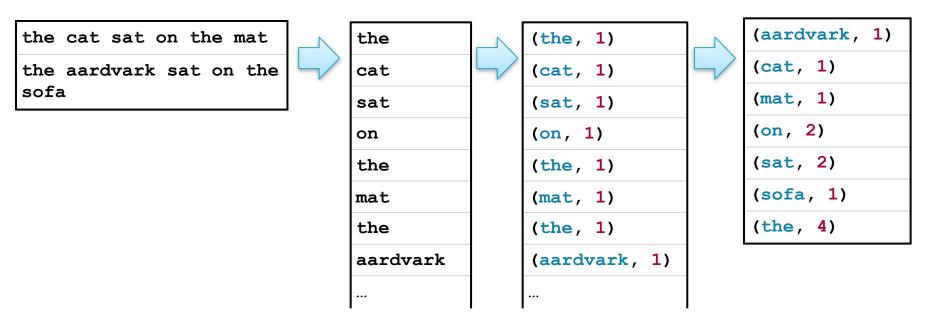
#### Use flatMap



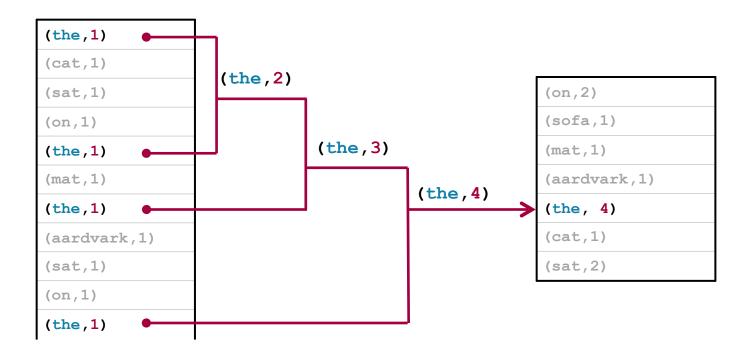
- flatMap, followed by map to map the word as key and output with a '1'
  - This is the same technique you use with Hadoop MapReduce when implementing WordCount



- After flatMap and map we have key-value pairs that can be the input to a reduce
- followed by reduceByKey specifying addition of two values, and that sum gets added to another value, and so on:



- The function passed to reduceByKey combines values from two keys
  - Function must be binary



```
> val counts = sc.textFile(file).
flatMap(line => line.split("\\W")).
map(word => (word, 1)).
reduceByKey((v1, v2) => v1 + v2)
```

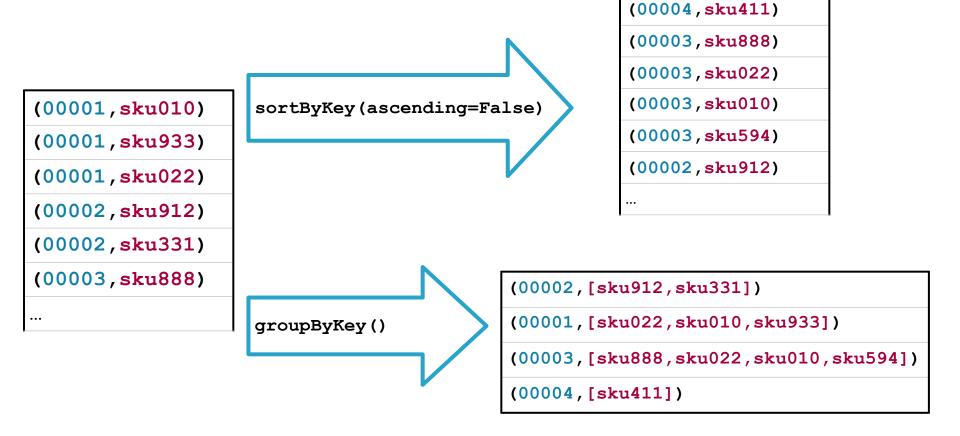
OR

```
> val counts = sc.textFile(file).
  flatMap(_.split("\\W")).
  map((_, 1)).
  reduceByKey(_ + _)
```

- Word count is challenging over massive amounts of data
  - Using a single compute node would be too time-consuming
  - Number of unique words could exceed available memory
- Statistics are often simple aggregate functions
  - Distributive in nature
  - -e.g., max, min, sum, count
- Map-reduce breaks complex tasks down into smaller elements which can be executed in parallel
- Many common tasks are very similar to word count
  - -e.g., log file analysis

- Key-Value Pair RDDs
- Map-Reduce
- Other Pair RDD Operations

- In addition to map and reduce functions, Spark has several operations specific to Pair RDDs
  - -countByKey
    - Return a map with the count of occurrences of each key
  - -groupByKey
    - Group all the values for each key in an RDD
  - sortByKey
    - Sort in ascending or descending order
  - -join
    - Return an RDD containing all pairs with matching keys from two RDDs



#### Using join

```
(Casablanca, $3.7M)
(Star Wars, $775M)
(Annie Hall, $38M)
(Argo, $232M)
...
```

```
(Casablanca,1942)
(Star Wars,1977)
(Annie Hall,1977)
(Argo,2012)
...
```

```
{\tt RDD:} \ {\tt movieGrossAndYearForEachMovie}
```

```
(Casablanca, ($3.7M,1942))
(Star Wars, ($775M,1977))
(Annie Hall, ($38M,1977))
(Argo, ($232M,2012))
...
```

#### Some other pair RDD operations

- -keys return an RDD of just the keys, without the values
- -values return an RDD of just the values, without keys
- -lookup(key) return the value(s) for a key
- -leftOuterJoin, rightOuterJoin, fullOuterJoin join, including keys defined in the left, right or either RDD respectively
- -mapValues, flatMapValues execute a function on just the values, keeping the key the same
- See the PairRDDFunctions class Scaladoc for a full list

#### Homework

Study for midterm exam.