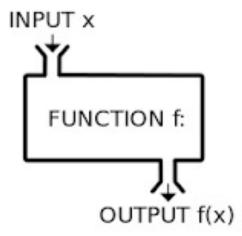
SPARK ESSENTIALS

Review of concepts

- What is Functional Programming Python
 - Lambda
 - Map, Filter, Reduce
- List Comprehensions
- Sliced

Functional Programming

- Functional programming is a **Style** whose underlying model of computation is the *function*.
- Functions take input and produce output, without any side effects.
- No state
- Immutable data
- Function as first-class citizen
- Recursion
- Purity ...



https://marcobonzanini.com/2015/06/08/functional-programming-in-python/



Python Functional Programming

Python is a multi paradigm programming language. As a Python programmer why uses functional programming in Python?

Python is not a functional language but have a lot of features that enables us to applies functional principles in the development, turning our code more elegant, concise, maintanable, easier to understand and test.

Lambda function

- Syntax : lambda argument_list: expression
 - argument_list: comma separated list of arguments
 - expression is an arithmetic expression using these arguments.
- The function can be assigned to a variable

Lambda function

Example

```
>> f = lambda x,y: x + y
>> f(1,2)
Out[1]: 3
```

- Only one expression in the lambda body
- Advantage of the lambda can be seen when it is used in combination with other functions (e.g. map, reduce, filter)

The map() function

- Syntax: r = map(func, seq)
 - func is the name of a function
 - Seq is a sequence (e.g. a list)
- Map applies the function func to all the elements of the sequence seq and returns a new list with the elements changed by func

The map() function

```
Example(1):
>> nums = [1,2,3,4]
>> squares = map(lambda x: x * x, nums)
>> print squares
Out[1]: [1, 4, 9, 16]
Example(2):
>> Celsius = [39.2, 36.5, 37.3, 37.8]
>> Fahrenheit = map(lambda x: (float(9) / 5) * x + 32, Celsius)
>> print Fahrenheit
Out[1]: [102.56, 97.70000000000003, 99.1400000000001, 100.039999999999]
\rightarrow C = map(lambda x: (float(5) / 9) * (x - 32), Fahrenheit)
>> print C
Out[1]: [39.20000000000003, 36.5, 37.3000000000004, 37.7999999999999] >>>
```

The filter() function

- Syntax: f = filter(function, list)
 - Function that returns true or false applied to every element of the list
 - List to applied the function

 Filter returns a new list with the "True" elements returned from applying the function to the list

The filter() function

Example:

```
>> fib= [0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55]
>> result= filter(lambda x: x % 2, fib)
>> print result
Out[1]: [1, 1, 3, 5, 13, 21, 55]
```

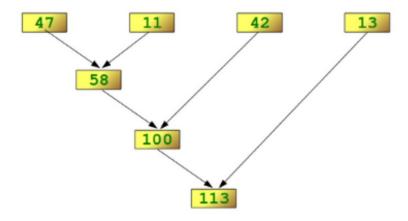
The reduce() function

- Syntax: r = reduce(func, seq)
 - func is the name of a function
 - Seq is a sequence (e.g. a list)
- Reduce continually applies the function func to the sequence seq, and returns a single value.

The reduce() function

Example:

```
>> result= reduce(lambda x,y: x + y, [47, 11, 42, 13])
Out[1]: 113
```



List Comprehensions

```
doubled odds = []
 for n in numbers:
     if n % 2 == 1:
         doubled odds.append(n * 2)
 doubled odds = [n * 2 for n in numbers if n % 2 == 1]
We copy-paste from a for loop into a list comprehension by:
1. Copying the variable assignment for our new empty list
2. Copying the expression that we've been append -ing into this new list
3. Copying the for loop line, excluding the final:
4. Copying the if statement line, also without the :
```

List slicing

```
    Slicing: Extracting parts of list

  Syntax:
               list[start:end]
               list[start:]
               list[end:]
               list[:]

    start inclusive and excluding end

  Slicing returns a new list
                        >>> colors = ['yellow', 'red', 'blue', 'green', 'black']
                        >>> colors[0:]
                         ['yellow', 'red', 'blue', 'green', 'black']
                        >>> colors[:4]
                         ['yellow', 'red', 'blue', 'green']
                        >>> colors[1:3]
                         ['red', 'blue']
                        >>> colors[:]
                         ['yellow', 'red', 'blue', 'green', 'black']
```

Spark Essentials

- SparkContext
- RDDs
- Operations:
 - Basic transformations
 - Basic actions
- Persistence

SparkContext

- Main entry point to Spark functionality
- Created for you in Spark shells and notebooks as variable sc
- In standalone programs, you'd make your own
 - (Last slide)
 - Now we are assuming that we are working either in the shell or with notebooks

Creating RDDs

```
# Turn a local collection into an RDD
>> rdd 1 = sc.parallelize([1, 2, 3])
# Load text file from local FS, HDFS, or S3
>> rdd 2 = sc.textFile("file.txt")
>> rdd 3 = sc.textFile("directory/*.txt")
>> rdd 4 =
sc.textFile("hdfs://namenode:9000/path/file"
# Transforming an existing RDD
>> rdd 5 = rdd2.filter(function)
```

RDD operations

- Transformations
 - lazy operation to build RDI
- Actions
 - Computes a result based on existing RDD or write it to storage

Transformations

```
map (func)
flatMap(func)
filter(func)
groupByKey()
reduceByKey(func)
mapValues(func)
sample(...)
union(other)
distinct()
sortByKey()
...
```

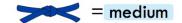
Actions

```
reduce(func)
collect()
count()
first()
take(n)
saveAsTextFile(path)
countByKey()
foreach(func)
...
```



sample

randomSplit



Essential Core & Intermediate Spark Operations

General

Math / Statistical

Set Theory / Relational

Data Structure / I/O

- map
- filter
- flatMap
- mapPartitions
- mapPartitionsWithIndex
- groupBy
- sortBy

- union
- intersection
- subtract
- distinct
- cartesian
- zip

keyBy

- zipWithIndex
- zipWithUniqueID
- zipPartitions
- coalesce
- repartition
- repartitionAndSortWithinPartitions
- pipe



- reduce
- collect
- aggregate
- fold
- first
- take
- forEach
- top
- treeAggregate
- treeReduce
- forEachPartition
- collectAsMap

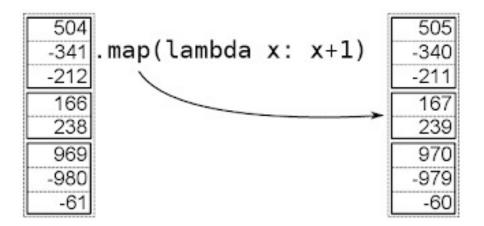
- count
- takeSample
- max
- min
- sum
- histogram
- mean
- variance
- stdev
- sampleVariance
- countApprox
- countApproxDistinct

- takeOrdered
- saveAsTextFile
- saveAsSequenceFile
- saveAsObjectFile
- saveAsHadoopDataset
- saveAsHadoopFile
- saveAsNewAPIHadoopDataset
- saveAsNewAPIHadoopFile

pySpark map(), filter(), reduce()

Note: Different syntax than Python

- Map syntax:
 - rdd.map(function)
- Filter syntax:
 - rdd.filter(function)
- Reduce syntax:
 - rdd.reduce(function)



Passing functions to Spark

 With lambda syntax allows us to define "simple" functions inline. But we can pass defined functions.

```
def hasHadoop( line ):
    return "Hadoop" in line

>> lines = sc.textFile( "README.txt" )
>> hadoopLines = lines.filter( hasHadoop )
```

Basic Transformations

```
nums = sc.parallelize([1, 2, 3])
# Pass each element through a function
squares = nums.map(lambda x: x * x) # => [1, 4, 9]
# Keep elements passing a predicate
even = squares.filter(lambda x: \times % 2 == 0) # => [4]
# Map each element to zero or more others
nums.flatMap(lambda x: range(0, x)) # => [0, 0, 1,
[0, 1, 2]
# Map each element to zero or more others
nums.map(lambda x: range(0, x)) \# = [0], [0, 1],
|0, 1, 2||
```

map() vs flatMap()

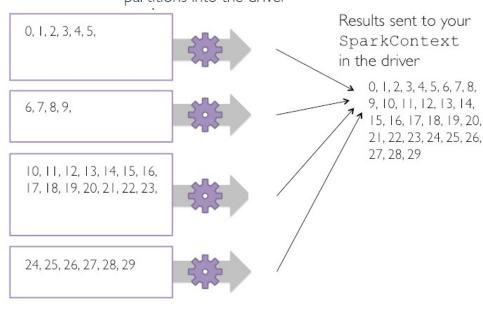
flatMap: Similar to map, it returns a new RDD by applying a function to each element of the RDD, but output is flattened.

```
>> rdd = sc.paralleliz([1, 2, 3])
>> rdd.map(lambada x: [x, x * 2])
Out[1]: [ [1, 2], [2, 4], [3, 6]]
>> rdd.flatMap(lambada x: [x, x * 2])
Out[1]: [ 1, 2, 2, 4, 3, 6]
```

Basic Actions

```
nums = sc.parallelize([5, 1, 3, 2])
# Retrieve RDD contents as a local collection
 → Results must fit in memory on the local machine
nums.collect() \# \Rightarrow [5, 1, 3, 2]
# Return first K elements
nums.take(2) \# \Rightarrow [5, 1]
# Return first K elements ordered
nums.takeOrdered(4) # => [1, 2, 3, 5]
# Return first K elements by applying
#a particular order
nums.takeOrdered(4, lamda n:-n)
\# \Rightarrow [5, 3, 2, 1]
# Count number of elements
nums.count() # => 4
# Merge elements with an associative function
nums.reduce(lambda x, y: x + y) # => 12
# Write elements to a text file
nums.saveAsTextFile("hdfs://file.txt")
```

collect(): Gathers the entries from all partitions into the driver



Persistence

- Spark recomputes the RDDs each time we call an action → expensive and can also cause data to be read from the disk again
- We can avoid this caching the data:
 - caching ()
 - persintace()
- Fault tolerant: In case of failure, Spark can rebuild the RDD
- Super Fast: will allow multiple operations on the same data set

Persistence

- RDDs can be cached using <u>cache</u> operation.
 They can also be persisted using <u>persist</u> operation.
- With cache(), you use only the default storage level MEMORY ONLY.
- With persist(), you can specify which storage level you want, (<u>rdd-persistence</u>).
- Use persist() if you want to assign another storage level than MEMORY_ONLY to the RDD (which storage level to choose)

Persistence Example

```
>> lines = sc. textFile("README.md", 4)
                                            >> lines = sc. textFile("README.md", 4)
                                            >> lines.persist() # ~lines.cache()
>> lines.count()
                                            >> lines.count()
Out[1]: 1024
                                           Out[1]: 1024
                                            >> pythonLines = lines.filter(lambda line:
>> pythonLines = lines.filter(lambda
                                            "Python" in line)
line: "Python" in line)
                                            >> pythonLines.count()
>> pythonLines.count()
                                           Out[1]: 50
Out[1]: 50
Causes Spark to reload lines from disk used.
                                            Spark will avoid re-computing lines every
                                            time it is used.
```

SparkContext - Cluster execution

A **SparkContext** represents the connection to a Spark cluster.

- Interactive Shell/ Notebooks -- the spark context is created for you.
- Spark Application running in a Apache Cluster -- you have to create it