# Practice 8 PageRank

# What is Scala

## > Scala

Modern multi-paradigm programming language that is a combination of object-oriented and functional programming.
 It is highly scalable which is why it is called Scala.

## > Scala Features

- Object- Oriented: Every value in Scala is an object so it is a purely object-oriented programming language. The behavior and type of objects are depicted by the classes and traits in Scala.
- Functional: It is also a functional programming language as every function is a value and every value is an object. It provides the support for the high-order functions, nested functions, anonymous functions etc.
- Statically Typed: The process of verifying and enforcing the constraints of types is done at compile time in Scala. Unlike other statically typed programming languages like C++, C etc., Scala doesn't expect the redundant type of information from the user. In most cases, the user has no need to specify a type.

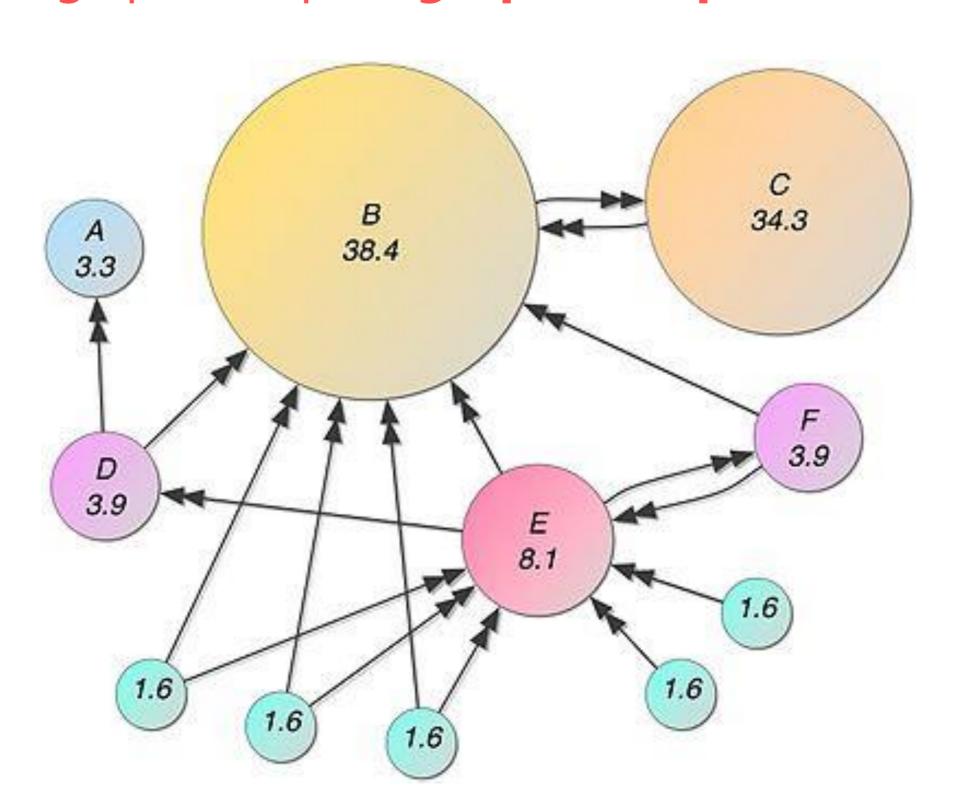
# What is Scala

## > Scala example

```
scala> textFile.map(line => line.split(" ").size).reduce((a, b) => if (a > b) a else b)
res4: Long = 15
scala> import java.lang.Math
import java.lang.Math
scala> textFile.map(line => line.split(" ").size).reduce((a, b) => Math.max(a, b))
res5: Int = 15
scala> val wordCounts = textFile.flatMap(line => line.split(" ")).groupByKey(identity).count()
wordCounts: org.apache.spark.sql.Dataset[(String, Long)] = [value: string, count(1): bigint]
scala> wordCounts.collect()
res6: Array[(String, Int)] = Array((means,1), (under,2), (this,3), (Because,1), (Python,2), (agree,1), (cluste
```

# **Problem**

- > Calculate the pagerank of all nodes in network dataset and report the ID and PageRank score of the top 10 pages.
  - Use predefined function in org.apache.spark.graphx.GraphLoader



# Dataset for PageRank

### Citation Network

Node = paper, Edge = citation (directed)

## > About Dataset

- We used CiteSeer network dataset and use edge list only.
- Then, we eliminated some nodes which are not in connected component and made the nodes have unique integer number(ID).

#### CiteSeer for Document Classification

- The CiteSeer dataset consists of 3312 scientific publications classified into one of six classes. The citation network consists of 4732 links. Each publication in the dataset is described by a 0/1-valued word vector indicating the absence/presence of the corresponding word from the dictionary. The dictionary consists of 3703 unique words. The README file in the dataset provides more details.
- Download link:
  - https://linqs-data.soe.ucsc.edu/public/lbc/citeseer.tgz
- · Related papers:
  - Qing Lu, and Lise Getoor. "Link-based classification." ICML, 2003.
  - Prithviraj Sen, et al. "Collective classification in network data." Al Magazine, 2008.

# Dataset for PageRank

## > Dataset representation



\* 'A B' means 'A has been cited by B'



> You can download the pre-processed dataset on iCampus

## **Original dataset**

100157	100157
100157	364207
100157	38848
100157	bradshaw97introduction
100157	bylund99coordinating
100157	dix01 metaagent
100157	gray99finding
100157	labrou01standardizing
100157	labrou99agent
100157	nodine98overview
100157	nodine99active
100157	wagner 97 artificial
100598	455651
100598	marquez00machine
100598	punyakanok01use
((Oiteree en eite ell file	

"Citerseer.cites" file

# Problem

- > Use PageRank in GraphLoader
  - How to run a scala code

1. spark-shell (in terminal)

2. :load ScriptName.scala (in spark-shell)

# **Practice 8**

1. Use predefined classes in org.apache.spark.graphx : GraphLoader ()

2. Find the PageRank of the dataset and report the ID and PageRank score of the top 10 pages.

# Submission

1. You have to submit "result.txt" file on iCampus

Windows

- 2. In your result.txt file, there must be the pairs of node ID and its PageRank.
- 3. Deadline: June 4th 23:59 P.M.
- 4. Your result.txt file must like following

```
      (23.428150459626647,2654)
      (23.428150459626647,2654)

      (18.713274227039285,3204)
      (18.713274227039285,3204)

      (14.857537396076575,2874)
      (14.857537396076575,2874)

      (11.979864744374963,2968)
      (11.979864744374963,2968)
```

Linux

# Solution

```
> Import packages
import org.apache.spark.graphx.GraphLoader
import java.io._
> Load the edges as a graph
val graph = GraphLoader.edgeListFile(sc, "practice8_pagerank.txt")
Run PageRank
  val ranks = graph.pageRank(0.001).vertices
  ranks.foreach(println)
                                  Show the pagerank of each
                                     vertex in the graph
```

# Solution

```
> Swap (id, rank) -> (rank, id)
  val swappedRanks = ranks.map(_.swap)
> Sort
 val sortedRanks = swappedRanks.sortByKey(false)
                                                            Sort vertex by importance
                                                              for showing the most
  Print the result
                                                                important vertex
  val pw = new PrintWriter(new File("result.txt"))
 for(pr_node <- sortedRanks.collect().take(10)){</pre>
                                                            Get top 10 nodes and their
      pw.write(pr_node.toString)
                                                                    pagerank
      pw.println("\n")
  pw.close()
  sc.stop()
```

# Solution

## > Result

- We show only top 4 nodes, but you can see there are 10 nodes which have highest pagerank
- Your result.txt file must be like following

(23.428150459626647,2654)

(18.713274227039285,3204)

(14.857537396076575,2874)

(11.979864744374963,2968)

Windows

(23.428150459626647,2654) (18.713274227039285,3204) (14.857537396076575,2874)

Linux

(11.979864744374963,2968)