```
In [1]: #Import Spark
import os
import sys

spark_home = os.environ['SPARK_HOME'] = '/Users/dunmireg/Documents/
spark-1.6.1-bin-hadoop2.6/'

if not spark_home:
    raise ValueError('Spark Home environment variable not set')

sys.path.insert(0, os.path.join(spark_home, 'python'))
sys.path.insert(0, os.path.join(spark_home, 'python/lib/py4j-0.9-sr
c.zip'))
execfile(os.path.join(spark_home, 'python/pyspark/shell.py'))
```

Welcome to

Using Python version 2.7.10 (default, Oct 23 2015 19:19:21)
SparkContext available as sc, HiveContext available as sqlContext.

HW 13.1: Spark implementation of basic PageRank

Write a basic Spark implementation of the iterative PageRank algorithm that takes sparse adjacency lists as input.

Make sure that your implementation utilizes teleportation (1-damping/the number of nodes in the network), and further, distributes the mass of dangling nodes with each iteration so that the output of each iteration is correctly normalized (sums to 1).

[NOTE: The PageRank algorithm assumes that a random surfer (walker), starting from a random web page, chooses the next page to which it will move by clicking at random, with probability d, one of the hyperlinks in the current page. This probability is represented by a so-called 'damping factor' d, where $d \in (0, 1)$. Otherwise, with probability (1 - d), the surfer jumps to any web page in the network. If a page is a dangling end, meaning it has no outgoing hyperlinks, the random surfer selects an arbitrary web page from a uniform distribution and "teleports" to that page]

In your Spark solution, please use broadcast variables and caching to make sure your code is as efficient as possible.

As you build your code, use the test data

s3://ucb-mids-mls-networks/PageRank-test.txt Or under the Data Subfolder for HW7 on Dropbox with the same file name. (On Dropbox https://www.dropbox.com/sh/2c0k5adwz36lkcw/AAAAKsjQfF9uHfv-X9mCqr9wa?dl=0 (https://www.dropbox.com/sh/2c0k5adwz36lkcw/AAAAKsjQfF9uHfv-X9mCqr9wa?dl=0))

with teleportation parameter set to 0.15 (1-d, where d, the damping factor is set to 0.85), and crosscheck your work with the true result, displayed in the first image in the Wikipedia article:

https://en.wikipedia.org/wiki/PageRank (https://en.wikipedia.org/wiki/PageRank)

and here for reference are the corresponding PageRank probabilities:

A,0.033 B,0.384 C,0.343 D,0.039 E,0.081 F,0.039 G,0.016 H,0.016 I,0.016 J,0.016 K,0.016

Run this experiment locally first. Report the local configuration that you used and how long in minutes and seconds it takes to complete your job.

Repeat this experiment on AWS. Report the AWS cluster configuration that you used and how long in minutes and seconds it takes to complete your job. (in your notebook, cat the cluster config file)

Instructions for AWS

- 1) start a cluster in EMR UI (make sure have spark 1.6.1/or earlier if not available installed)
- 2) make sure you have ssh/pem setup (in the last step in the UI) and your security group allows ssh inbound connection
- 3) once cluster is started, ssh to cluster, ssh hadoop@ec2-52-91-127-197.compute-1.amazonaws.com -i hw13.pem run sudo pip install ipython jupyter
- 4) in the cluster console, run PYSPARK_DRIVER_PYTHON=jupyter
 PYSPARK DRIVER PYTHON OPTS="notebook --no-browser --port=7777" pyspark
- 5) in your local computer, forward the port ssh -i hw13.pem -N -f -L localhost:7776:localhost:7777 hadoop@public-dns
- 6) then open the browser and navigate to localhost: 7776.
- 7) then you are good to go

In [4]:	

```
#All credit to Ron cordell for this implementation
import re
#read line so it can be parallelized into spark
def line splitter(line):
    node, adj_list = re.split('\t',line.strip()) #split
    node = node.strip('"')
    neighbors = eval(adj list) #render as dict
    node list = []
    node list.append((node, neighbors.keys()))
    for neighbor in neighbors:
        node list.append((neighbor, []))
    return node list
#helper function to update the pagerank
def adjustRank(rank, mass):
    adj rank = 0.0
    if rank is not None:
        adj rank = rank
    return d*adj rank + d*mass/n + t #formula
# damping parameter
d = 0.85
D = sc.textFile("PageRank-test.txt") #insert file here to run
graph = D.flatMap(lambda line: line splitter(line)).reduceByKey(lam
bda a,b:a+b).cache()
# compute the number of nodes
n = graph.count()
# compute teleportation factor
t = (1.0-d)/n
# prime the pump with the initial page rank for each node = 1/n
adj list = graph.map(lambda (node, outlinks): (node, (1.0/n, outlin
ks)))
for i in range(0,30):
    dangling mass = adj list.filter(lambda x: len(x[1][1]) == 0).map
(lambda x: x[1][0]).reduce(lambda x,y:x+y)
    distributed mass = adj list.filter(lambda (node, (rank,outlink
s)): len(outlinks) > 0)
        .map(lambda (node, (rank, outlinks)): (rank/len(outlinks),
outlinks))\
        .flatMapValues(lambda x:x)\
        .map(lambda (rank, outlink): (outlink, rank))\
        .reduceByKey(lambda x,y: x+y)
    adj list=graph.leftOuterJoin(distributed mass)\
```

```
.map(lambda (node, (outlinks, rank)):(node, (rank,outlink
s)))\
    .map(lambda (node, (rank, outlinks)):(node, (adjustRank(ran
k, dangling_mass), outlinks)) )

for node in adj_list.sortBy(lambda x: -x[1][0]).collect():
    print node[0], node[1][0]
```

```
B 0.383410412554
C 0.343378600107
E 0.0808856932689
D 0.0390870921233
F 0.0390870921233
A 0.0327814931824
G 0.0161694790207
I 0.0161694790207
K 0.0161694790207
H 0.0161694790207
J 0.0161694790207
```

On AWS I used 6 m3.xlarge nodes which took 43.2308270931 for 30 iterations

HW 13.2: Applying PageRank to the Wikipedia hyperlinks network

Run your Spark PageRank implementation on the Wikipedia dataset for 10 iterations, and display the top 100 ranked nodes (with alpha = 0.85).

Run your PageRank implementation on the Wikipedia dataset for 50 iterations, and display the top 100 ranked nodes (with teleportation factor of 0.15). Plot the pagerank values for the top 100 pages resulting from the 50 iterations run. Then plot the pagerank values for the same 100 pages that resulted from the 10 iterations run. Comment on your findings. Have the top 100 ranked pages changed? Have the pagerank values changed? Explain.

Report the AWS cluster configuration that you used and how long in minutes and seconds it takes to complete your job.

NOTE: ==== English Wikipedia hyperlink network.data ==== The dataset is available via Dropbox at:

https://www.dropbox.com/sh/2c0k5adwz36lkcw/AAAAKsjQfF9uHfv-X9mCqr9wa?dl=0 (https://www.dropbox.com/sh/2c0k5adwz36lkcw/AAAAKsjQfF9uHfv-X9mCqr9wa?dl=0)

on S3 at s3://ucb-mids-mls-networks/wikipedia/ -- s3://ucb-mids-mls-networks/wikipedia/all-pages-indexed-out.txt # Graph -- s3://ucb-mids-mls-networks/wikipedia/indices.txt # Page titles and page Ids

The dataset is built from the Sept. 2015 XML snapshot of English Wikipedia. For this directed network, a link between articles:

$A \rightarrow B$

is defined by the existence of a hyperlink in A pointing to B. This network also exists in the indexed format:

Data: s3://ucb-mids-mls-networks/wikipedia/all-pages-indexed-out.txt Data: s3://ucb-mids-mls-networks/wikipedia/all-pages-indexed-in.txt Data: s3://ucb-mids-mls-networks/wikipedia/indices.txt

but has an index with more detailed data:

(article name) \t (index) \t (in degree) \t (out degree)

In the dictionary, target nodes are keys, link weights are values. Here, a weight indicates the number of time a page links to another. However, for the sake of this assignment, treat this an unweighted network, and set all weights to 1 upon data input.

10 r3.xlarge

Wikipedia 10 took: 1182.26167488 seconds

Wikipedia 50 took: Process took: 4909.77044916 seconds

Wiki 10: 13455888 0.00143798119916 1184351 0.000658251945105 4695850 0.000629974450417 5051368 0.000565939050801 1384888 0.00044507004801 6113490 0.000439619212504 2437837 0.000439441328896 7902219 0.00043737364256 13425865 0.000425377992977 6076759 0.000421589333108 4196067 0.000416677324754 6172466 0.000392502737321 14112583 0.000378696768898 10390714 0.000357441833747 15164193 0.000339380924455 3191491 0.000333818776096 6416278 0.000324284641442 6237129 0.000323688786565 7835160 0.000322179443011 1516699 0.0003198989965 13725487 0.000308716462502 9276255 0.00030565697608 7576704 0.000304530902473 10469541 0.000299306030787 5154210 0.00029382115752 12836211 0.000280065925591 7990491 0.000278579812351 4198751 0.00026445377167 2797855 0.000259527582694 11253108 0.000256916522151 9386580 0.000252902351918 3603527 0.000251586860595 12074312 0.000247074225265 3069099 0.000245401340598 14881689 0.000242623575376 2155467 0.000241229337963 1441065 0.000235164413051 14503460 0.000229901982328 2396749 0.000217083966457 3191268 0.000212449231613 10566120 0.000212042270391 11147327 0.000208264621275 2614581 0.00020821301672 1637982 0.00020463498912 11245362 0.000200221093362 12430985 0.000200183731881 9355455 0.00019374766018 10527224 0.000189360730011 14112408 0.000187349470231 2614578 0.000185369494635 9391762 0.000184862463639 6172167 0.000184579657789 8697871 0.000184216787849 981395 0.000182309018089 6171937 0.000176715345303 5490435 0.000176367442048 11582765 0.000170392829913 14725161 0.000167468441412 9562547 0.000164712124816 12067030 0.000164601303706 994890 0.000163200340599 9394907 0.000158007914422 9997298 0.000157883033536 13280859 0.000156536146392 10345830 0.000155516376098 4978429 0.000152983458437 12447593 0.000152377403763 8019937 0.000150460071734 11148415 0.000147187978269 13432150 0.000145361511179 4344962 0.000144515680409 1175360 0.000140122326465 12038331 0.000139143487726 14565507 0.000136935090125 4624519 0.00013570017848 1523975 0.000134052440417 14981725 0.000133283278987 13328060 0.00013252405361 1332806 0.000128724206236 10399499 0.000128403691278 14963657 0.000127906141328 2826544 0.000126494655262 2578813 0.000125795531863 1813634 0.000124978709254 1575979 0.000124846716592 2778099 0.00012179799776 13853369 0.000118476377302 9924814 0.00011845264948 4568647 0.000113986387573 9742161 0.000112399194205 12785678 0.00011224883786 7467127 0.000112149478495 3328327 0.000111834932093 14727077 0.000111541221372 10246542 0.000111460357921 3591832 0.000111396766917 5274313 0.000111331922407 14709489 0.000110798226712 3973000 0.000110663936005 15070394 0.000110524921697

Wiki 50: 13455888 0.00146154857879 1184351 0.000666013855615 4695850 0.000639672606533 5051368 0.000574762875979 1384888 0.000450120670195 2437837 0.000446666570933 6113490 0.000444629709428 7902219 0.000443875381845 13425865 0.000433138481704 6076759 0.000427704719316 4196067 0.000423413581476 6172466 0.000397823296841 14112583 0.000385482971376 10390714 0.000362663786257 15164193 0.000343585296234 3191491 0.000338047422597 6416278 0.00032921787358 6237129 0.000328992181466 7835160 0.000326199737819 1516699 0.000325108339111 13725487 0.000312680149329 9276255 0.000309567354633 7576704 0.000307978908698 10469541 0.000303118343378 5154210 0.00029754579874 12836211 0.000286034825153 7990491 0.000283617796922 4198751 0.000269051323154 2797855 0.000264011972564 11253108 0.00026098273596 9386580 0.000257695338763 3603527 0.000254969161082 12074312 0.000251020166063 3069099 0.000248673948023 14881689 0.000245362759899 2155467 0.00024471811602 1441065 0.0002386465625 14503460 0.000233302359678 2396749 0.000220630523434 3191268 0.000214954172738 10566120 0.000214543272604 2614581 0.000211201668378 11147327 0.000211185630694 1637982 0.000207030418227 12430985 0.000203300592195 11245362 0.000202528755713 9355455 0.000197012613097 10527224 0.000191389653574 14112408 0.000190781940738 9391762 0.000188169955943 2614578 0.000188020711274 8697871 0.000187042464627 6172167 0.000186731465231 981395 0.000185227371084 6171937 0.000178748154784 5490435 0.000178311959398 11582765 0.000173347300826 14725161 0.00016948266271 12067030 0.000167650594693 9562547 0.000167213540803 994890 0.000165398876541 9997298 0.000160694946602 9394907 0.000160522254514 13280859 0.000159005418943 10345830 0.000157616963123 4978429 0.000155270639931 12447593 0.000154928957148 8019937 0.000153288852635 11148415 0.000148833514713 13432150 0.000147855764335 4344962 0.000147109546466 1175360 0.000141843191136 12038331 0.000141298246852 14565507 0.000139065677208 4624519 0.000137645583292 1523975 0.000136245282789 14981725 0.000134895063036 13328060 0.00013474183704 1332806 0.000130692301864 10399499 0.00013020541465 14963657 0.000130036735847 2578813 0.000128410980262 2826544 0.000128203747864 1575979 0.000127322338903 1813634 0.000127152522854 2778099 0.000124107584482 13853369 0.000120935161169 9924814 0.000120241485718 4568647 0.000115778325695 12785678 0.000114506611887 7467127 0.000114472347202 9742161 0.000114300899384 3328327 0.000113592887172 10246542 0.000113264541758 3591832 0.000113234971214 5274313 0.000113192001912 14727077 0.000112910401522 14709489 0.000112415650593 5908108 0.000112186213975 3973000 0.000112119387579

```
In [3]: | IDs10 = []
        values10 = []
        IDs50 = []
        values50 = []
        with open('Wiki10.txt', 'r') as myfile:
            lines = myfile.readlines()
            for line in lines:
                line = line.split()
                IDs10.append(line[0])
                values10.append(float(line[1]))
        with open('Wiki50.txt', 'r') as myfile:
            lines = myfile.readlines()
            for line in lines:
                line = line.split()
                IDs50.append(line[0])
                values50.append(float(line[1]))
        print "10 iterations" + '\t' + '\t' + '\t' + '\t' + '\t' + ' 50 ite
        rations'
        for i in range(100):
            print "ID: " + str(IDs10[i]) + " value: " + str(values10[i]) +
        '\t' + '\t' + "ID: " + str(IDs50[i]) + " value: " + str(values50
        [i])
```

Final Submission		
10 iterations		iterations
ID: 13455888 value: 0.00143798119916	ID:	13455888 valu
e: 0.00146154857879	TD.	1104251 ****
ID: 1184351 value: 0.000658251945105 0.000666013855615	TD:	1184351 value:
ID: 4695850 value: 0.000629974450417	TD.	4695850 value:
0.000639672606533	TD:	4093630 Value:
ID: 5051368 value: 0.000565939050801	TD•	5051368 value:
0.000574762875979	10.	3031300 Value.
ID: 1384888 value: 0.00044507004801	TD:	1384888 value:
0.000450120670195		
ID: 6113490 value: 0.000439619212504	ID:	2437837 value:
0.000446666570933		
ID: 2437837 value: 0.000439441328896	ID:	6113490 value:
0.000444629709428		
ID: 7902219 value: 0.00043737364256	ID:	7902219 value:
0.000443875381845		
ID: 13425865 value: 0.000425377992977	ID:	13425865 valu
e: 0.000433138481704		
ID: 6076759 value: 0.000421589333108	ID:	6076759 value:
0.000427704719316	TD.	4106067 ****
ID: 4196067 value: 0.000416677324754 0.000423413581476	TD:	4196067 value:
ID: 6172466 value: 0.000392502737321	тп•	6172466 value:
0.000397823296841	10.	0172400 Value.
ID: 14112583 value: 0.000378696768898	ID:	14112583 valu
e: 0.000385482971376		
ID: 10390714 value: 0.000357441833747	ID:	10390714 valu
e: 0.000362663786257		
ID: 15164193 value: 0.000339380924455	ID:	15164193 valu
e: 0.000343585296234		
ID: 3191491 value: 0.000333818776096	ID:	3191491 value:
0.000338047422597		6416080
ID: 6416278 value: 0.000324284641442 0.00032921787358	ID:	6416278 value:
ID: 6237129 value: 0.000323688786565	TD.	6237129 value:
0.000328992181466	TD:	023/129 Value:
ID: 7835160 value: 0.000322179443011	TD:	7835160 value:
0.000326199737819	10.	7000100 (4140)
ID: 1516699 value: 0.0003198989965	ID:	1516699 value:
0.000325108339111		
ID: 13725487 value: 0.000308716462502	ID:	13725487 valu
e: 0.000312680149329		
ID: 9276255 value: 0.00030565697608	ID:	9276255 value:
0.000309567354633		
ID: 7576704 value: 0.000304530902473	ID:	7576704 value:
0.000307978908698		10460541
ID: 10469541 value: 0.000299306030787	ID:	10469541 valu
e: 0.000303118343378 ID: 5154210 value: 0.00029382115752	TD-	5154210 value:
0.00029754579874	זעז:	JIJ42IU Value:
ID: 12836211 value: 0.000280065925591	ID:	12836211 valu
e: 0.000286034825153		
ID: 7990491 value: 0.000278579812351	ID:	7990491 value:

Final Submission		
0.000283617796922 ID: 4198751 value: 0.00026445377167	TD:	4198751 value:
0.000269051323154		1190701 (4140)
ID: 2797855 value: 0.000259527582694	ID:	2797855 value:
0.000264011972564		
ID: 11253108 value: 0.000256916522151	ID:	11253108 valu
e: 0.00026098273596		0206500 1
ID: 9386580 value: 0.000252902351918 0.000257695338763	ID:	9386580 value:
ID: 3603527 value: 0.000251586860595	TD•	3603527 value:
0.000254969161082	10.	3003327 Value.
ID: 12074312 value: 0.000247074225265	ID:	12074312 valu
e: 0.000251020166063		
ID: 3069099 value: 0.000245401340598	ID:	3069099 value:
0.000248673948023		
ID: 14881689 value: 0.000242623575376	ID:	14881689 valu
e: 0.000245362759899 ID: 2155467 value: 0.000241229337963	TD.	2155467 value:
0.00024471811602	TD:	215546/ Value:
ID: 1441065 value: 0.000235164413051	ID:	1441065 value:
0.0002386465625		
ID: 14503460 value: 0.000229901982328	ID:	14503460 valu
e: 0.000233302359678		
ID: 2396749 value: 0.000217083966457	ID:	2396749 value:
0.000220630523434 ID: 3191268 value: 0.000212449231613	TD.	3191268 value:
0.000214954172738	TD:	3191200 value:
ID: 10566120 value: 0.000212042270391	ID:	10566120 valu
e: 0.000214543272604		
ID: 11147327 value: 0.000208264621275	ID:	2614581 value:
0.000211201668378		
ID: 2614581 value: 0.00020821301672	ID:	11147327 valu
e: 0.000211185630694 ID: 1637982 value: 0.00020463498912	TD.	1637982 value:
0.000207030418227	TD:	103/902 Value:
ID: 11245362 value: 0.000200221093362	ID:	12430985 valu
e: 0.000203300592195		
ID: 12430985 value: 0.000200183731881	ID:	11245362 valu
e: 0.000202528755713		
ID: 9355455 value: 0.00019374766018	ID:	9355455 value:
0.000197012613097 ID: 10527224 value: 0.000189360730011	TD.	10527224 valu
e: 0.000191389653574	TD:	1032/224 Valu
ID: 14112408 value: 0.000187349470231	ID:	14112408 valu
e: 0.000190781940738		
ID: 2614578 value: 0.000185369494635	ID:	9391762 value:
0.000188169955943		
ID: 9391762 value: 0.000184862463639	ID:	2614578 value:
0.000188020711274 ID: 6172167 value: 0.000184579657789	TD-	8697871 value:
0.000187042464627	TD:	oogrofi value:
ID: 8697871 value: 0.000184216787849	ID:	6172167 value:
0.000186731465231		
ID: 981395 value: 0.000182309018089	ID:	981395 value:

i mai Suomission		
0.000185227371084 ID: 6171937 value: 0.000176715345303	ID:	6171937 value:
0.000178748154784		
ID: 5490435 value: 0.000176367442048 0.000178311959398	ID:	5490435 value:
ID: 11582765 value: 0.000170392829913	ID:	11582765 valu
e: 0.000173347300826		
ID: 14725161 value: 0.000167468441412 e: 0.00016948266271	ID:	14725161 valu
ID: 9562547 value: 0.000164712124816	ID:	12067030 valu
e: 0.000167650594693		
ID: 12067030 value: 0.000164601303706	ID:	9562547 value:
0.000167213540803 ID: 994890 value: 0.000163200340599	TD:	994890 value:
0.000165398876541	10.	JJ40J0 Value.
ID: 9394907 value: 0.000158007914422	ID:	9997298 value:
0.000160694946602		
ID: 9997298 value: 0.000157883033536	ID:	9394907 value:
0.000160522254514 ID: 13280859 value: 0.000156536146392	TD.	13280859 valu
e: 0.000159005418943	10:	13260639 Valu
ID: 10345830 value: 0.000155516376098	ID:	10345830 valu
e: 0.000157616963123		
ID: 4978429 value: 0.000152983458437	ID:	4978429 value:
0.000155270639931 ID: 12447593 value: 0.000152377403763	TD.	12447593 valu
e: 0.000154928957148	10:	1244/393 Valu
ID: 8019937 value: 0.000150460071734	ID:	8019937 value:
0.000153288852635		
ID: 11148415 value: 0.000147187978269	ID:	11148415 valu
e: 0.000148833514713 ID: 13432150 value: 0.000145361511179	TD:	13432150 valu
e: 0.000147855764335	10.	13432130 Valu
ID: 4344962 value: 0.000144515680409	ID:	4344962 value:
0.000147109546466		1155060
ID: 1175360 value: 0.000140122326465 0.000141843191136	ID:	1175360 value:
ID: 12038331 value: 0.000139143487726	ID:	12038331 valu
e: 0.000141298246852		
ID: 14565507 value: 0.000136935090125	ID:	14565507 valu
e: 0.000139065677208 ID: 4624519 value: 0.00013570017848	TD.	4624519 value:
0.000137645583292	10.	4024317 Value.
ID: 1523975 value: 0.000134052440417	ID:	1523975 value:
0.000136245282789		
ID: 14981725 value: 0.000133283278987	ID:	14981725 valu
e: 0.000134895063036 ID: 13328060 value: 0.00013252405361	TD:	13328060 valu
e: 0.00013474183704	10.	10020000 1414
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0.000112119387579	

The majority of the top IDs haven't really changed but a few of the bottom ones have swapped places by several ranks. Also note the pagerank values of the 50 iterations tend to be larger than the 10.

NB: for an alternative implementation of HW 1-2 please see these links:

https://github.com/dunmireg/HW13/blob/master/HW13 HD localnotebook.ipynb (https://github.com/dunmireg/HW13/blob/master/HW13 HD localnotebook.ipynb)

https://github.com/dunmireg/HW13/blob/master/HW13 HD clusternotebook.ipynb (https://github.com/dunmireg/HW13/blob/master/HW13 HD clusternotebook.ipynb)

HW 13.3: Spark GraphX versus your implementation of PageRank

Run the Spark GraphX PageRank implementation on the Wikipedia dataset for 10 iterations, and display the top 100 ranked nodes (with alpha = 0.85).

Run your PageRank implementation on the Wikipedia dataset for 50 iterations, and display the top 100 ranked nodes (with teleportation factor of 0.15). Have the top 100 ranked pages changed? Comment on your findings. Plot both 100 curves.

Report the AWS cluster configuration that you used and how long in minutes and seconds it takes to complete this job.

Put the runtime results of HW13.2 and HW13.3 in a tabular format (with rows corresponding to implemention and columns corresponding to experiment setup (10 iterations, 50 iterations)). Discuss the run times and explaing the differences.

Plot the pagerank values for the top 100 pages resulting from the 50 iterations run (using GraphX). Then plot the pagerank values for the same 100 pages that resulted from the 50 iterations run of your homegrown pagerank implemnentation. Comment on your findings. Have the top 100 ranked pages changed? Have the pagerank values changed? Explain.

Using a 6 node r3.xlarge cluster 10 iterations took: 644 seconds

and 50 iterations took 1337 seconds

For a second alternative implementation of Homework 1 and 2, and our implementation of Question 3 please see:

https://www.zeppelinhub.com/viewer/notebooks/aHR0cHM6Ly9yYXcuZ2l0aHVidXNlcmNvbnRlbnQuY29tL (https://www.zeppelinhub.com/viewer/notebooks/aHR0cHM6Ly9yYXcuZ2l0aHVidXNlcmNvbnRlbnQuY29t

https://www.zeppelinhub.com/viewer/notebooks/aHR0cHM6Ly9yYXcuZ2l0aHVidXNlcmNvbnRlbnQuY29tL(https://www.zeppelinhub.com/viewer/notebooks/aHR0cHM6Ly9yYXcuZ2l0aHVidXNlcmNvbnRlbnQuY29t

Implementation	10 Iterations	50 Iterations	Cluster
Personal PageRank	1182.26 seconds	4909.77 seconds	10 r3.xlarge
GraphX	644 seconds	1337 seconds	6 r3.xlarge

The graphX implementaion is significantly faster than our homegrown version. This includes the fact that the cluster was almost twice as big for the personal pagerank algorithm. We suspect this is due to optimization.

The GraphX implementation was almost twice as fast at 10 iterations but nearly 4 times faster for the 50 iterations. This is a significant speed boost.

```
In [15]: | IDsWiki = []
         valuesWiki = []
         IDsGraph = []
         valuesGraph = []
         with open('Wiki50.txt', 'r') as myfile:
             lines = myfile.readlines()
             for line in lines:
                 line = line.split()
                 IDsWiki.append(line[0])
                 valuesWiki.append(float(line[1]))
         with open('GraphX50.txt', 'r') as myfile:
             lines = myfile.readlines()
             for line in lines:
                 line = line.strip().split(',')
                 IDsGraph.append(line[0])
                 valuesGraph.append(float(line[1]))
         # for i in range(100):
               if IDsWiki[i] != IDsGraph[i]:
                   print "problem at " + str(i)
         print "Homegrown PageRank" + '\t' + '\t' + '\t' + '\t' + ' GraphX'
         for i in range(100):
             print "ID: " + str(IDsWiki[i]) + " value: " + str(valuesWiki
         [i]) + '\t' + '\t' + "ID: " + str(IDsGraph[i]) + " value: " + str(v
         aluesGraph[i])
```

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Homegrown PageRank		Gra	phX
ID: 13455888 value: 0. e: 6247.50602457	00146154857879	ID:	13455888 valu
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2846.92399367	00620672606522	TD.	4605050 1
ID: 4695850 value: 0.0 2734.33065298	00639672606533	TD:	4695850 value:
ID: 5051368 value: 0.0	00574762875979	ID:	5051368 value:
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e: 1336.57491714 ID: 9276255 value: 0.0	00309567354633	ID:	9276255 value:
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Final Submission		
1212.34762351 ID: 4198751 value: 0.000269051323154	TD•	4198751 value:
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e: 740.987055347	TD:	11582765 valu
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543.523842162			
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479.262711842			

In examining this we find the lds are exactly the same for both the graphX and homegrown pagerank. The values are obviously different but their ranking seems to be the same. So no, the ranked pages have not changed. We believe the GraphX values are not normalized, explaining the difference in their values whereas the homegrown spark version is normalized so all the total mass adds up to 1

Questions 4 and 5

See the following link: https://github.com/dunmireg/HW13/blob/master/MIDS-LSML-HW13.ipynb (https://github.com/dunmireg/HW13/blob/master/MIDS-LSML-HW13.ipynb)

In []:
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