CS-432/532 Introduction to Web Science: Assignment #4: The "Friendship Paradox"

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Problem 1

Determine if the friendship paradox holds for my Facebook account.* Compute the mean, standard deviation, and median of the number of friends that my friends have. Create a graph of the number of friends (y-axis) and the friends themselves, sorted by number of friends (x-axis). (The friends don't need to be labeled on the x-axis: just f1, f2, f3, ... fn.) Do include me in the graph and label me accordingly.

* = This used to be more interesting when you could more easily download your friend's friends data from Facebook. Facebook now requires each friend to approve this operation, effectively making it impossible.

I will email to the list the XML file that contains my Facebook friendship graph ca. Oct, 2013. The interesting part of the file looks like this (for 1 friend):

It is in GraphML format: http://graphml.graphdrawing.org/

1.1 Approach

EXTRACTING DATA

Dr. Nelson's facebook account data was utilized to solve this problem. The data was in a XML structure in the file < mln.graphml >; a Python XML parser was employed to parse its content. The object was stored in variable root (line 21) and all **number-of-friends** of Dr. Nelson's friends were stored in a dictionary (lines 25-30). Finally, this object was sorted and written into a file: $< ffriendship_paradox.dat >$ (lines 35-42).

Dr. Nelson's total **number-of-friends** in the structure was 165. However, there were 10 nodes within the structure that did not have any **friends_count** (10). They were not considered in the data, so the total number of friends in this sample is 155.

Listing 1: GetFriends.py

```
tree = Tree.parse('mln.graphml')
root = tree.getroot()
ns = {'structure': 'http://graphml.graphdrawing.org/xmlns'}
my_friends = root[len(root) - 1].findall('structure:node', ns)
friend = 0
friend_dict = {0: len(my_friends)}
for child in my_friends:
    friend += 1
for element in child.findall('structure:data', ns):
```

```
if element.attrib['key'] == 'friend_count':
              friend_dict[friend] = int(element.text)
31
  print('Total Friends: %d, Missing Data: %d' % (len(my_friends), len(my_friends) -
     len(friend_dict)))
  print(sorted(friend_dict.items(), key=operator.itemgetter(1)))
33
  friend = 0
  with open('ffriendship_paradox.dat', 'w') as file:
35
      file.write('Friend\tNo.Friends\n')
      for friend_tuple in sorted(friend_dict.items(), key=operator.itemgetter(1)):
37
          friend += 1
38
          if friend_tuple[0] == 0:
              file.write('mln\t%d\n' % friend_tuple[1])
4(
41
          else:
              file.write('f%d\t%d\n' % (friend, friend_tuple[1]))
```

1.2 Solution

Figure 1: Dr. Nelson's Facebook Friends Graph

3200 3000 2500 Number of Friends 2000 1500 1000 500 mean 358 median f1 f14 f24 f34 mln f54 f64 f74 f84 f94 f104 f124 f144 Friends

Friendship Paradox for Dr. Nelson

The data was sorted by **Number-of-Friends** field (see Table 1), the graph shows characteristics of a power-law distribution. The **Number-of-Friends** corresponding to mln is 165 and mln is position at the left side of the median 259. Then, the friendship paradox holds true for mln Facebook account.

The mean, standard deviation, and median was computed using R. The file < facebook.R > is the script that makes the computation and graph plotting. The file is attached to this document. Calculation for the number of friends that mln friends has:

• Mean: 358.2

• σ : 259

• Median: 370.7

 σ has a high value. It is skew to the magnitude of the last 5 friends. See Table 1. However, even if we were to disregard this data, mln's Number of Friends are so far left from the mean and the median that the friendship paradox holds true for mln Facebook account, in which his friends have more friends than him.

Problem 2

Determine if the friendship paradox holds for your Twitter account. Since Twitter is a directed graph, use "followers" as value you measure (i.e., "do your followers have more followers than you?").

Generate the same graph as in question #1, and calculate the same mean, standard deviation, and median values.

For the Twitter 1.1 API to help gather this data, see:

```
https://dev.twitter.com/docs/api/1.1/get/followers/list
```

If you do not have followers on Twitter (or don't have more than 50), then use my twitter account "phone-dude_mln".

2.1 Approach

EXTRACTING DATA

My sister Twitter account was employed to test if the paradox holds. A REST Twitter API [1]. A cURL command was used to execute the request:

```
curl --get 'https://api.twitter.com/1.1/followers/list.json' --data
    'amp%3Binclude_user_entities=false&amp%3Bscreen_name=
    roselenavargas&amp%3Bskip_status=true&cursor=-1'
    --header 'Authorization: OAuth oauth_consumer_key="L9IMyConSuMerKeyT1MC",
    oauth_nonce="33b5ab398f8145ac453017896ebe99a1",
    oauth_signature="IQgpTL2PBvHzeu%2BKa7veW7EIxrw%3D", oauth_signature_method="HMAC-SHA1",
    oauth_timestamp="1456395318", oauth_token="412664119-
    YJ5IeQTheAuThoRizaTionToken", oauth_version="1.0"' --verbose > twitter
```

The result was redirected to a file < twitter >. Since the result is a JSON object, the original approach was to utilize a Python library, such as pickel or json to create a structure to iterate and extract its data. Both approaches came empty, in which the number of followers extracted was less than in the data structure. This most likely was due to some type of encoding problem in the file. However, to move forward a third approach proved to be effective: the use of regex.

< GetFollowers.py > is the Python program that helped extract the information needed to make the proof. The heart of the code is in line 25:

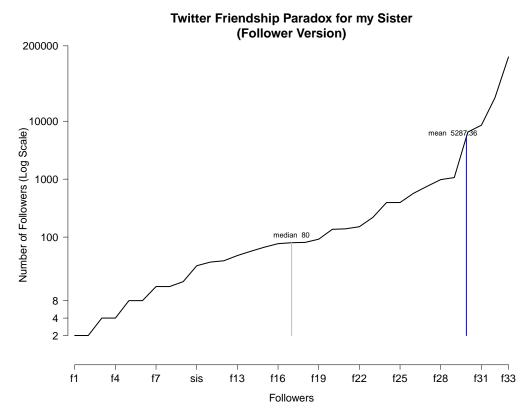
Listing 2: GetFollowers.py

```
followers_dict = {}
for followers in re.findall('\"followers_count\":\d+', x):
    counter += 1
followers_dict[counter] = int(re.findall('\d+', followers)[0])
print(counter, followers, re.findall('\d+', followers)[0])
followers_dict[0] = counter
```

An iterable object was formed with the regular expression '\"followers_count\\":\d+', extracting followers_count amounts. Finally, my sister's information along her followers was placed in file $< twitter_p aradox.dat >$. Table 2 shows the data set result.

2.2 Solution

Figure 2: Sister's Twitter Followers Graph



The data was sorted by **Number-of-Friends**, the graph shows characteristics of a power-law distribution. The **Number-of-Friends** corresponding to sis is 32 (see Table 2). Since my sister on the x-axis is positioned at the left side of the median 259 and the mean 5287, then the friendship paradox holds for my sister Twitter account, in which her friends have more followers than her.

The mean, standard deviation, and median was computed using R. The file < twitter.R > is the script that makes the computation and graph plotting. The file is attached to this document. Calculation for the number of friends that my sister's friends have:

• Mean: 5,287.35

• σ : 22,623

• Median: 80

Problem 3 Extra Credit

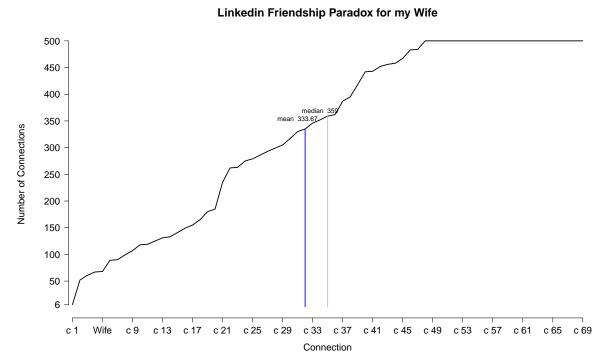
Repeat question #1, but with your LinkedIn profile.

3.1 Approach

All the connections from my wife's LinkedIn account were downloaded in file < source.txt >. I couldn't cross reference the internal connection ID with their public ID, so a manual computation on the number of connection was employed to get the data.

3.1 Solution

Figure 3: Wife Linkedin Connection Graph



The data was sorted by **Connection Count**. The **Connection Count** corresponding to *Wife* is **68** and *my wife* is positioned at the left side of the median **359**. Then, the friendship paradox holds for my wife Linkedin account in which her connections have more connections than her.

Problem 3 Extra Credit

Repeat question #2, but change "followers" to following? In other words, are the people I am following following more people?

4.1 Approach

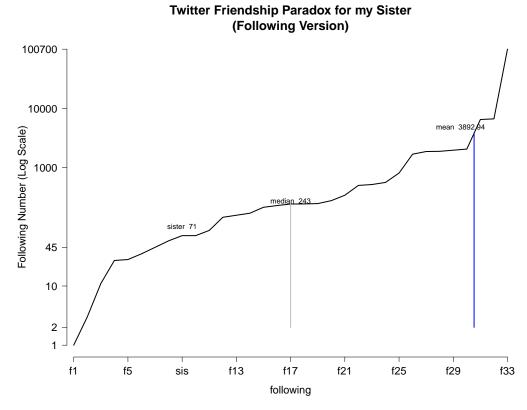
The same approach as in 2.1 was followed. The only difference was that instead of using **follower_count** as in the *regex* filter we used **friends_count** which points to the number of particular person may be following. See listing below:

Listing 3: GetFollowing.py

```
following_dict = {counter: 71}
for following in re.findall('\"friends_count\":\d+', x):
    counter += 1
following_dict[counter] = int(re.findall('\d+', following)[0])
print(counter, following, re.findall('\d+', following)[0])
```

4.1 Solution

Figure 4: Sister's Twitter Following Friends Graph



Plotting the data sorted by **Number-Following**, the graph shows characteristics of a power-law distribution. The **Number-Following** corresponding to sis is 71 and my sister is position at the left side of the median 243. Then, the friendship paradox holds for my sister Twitter account in which her friends are following more people than her.

The mean, standard deviation, and median was computed using R. The file < twitter 2.R > is the script that makes the computation and graph plotting. The file is attached to this document. Calculation for the number of friends that my sister's friends have:

• Mean: 3,892.94

• σ : 17,449

• Median: 243

Table 1: Facebook Number of Friends that Dr. Nelson's friends have

Friend	No. Friends	Friend	No. Friends	Friend	No. Friends	Friend	No. Friends
f1	7	f48	170	f95	328	f142	763
f2	15	f49	172	f96	337	f143	770
f3	25	f50	181	f97	348	f144	784
f4	30	f51	182	f98	351	f145	819
f5	38	f52	183	f99	353	f146	833
f6	39	f53	186	f100	359	f147	844
f7	40	f54	187	f101	363	f148	873
f8	41	f55	190	f102	366	f149	909
f9	41	f56	195	f103	380	f150	1194
f10	42	f57	197	f104	387	f151	1346
f11	43	f58	204	f105	400	f152	1512
f12	54	f59	207	f106	404	f153	1521
f13	58	f60	208	f107	409	f154	1626
f14	59	f61	220	f108	415	f155	3187
f15	60	f62	227	f109	420	1100	9101
f16	62	f63	229	f110	421		
f17	65	f64	231	f1110	424		
f18	68	f65	231	f1112	425		
f19	68	f66	233	f1112	425		
f20	77	f67	233	f114	427		
f21	80	f68	235	f115	436		
f22	85	f69	236	f116	443		
f23	86	f70	240	f117	448		
f24	87	f71	240	f1117	449		
f25	89	f72	241	f1118	449		
f26	93	f73	242	f120	496		
f27	94	f74	244	f120	510		
f28	94	f75	245	f121	524		
f29	96	f76	250	f123	528		
f30	97	f77	255 255	f123	538		
f31	104	f78	259 259	f124	539		
f32	104	f79	274	f126	552		
f33	104	f80	275	f120	555 555		
f34	108	f81	276				
f35		f82	278	f128 f129	555 561		
f36	111		295				
	123	f83		f130 f131	562		
f37 f38	124	f84 f85	297 297		568		
	128	1		f132	575 576		
f39 f40	131 143	f86 f87	308 312	f133 f134	576 580		
f41	144	f88	315	f135	592		
f42	147	f89	317	f136	615		
f43	155	f90	321	f137	619		
mln	165	f91	321	f138	622		
f45	165	f92	322	f139	624		
f46	168	f93	324	f140	705 750		
f47	168	f94	327	f141	752		

Data was sorted by Number of Friends that mln has. This data was use to plot Figure 1

Table 2: Twitter Followers Data

f1 2 f2 2 f3 4 f4 4 f5 8 f6 8 f6 8 f7 14 f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361 f33 128602	Friend	Followers
f3 4 f4 4 f5 8 f6 8 f7 14 f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f1	2
f4 4 f5 8 f6 8 f7 14 f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f2	2
f5 8 f6 8 f7 14 f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f3	4
f6 8 f7 14 f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f4	4
f7 14 f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f5	8
f8 14 f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f6	8
f9 17 sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f7	14
sister 32 f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f8	14
f11 37 f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f9	17
f12 39 f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	sister	32
f13 48 f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f11	37
f14 57 f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f12	39
f15 66 f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f13	48
f16 77 f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f14	57
f17 80 f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f15	66
f18 81 f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f16	77
f19 92 f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f17	80
f20 136 f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f18	81
f21 139 f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f19	92
f22 151 f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f20	136
f23 220 f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f21	139
f24 394 f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f22	151
f25 395 f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f23	220
f26 571 f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f24	394
f27 753 f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f25	395
f28 981 f29 1066 f30 6504 f31 8527 f32 25361	f26	571
f29 1066 f30 6504 f31 8527 f32 25361	f27	753
f30 6504 f31 8527 f32 25361	f28	981
f31 8527 f32 25361	f29	1066
f32 25361	f30	6504
	f31	8527
f33 128602		25361
155 125002	f33	128602

Data for Figure 2. $\,$

Table 3: My Sister's Linkedin Connection Data

Connection	Count	Connection	Count
C1	6	C48	500
C2	52	C49	500
C3	61	C50	500
C4	67	C51	500
Wife	68	C52	500
C6	89	C53	500
C7	90	C54	500
C8	99	C55	500
C9	107	C56	500
C10	118	C57	500
C11	119	C58	500
C12	125	C59	500
C13	131	C60	500
C14	133	C61	500
C15	141	C62	500
C16	149	C63	500
C17	155	C64	500
C18	165	C65	500
C19	180	C66	500
C20	185	C67	500
C21	235	C68	500
C22	262	C69	500
C23	263		
C24	275		
C25	279		
C26	286		
C27	293		
C28	299		
C29	305		
C30	317		
C31	330		
C32	335		
C33	346		
C34	352		
C35	359		
C36	362		
C37	387		
C38	395		
C39	418		
C40	442		
C41	443		
C42	452		
C43	456		
C44	458		
C45	467		
C46	483		
C47	484		

Data for Figure 3. It was sorted by Connection Count.

Table 4: Twitter Following Friends Data

Friend	Following
f1	1
f2	3
f3	11
f4	27
f5	28
f6	35
f7	45
f8	58
sis	71
f10	71
f11	87
f12	145
f13	157
f14	170
f15	214
f16	229
f17	243
f18	244
f19	247
f20	277
f21	342
f22	501
f23	519
f24	563
f25	811
f26	1691
f27	1875
f28	1889
f29	1974
f30	2061
f31	6514
f32	6687
f33	100677

Data for Figure 4. It was sorted by number of friends a friend is following.

References

[1] Twitter API. (n.d.) Retrieved February 24, 2016, from https://dev.twitter.com/docs/api/1.1/get/followers/list