Analiza društvene mreže Reddit

Univerzitet u Beogradu, Elektrotehnički fakultet, veb adresa ovde.

Predmet: Analiza socijalnih mreža, veb adresa ovde.

Tekst projektnog zadatka se može naći na veb sajtu predmeta ovde.

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Instalacija i učitavanje korišćenih biblioteka

Neophodne bilioteke se mogu instalirati korišćenjem Python package instalera pip .

```
In [1]: import os
   import matplotlib.pyplot as plt
   import networkx as nx
   import numpy as np
   import pandas as pd
```

Format skupa podataka

Podaci su preuzeti sa sajta predmeta i otpakovani u odgovarajuće direktorijume. Kako postoji više datoteka sa istim tipom podataka, učitavamo ih sve zajedno i spajamo u jedinstvenu strukturu podataka.

```
In [2]: reddit_data_path = "ASM_PZ2_podaci_2122/reddit2008"
    submission_dataPath = f"{reddit_data_path}/submissions_2008_asm/"
    comments_dataPath = f"{reddit_data_path}/comments_2008_asm_v1.1/comments_2008/"

    def loadDataSet(folderPath):
        allFileData = pd.DataFrame([])
        for fileName in os.listdir(folderPath):
            singleFileData = pd.read_csv(folderPath + fileName, low_memory=False)
            allFileData = pd.concat([allFileData, singleFileData])

    return allFileData

def groupby_count(data_frame, groupby_list):
    return data_frame.groupby(groupby_list).size().reset_index(name="counts")

def groupby_count_sorted(data_frame, groupby_list):
    return groupby_count(data_frame, groupby_list).sort_values('counts', ascending='counts')
```

Filtriranje skupa podataka

Na društvenoj mreži Reddit je moguće obrisati nalog - u tom slučaju se mogu dobiti podaci koji sadrže [deleted] kao korisničko ime. Ovakvi podaci nepovoljno utiču na analizu jer su ti nalozi mogli pripadati proizvoljnom broju korisnika i stoga ih filtriramo i ne koristimo u daljoj analizi.

Statistička obrada podataka

```
allSubredditIds = np.union1d(submissionData['subreddit_id'], commentsData['subreddi
In [4]:
        print(f"Number of different subreddits: {len(allSubredditIds)}")
        commentsPerSubreddit = groupby_count_sorted(commentsData, ["subreddit_id"])
        print(f"Comments per subreddit:\n{commentsPerSubreddit[:1]}")
        # subreddit - author - count interactions
        interactionsPerAuthorPerSubreddit = groupby count(allData, ["subreddit id", "author
        # subreddit - count authors
        authorsPerSubreddit = groupby_count_sorted(interactionsPerAuthorPerSubreddit, ["sub
        print(f"Authors per subreddit:\n{authorsPerSubreddit[:1]}")
        print(f"AVG number users per subreddit:\n{authorsPerSubreddit['counts'].sum() / len
        submissionsPerAuthor = groupby_count_sorted(filteredSubmissions, ['author'])
        commentsPerAuthor = groupby count sorted(filteredComments, ['author'])
        print(f"Max submissions per author:\n{submissionsPerAuthor[:1]}")
        print(f"Max comments per author:\n{commentsPerAuthor[:1]}")
        # author - subreddit - count interactions
        interactionsPerSubredditPerAuthor = groupby_count(allData, ['author', 'subreddit_id
        # author - count subreddits
        subredditsPerAuthor = groupby_count_sorted(interactionsPerSubredditPerAuthor, ['aut
        print(f"Subreddits per author:\n{subredditsPerAuthor[:1]}")
```

```
Number of different subreddits: 5032
Comments per subreddit:
     subreddit_id
                    counts
2689
             t5 6 1884629
Authors per subreddit:
     subreddit id counts
4354
             t5_6 163779
AVG number users per subreddit:
128.78398251192368
Max submissions per author:
      author counts
84823
         gst
               18870
Max comments per author:
                author counts
12603 NoMoreNicksLeft
                         13480
Subreddits per author:
         author counts
26173 MrKlaatu
                    181
```

In [5]: paersonCalculation = submissionsPerAuthor.copy().merge(commentsPerAuthor.copy(), on
 paersonCalculation.plot.scatter(y="counts_submissions", x="counts_comments")
 print(f"Pearson correlation matrix:\n{paersonCalculation.corr(method='pearson')}")

counts_comments

Pearson correlation matrix:

```
counts submissions
                                       1.000000
                                                             0.150412
counts_comments
                                       0.150412
                                                             1.000000
   17500
   15000
counts submissions
   12500
   10000
    7500
    5000
    2500
       0
                           4000
                                   6000
                                            8000
                   2000
                                                   10000
                                                           12000
                                                                    14000
            0
                                counts_comments
```

counts_submissions

In [6]: filterNonOver18 = submissionData["over_18"] == False
 filteredSubmissionsNonOver18 = submissionData[filterNonOver18]
 extractedCommentsData = pd.DataFrame(commentsData["link_id"].map(lambda element: el
 commentsDataSubmissionId = groupby_count_sorted(extractedCommentsData, ["link_id"])
 filteredSubmissionsNonOver18JoinedComments = filteredSubmissionsNonOver18.merge(com
 topSubmissionsNonOver18JoinedComments = filteredSubmissionsNonOver18JoinedComments.
 print(topSubmissionsNonOver18JoinedComments)
 topSubmissionsNonOver18JoinedComments.to_csv("top_submission_non_over_18.csv")

	Unnamed: 0	id								url	\	
290534	191390	7kpe5	https://www.reddit.com/r/AskF				/AskRedd	ddit/comments/7k				
574617	111467	6nz1k	http://hundredpushups.com									
57505	135995	675oj	https://www.reddit.com/r/reddit.com/comments/6									
942221	191265	78n1v	http://www.time.com/time/nation/article/0,8599									
133696	36220	7beo2	https://www.reddit.com/r/politics/comments/7be									
488720	73206	6jbc0	https://www.reddit.com/r/reddit.com/comments/6									
764729	197408	6yn6t	http://www.cnbc.com/id/26454655?									
721214	84267	6w7e6	https://www.reddit.com/r/reddit.com/comments/6									
136018	41921	7bj3f	http://www.latimes.com/news/local/la-me-gaymar									
772368	4466	6z2e2	http://www.nytimes.com/reuters/us/internationa									
200524	/ /4 5 11.1	,	. /71	- /·			malink		uthor	\		
290534	/r/AskReddit/comments/7kpe5/i_am_constantly_wo								[deleted]			
574617	/r/science/comments/6nz1k/got_six_weeks_try_th zekel											
57505	/r/reddit.com/comments/675oj/post_the_funniest matiasklein											
942221	/r/WTF/comments/78n1v/a_black_community_in_oh indorock											
133696	/r/politics/comments/7beo2/obama_wins_the_pres willjohnston											
488720 764729	/r/reddit.com/comments/6jbc0/voice_a_true_yet [deleted]											
721214	/r/politics/comments/6yn6t/its_official_sarah [deleted]											
136018	<pre>/r/reddit.com/comments/6w7e6/just_out_of_curio [deleted] /r/reddit.com/comments/7bj3f/who_else_disguste [deleted]</pre>											
772368	/r/reddit.com/comments/6z2e2/palin_says_her_da nucleophile											
772300	/1/1 edd1c.co	iii/ Colline	1103/02	.262/ paii	11_34	y 3_11C1	ua	Hucieo	philite			
	created_utc	subre	eddit s	ubreddit	id	num c	omments	score	over	18 \	\	
290534	1229752197	AskRe		t5_2q	_	_	3467	515		- lse		
574617	1213826517	sci	lence	t5_m	ouw		33329	1621	Fa:	lse		
57505	1201730171	reddit	.com	t	5_6		2039	1098	Fa	lse		
942221	1224677051		WTF t5_2qh6		h61		3657	1496	Fai	lse		
133696	1225857637	poli	itics t5_2cne				1934	8538	Fa:	lse		
488720	1210747619	reddit	.com	t	5_6		1808	274	Fa:	lse		
764729	1220020890	•	itics	t5_2c	neq		1668	1073	Fa:	lse		
721214	1218660724	reddit	.com		5_6		1724	258		lse		
136018	1225899136	reddit			5_6		1605	5992		lse		
772368	1220285161	reddit	.com	t	5_6		1425	1517	Fa	lse		
		ı										
200524	distinguished		- 1 C A -				locked	_		count		
290534	NaN			kReddit		False	False		alse	306		
574617	NaN			nups.com		False	False		alse	289		
57505	NaN			ldit.com		False	False		alse	207		
942221	NaN			ime.com		False	False		alse	204		
133696	NaN			olitics		False	False		alse	200		
488720	NaN			ldit.com		False	False		alse	193		
764729	NaN			nbc.com		False	False		alse	183		
721214	NaN			ldit.com		False	False		alse	182		
136018	NaN			mes.com		False	False		alse	173		
772368	NaN	l	nyti	mes.com		False	False	F	alse	155	80	

Modelovanje podataka grafovima

Skup podataka modelujemo pomoću 4 različita grafa:

- 1. SNet (Subreddit network) sadrži kompletne podatke, sve sabredite i interakcije sa njima
- 2. SNetF (Subreddit network filtered) filtrirani SNet na osnovu broja korisnika koji definišu interakciju izmedju više sabredita
- 3. SNetT (Subreddit network targeted) filtrirani SNet na osnovu odabranih sabredita i grana kojima su povezani
- 4. UserNet sadrži interakcije između korisnika komentare na objave ili na komentare

Iz SNet eliminišemo sve čvorove koji nemaju nijednu granu kako bismo omogućili dalju analizu povezanosti čvorova grafa.

```
In [7]:
        snet = nx.Graph()
        snet.add_nodes_from(allSubredditIds)
        authorSubredditIdGroups = allData.groupby(["author", "subreddit_id"]).groups
        groups = dict()
        for author, subredditId in authorSubredditIdGroups:
            if author not in groups:
                groups[author] = [subredditId]
                groups[author].append(subredditId)
        for key in groups:
            subreddit ids = groups[key]
            for i in range(0, len(subreddit_ids)):
                for j in range(i + 1, len(subreddit_ids)):
                    if snet.has_edge(subreddit_ids[i], subreddit_ids[j]):
                         snet.edges[subreddit ids[i], subreddit ids[j]]['weight'] += 1
                    else:
                         snet.add_edge(subreddit_ids[i], subreddit_ids[j], weight=1)
        snet.remove_nodes_from(list(nx.isolates(snet)))
```

Odabir praga za filtriranje grana po težini

Prag za filtriranje je uzet kao prosečna vrednost svih težina grana u grafu SNet.

```
In [8]: average_weight = sum([tags["weight"] for u, v, tags in snet.edges(data=True)]) / le
w_threshold = average_weight

# TODO: check if commented out works
# snetf = snet.edge_subgraph([(u, v) for u, v, tags in snet.edges(data=True) if tag

snetf = nx.Graph()
snetf.add_nodes_from(allSubredditIds)
snetf.add_edges_from([(u, v, tags) for u, v, tags in snet.edges(data=True) if tags[
snetf.remove_nodes_from(list(nx.isolates(snetf)))
```

Odabir sabredita od interesa

Sabrediti od interesa za analizu pomoću SNetT su uzeti iz teksta projektog zadatka.

```
In [9]:
        targetSubreddits = {
             "reddit.com",
             "pics",
             "worldnews",
             "programming",
             "business",
             "politics",
             "obama",
             "science",
             "technology",
             "WTF",
             "AskReddit",
             "netsec",
             "philosophy",
             "videos",
             "offbeat",
             "funny",
             "entertainment",
             "linux",
             "geek",
             "gaming",
             "comics",
             "gadgets",
             "nsfw",
             "news",
             "environment",
             "atheism",
             "canada",
             "math",
             "Economics",
             "scifi",
             "bestof",
             "cogsci",
             "joel",
             "Health",
             "guns",
             "photography",
             "software",
             "history",
             "ideas",
         }
         targetSubredditIds = [allData[allData["subreddit"] == targetSubreddit]["subreddit_i
         snett = snet.subgraph(targetSubredditIds)
```

UserNet

Modelujemo interakcije korisnika društvene mreže usmerenim grafom.

Dobijene grafove čuvamo u standardnom gml formatu na disku kako bismo im mogli pristupati i iz eksternih alata.

Generisanje Erdos-Renyi mreža

Erdos-Renyi mreže koristimo za poređenje sa mrežama reddit podataka.

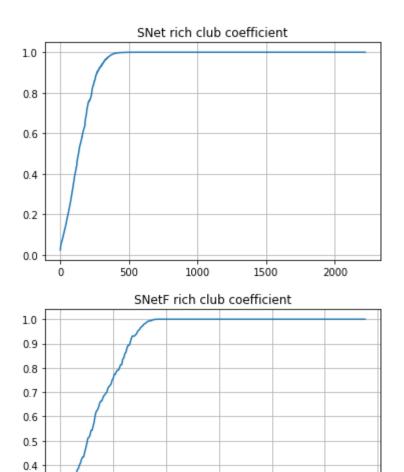
Verovatnoća za stvaranje grana u grafu je odabrana kao gustina grafova koje ispitujemo.

```
In [102... # erdos_renyi_snet = nx.erdos_renyi_graph(n=snet.number_of_nodes(), p=nx.density(sn
# erdos_renyi_snetf = nx.erdos_renyi_graph(n=snetf.number_of_nodes(), p=nx.density(
# erdos_renyi_usernet = nx.erdos_renyi_graph(n=usernet.number_of_nodes(), p=nx.dens
In [103... # nx.write_gml(erdos_renyi_snet, "erdos_renyi_snet.gml")
# nx.write_gml(erdos_renyi_snetf, "erdos_renyi_snetf.gml")
# nx.write_gml(erdos_renyi_usernet, "erdos_renyi_usernet.gml")
```

Analiza modelovanih grafova

Klub bogatih

Kako bi izvrsavanje funkcija trajalo kratko, koristi se nenormalizovana varijanta. Kao sto se moze videti sa grafika dole, grafovi ispoljavaju klub bogatih (cvorovi su u klubu bogatih za vrednosti 1).



Asortativna analiza

50

100

150

0.3

Kako su vrednosti koeficijenata asortativnosti manje od 0, ove mreže su disortitativna - čvorovi ne teže da se povežu sa sličnim čvorovima.

200

250

300

```
In [34]: print(f"Degree assortativity (SNet) {nx.degree_assortativity_coefficient(snet)}")
    print(f"Degree assortativity (SNetF) {nx.degree_assortativity_coefficient(snetf)}")
    snet_node_degrees = {key: value for key, value in snet.degree}
    snet_average_neighbor_degree = nx.average_neighbor_degree(snet)

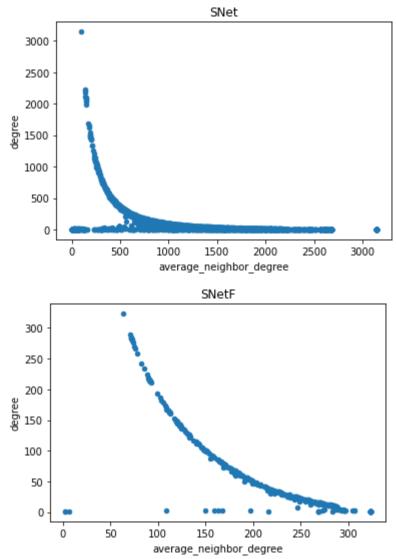
snet_degree_average_neighbor_degree = pd.DataFrame.from_records([snet_node_degrees, snet_degree_average_neighbor_degree.plot.scatter(x="average_neighbor_degree", y="de

snetf_node_degrees = {key: value for key, value in snetf.degree}
    snetf_average_neighbor_degree = nx.average_neighbor_degree(snetf)

snetf_degree_average_neighbor_degree = pd.DataFrame.from_records([snetf_node_degree snetf_degree_average_neighbor_degree.plot.scatter(x="average_neighbor_degree", y="d

Degree assortativity (SNet) -0.436456805812687
Degree assortativity (SNetF) -0.6177660601425986
```

Out[34]: <AxesSubplot:title={'center':'SNetF'}, xlabel='average_neighbor_degree', ylabel='de
 gree'>



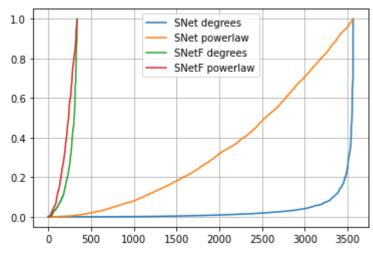
Stepen čvora i power law

Mreže ne prate power law raspodelu.

```
In [12]: snet_sorted_degrees = np.array(sorted([degree for node_id, degree in snet.degree]))
    snet_sorted_degrees_normalized = snet_sorted_degrees / snet_sorted_degrees[-1]
    snet_x_axis = np.linspace(0, len(snet_sorted_degrees), len(snet_sorted_degrees))
    snet_sorted_power_law = sorted(np.random.power(0.5, len(snet_sorted_degrees)))
    plt.plot(snet_x_axis, snet_sorted_degrees_normalized, label="SNet degrees")
    plt.plot(snet_x_axis, snet_sorted_power_law, label="SNet powerlaw")

snetf_sorted_degrees = np.array(sorted([degree for node_id, degree in snetf.degree]
    snetf_sorted_degrees_normalized = snetf_sorted_degrees / snetf_sorted_degrees[-1]
    snetf_x_axis = np.linspace(0, len(snetf_sorted_degrees), len(snetf_sorted_degrees))
    snetf_sorted_power_law = sorted(np.random.power(0.5, len(snetf_sorted_degrees)))
    plt.plot(snetf_x_axis, snetf_sorted_degrees_normalized, label="SNetF degrees")
    plt.plot(snetf_x_axis, snetf_sorted_power_law, label="SNetF powerlaw")

plt.legend()
    plt.grid()
```



Katz centrality

Isti čvorovi su centralni i po ovoj metrici, nezavisno od menjanja parametra beta.

```
t5_6 = reddit.com
t5_2qh16 = technology
t5_2cneq = politics
```

```
In [13]:
         def calculate katz centrality(G, beta scalar):
             beta dict = dict()
             for node_id in G.nodes:
                  beta_dict[node_id] = 1.0 if node_id != "t5_6" else beta_scalar
             return nx.katz centrality(G, alpha=0.0005, beta=beta dict, max iter=1000)
         count_of_most_important = 3
         print("SNet Katz central nodes:")
         snet_katz_dict_1 = calculate_katz_centrality(snet, 1.0)
         print(sorted(snet_katz_dict_1, key=lambda node_id: -snet_katz_dict_1[node_id])[:cou
         snet_katz_dict_5 = calculate_katz_centrality(snet, 50.0)
         print(sorted(snet_katz_dict_5, key=lambda node_id: -snet_katz_dict_5[node_id])[:cou
         snet katz dict 10 = calculate katz centrality(snet, 100.0)
         print(sorted(snet_katz_dict_10, key=lambda node_id: -snet_katz_dict_10[node_id])[:c
         print("SNetF Katz central nodes:")
         snetf_katz_dict_1 = calculate_katz_centrality(snetf, 1.0)
         print(sorted(snetf_katz_dict_1, key=lambda node_id: -snetf_katz_dict_1[node id])[:d
         snetf_katz_dict_5 = calculate_katz_centrality(snetf, 50.0)
         print(sorted(snetf_katz_dict_5, key=lambda node_id: -snetf_katz_dict_5[node_id])[:c
         snetf_katz_dict_10 = calculate_katz_centrality(snetf, 100.0)
         print(sorted(snetf_katz_dict_10, key=lambda node_id: -snetf_katz_dict_10[node_id])[
         SNet Katz central nodes:
         ['t5_6', 't5_2qh16', 't5_2cneq']
         ['t5_6', 't5_2qh16', 't5_2cneq']
         ['t5_6', 't5_2qh16', 't5_2cneq']
         SNetF Katz central nodes:
         ['t5_6', 't5_2cneq', 't5_2qh16']
         ['t5_6', 't5_2cneq', 't5_2qh16']
         ['t5_6', 't5_2cneq', 't5_2qh16']
```