main

April 8, 2023

1 Detecting Communities in the Reddit Hyperlink Dataset

1.1 For ECMM447 - Social Networks and Text Analysis

```
[34]: # Only run once to install necessary dependencies
      !pip install scikit-image
      !pip install python-louvain
      !pip install cdlib
     Requirement already satisfied: scikit-image in c:\users\aorun\desktop\uni\year
     4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (0.20.0)
     Requirement already satisfied: packaging>=20.0 in
     c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
     (23.0)
     Requirement already satisfied: scipy<1.9.2,>=1.8 in
     c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
     (1.9.1)
     Requirement already satisfied: imageio>=2.4.1 in c:\users\aorun\desktop\uni\year
     4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
     (2.27.0)
     Requirement already satisfied: pillow>=9.0.1 in c:\users\aorun\desktop\uni\year
     4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
     (9.5.0)
     Requirement already satisfied: tifffile>=2019.7.26 in
     c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
     (2023.3.21)
     Requirement already satisfied: PyWavelets>=1.1.1 in
     c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
     analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
     (1.4.1)
     Requirement already satisfied: numpy>=1.21.1 in c:\users\aorun\desktop\uni\year
```

4\term 2\ecmm447 - social networks and text

```
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
(1.23.5)
Requirement already satisfied: lazy_loader>=0.1 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
(0.2)
Requirement already satisfied: networkx>=2.8 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from scikit-image)
(3.0)
Requirement already satisfied: python-louvain in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (0.16)
Requirement already satisfied: numpy in c:\users\aorun\desktop\uni\year 4\term
2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from python-
louvain) (1.23.5)
Requirement already satisfied: networkx in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from python-
louvain) (3.0)
Collecting cdlib
 Downloading cdlib-0.2.6-py3-none-any.whl (228 kB)
     ------ 0.0/228.6 kB ? eta -:--:--
    ----- 194.6/228.6 kB 5.9 MB/s eta 0:00:01
    ----- 228.6/228.6 kB 7.0 MB/s eta 0:00:00
Collecting future
 Downloading future-0.18.3.tar.gz (840 kB)
    ----- 0.0/840.9 kB ? eta -:--:--
    ----- 225.3/840.9 kB 4.6 MB/s eta 0:00:01
    ----- 409.6/840.9 kB 5.1 MB/s eta 0:00:01
    ----- 491.5/840.9 kB 4.4 MB/s eta 0:00:01
    ----- 727.0/840.9 kB 5.1 MB/s eta 0:00:01
    ----- 840.9/840.9 kB 4.8 MB/s eta 0:00:00
 Preparing metadata (setup.py): started
 Preparing metadata (setup.py): finished with status 'done'
Requirement already satisfied: matplotlib in c:\users\aorun\desktop\uni\year
4\text{-} 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from cdlib)
(3.7.1)
Collecting python-Levenshtein
 Downloading python Levenshtein-0.20.9-py3-none-any.whl (9.4 kB)
Requirement already satisfied: scipy in c:\users\aorun\desktop\uni\year 4\term
2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from cdlib)
(1.9.1)
Collecting bimlpa
 Downloading bimlpa-0.1.2-py3-none-any.whl (7.0 kB)
```

```
Collecting eva-1cd
 Downloading eva_lcd-0.1.1-py3-none-any.whl (9.2 kB)
Collecting chinese-whispers
 Downloading chinese_whispers-0.8.1-py3-none-any.whl (7.8 kB)
Collecting pyclustering
 Downloading pyclustering-0.10.1.2.tar.gz (2.6 MB)
   ----- 0.0/2.6 MB ? eta -:--:--
   --- 0.2/2.6 MB 6.1 MB/s eta 0:00:01
   ----- 0.4/2.6 MB 5.5 MB/s eta 0:00:01
   ----- 0.6/2.6 MB 5.4 MB/s eta 0:00:01
   ----- 0.8/2.6 MB 5.5 MB/s eta 0:00:01
   ----- 0.9/2.6 MB 4.8 MB/s eta 0:00:01
   ----- 1.1/2.6 MB 4.9 MB/s eta 0:00:01
   ----- 1.3/2.6 MB 5.3 MB/s eta 0:00:01
   ----- 1.5/2.6 MB 5.4 MB/s eta 0:00:01
   ----- 1.6/2.6 MB 5.2 MB/s eta 0:00:01
   ----- 1.7/2.6 MB 4.7 MB/s eta 0:00:01
   ----- 1.9/2.6 MB 4.9 MB/s eta 0:00:01
     ----- 2.1/2.6 MB 4.9 MB/s eta 0:00:01
   ----- 2.3/2.6 MB 4.9 MB/s eta 0:00:01
   ----- 2.5/2.6 MB 5.0 MB/s eta 0:00:01
   ----- 2.6/2.6 MB 5.0 MB/s eta 0:00:00
 Preparing metadata (setup.py): started
 Preparing metadata (setup.py): finished with status 'done'
Collecting dynetx
 Downloading dynetx-0.3.1-py3-none-any.whl (39 kB)
Collecting pooch
 Downloading pooch-1.7.0-py3-none-any.whl (60 kB)
   ----- 0.0/60.9 kB ? eta -:--:-
   ----- 60.9/60.9 kB ? eta 0:00:00
Collecting cython
 Downloading Cython-0.29.34-py2.py3-none-any.whl (988 kB)
   ----- 0.0/988.1 kB ? eta -:--:-
   ----- 266.2/988.1 kB 8.3 MB/s eta 0:00:01
   ----- 481.3/988.1 kB 7.5 MB/s eta 0:00:01
   ----- 675.8/988.1 kB 7.1 MB/s eta 0:00:01
   ----- -- 911.4/988.1 kB 6.4 MB/s eta 0:00:01
   ----- 988.1/988.1 kB 6.9 MB/s eta 0:00:00
Collecting pulp
 Using cached PuLP-2.7.0-py3-none-any.whl (14.3 MB)
Collecting thresholdclustering
 Downloading thresholdclustering-1.1-py3-none-any.whl (5.3 kB)
Requirement already satisfied: networkx>=2.4 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from cdlib) (3.0)
Collecting tqdm
 Downloading tqdm-4.65.0-py3-none-any.whl (77 kB)
   ----- 0.0/77.1 kB ? eta -:--:-
```

```
----- 77.1/77.1 kB 4.2 MB/s eta 0:00:00
Collecting demon
 Downloading demon-2.0.6-py3-none-any.whl (7.3 kB)
Collecting angel-cd
 Downloading angel_cd-1.0.3-py3-none-any.whl (10 kB)
Collecting seaborn
 Using cached seaborn-0.12.2-py3-none-any.whl (293 kB)
Collecting python-igraph
 Downloading python-igraph-0.10.4.tar.gz (9.5 kB)
 Preparing metadata (setup.py): started
 Preparing metadata (setup.py): finished with status 'done'
Collecting nf1
 Downloading nf1-0.0.4-py3-none-any.whl (18 kB)
Requirement already satisfied: pandas in c:\users\aorun\desktop\uni\year 4\term
2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from cdlib)
(1.5.3)
Requirement already satisfied: numpy in c:\users\aorun\desktop\uni\year 4\term
2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from cdlib)
Requirement already satisfied: python-louvain>=0.16 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from cdlib) (0.16)
Collecting scikit-learn
 Downloading scikit_learn-1.2.2-cp39-cp39-win_amd64.whl (8.4 MB)
   ----- 0.0/8.4 MB ? eta -:--:-
   - ----- 0.2/8.4 MB 7.6 MB/s eta 0:00:02
   - ----- 0.3/8.4 MB 7.0 MB/s eta 0:00:02
   -- ----- 0.5/8.4 MB 4.7 MB/s eta 0:00:02
   -- ----- 0.6/8.4 MB 4.6 MB/s eta 0:00:02
   -- ----- 0.6/8.4 MB 3.5 MB/s eta 0:00:03
   --- 0.8/8.4 MB 3.9 MB/s eta 0:00:02
   ---- 1.1/8.4 MB 4.2 MB/s eta 0:00:02
   ---- 1.1/8.4 MB 4.3 MB/s eta 0:00:02
   ---- 1.2/8.4 MB 4.0 MB/s eta 0:00:02
   ----- 1.4/8.4 MB 4.2 MB/s eta 0:00:02
   ----- 1.7/8.4 MB 4.3 MB/s eta 0:00:02
   ----- 1.9/8.4 MB 4.6 MB/s eta 0:00:02
   ----- 2.2/8.4 MB 4.8 MB/s eta 0:00:02
   ----- 2.3/8.4 MB 4.7 MB/s eta 0:00:02
   ----- 2.5/8.4 MB 4.7 MB/s eta 0:00:02
   ----- 2.7/8.4 MB 4.8 MB/s eta 0:00:02
   ----- 3.0/8.4 MB 5.0 MB/s eta 0:00:02
   ----- 3.2/8.4 MB 5.1 MB/s eta 0:00:02
   ----- 3.3/8.4 MB 5.0 MB/s eta 0:00:02
   ----- 3.5/8.4 MB 4.9 MB/s eta 0:00:01
   ----- 3.7/8.4 MB 5.0 MB/s eta 0:00:01
```

```
----- 3.8/8.4 MB 5.0 MB/s eta 0:00:01
   ----- 4.0/8.4 MB 4.9 MB/s eta 0:00:01
   ----- 4.1/8.4 MB 4.8 MB/s eta 0:00:01
   ----- 4.1/8.4 MB 4.8 MB/s eta 0:00:01
   ----- 4.2/8.4 MB 4.6 MB/s eta 0:00:01
   ----- 4.4/8.4 MB 4.6 MB/s eta 0:00:01
   ----- 4.5/8.4 MB 4.6 MB/s eta 0:00:01
   ----- 4.6/8.4 MB 4.6 MB/s eta 0:00:01
   ----- 4.9/8.4 MB 4.6 MB/s eta 0:00:01
   ----- 5.1/8.4 MB 4.7 MB/s eta 0:00:01
   ----- 5.3/8.4 MB 4.7 MB/s eta 0:00:01
   ----- 5.6/8.4 MB 4.8 MB/s eta 0:00:01
   ----- 5.8/8.4 MB 4.8 MB/s eta 0:00:01
   ----- 5.9/8.4 MB 4.8 MB/s eta 0:00:01
   ----- 6.2/8.4 MB 4.8 MB/s eta 0:00:01
   ----- 6.5/8.4 MB 4.9 MB/s eta 0:00:01
   ----- 6.7/8.4 MB 5.0 MB/s eta 0:00:01
   ----- 6.8/8.4 MB 5.0 MB/s eta 0:00:01
   ----- 6.9/8.4 MB 4.9 MB/s eta 0:00:01
   ----- 7.2/8.4 MB 4.9 MB/s eta 0:00:01
   ----- 7.4/8.4 MB 5.0 MB/s eta 0:00:01
   ----- 7.7/8.4 MB 5.0 MB/s eta 0:00:01
   ----- -- 7.9/8.4 MB 5.0 MB/s eta 0:00:01
   ----- - 8.1/8.4 MB 5.0 MB/s eta 0:00:01
   ----- 8.3/8.4 MB 5.1 MB/s eta 0:00:01
   ----- 8.4/8.4 MB 5.0 MB/s eta 0:00:00
Collecting markov-clustering
 Downloading markov_clustering-0.0.6.dev0-py3-none-any.whl (6.3 kB)
Collecting networkx>=2.4
 Using cached networkx-2.8.8-py3-none-any.whl (2.0 MB)
Requirement already satisfied: decorator in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
dynetx->cdlib) (5.1.1)
Requirement already satisfied: kiwisolver>=1.0.1 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (1.4.4)
Requirement already satisfied: pyparsing>=2.3.1 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (3.0.9)
```

Requirement already satisfied: packaging>=20.0 in c:\users\aorun\desktop\uni\vear 4\term 2\ecmm447

matplotlib->cdlib) (4.39.3)

Requirement already satisfied: fonttools>=4.22.0 in

c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text

c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text

analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from

```
matplotlib->cdlib) (23.0)
Requirement already satisfied: python-dateutil>=2.7 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (2.8.2)
Requirement already satisfied: contourpy>=1.0.1 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (1.0.7)
Requirement already satisfied: importlib-resources>=3.2.0 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (5.12.0)
Requirement already satisfied: cycler>=0.10 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (0.11.0)
Requirement already satisfied: pillow>=6.2.0 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
matplotlib->cdlib) (9.5.0)
Requirement already satisfied: pytz>=2020.1 in c:\users\aorun\desktop\uni\year
4\text{-} 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
pandas->cdlib) (2023.3)
Requirement already satisfied: platformdirs>=2.5.0 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from pooch->cdlib)
(3.2.0)
Requirement already satisfied: requests>=2.19.0 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from pooch->cdlib)
(2.28.2)
Requirement already satisfied: igraph==0.10.4 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from python-
igraph->cdlib) (0.10.4)
Requirement already satisfied: texttable>=1.6.2 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
igraph==0.10.4->python-igraph->cdlib) (1.6.7)
Collecting Levenshtein==0.20.9
  Downloading Levenshtein-0.20.9-cp39-cp39-win_amd64.whl (101 kB)
    ----- 0.0/101.3 kB ? eta -:--:-
    ----- 61.4/101.3 kB 1.6 MB/s eta 0:00:01
    ----- 101.3/101.3 kB 1.9 MB/s eta 0:00:00
Collecting rapidfuzz<3.0.0,>=2.3.0
```

analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from

```
Downloading rapidfuzz-2.15.0-cp39-cp39-win_amd64.whl (1.1 MB)
    ----- 0.0/1.1 MB ? eta -:--:-
    ----- 0.2/1.1 MB 6.9 MB/s eta 0:00:01
    ----- 0.5/1.1 MB 5.6 MB/s eta 0:00:01
    ----- 0.7/1.1 MB 6.3 MB/s eta 0:00:01
    ----- 0.9/1.1 MB 6.3 MB/s eta 0:00:01
    ----- 1.1/1.1 MB 6.1 MB/s eta 0:00:00
Collecting threadpoolctl>=2.0.0
 Using cached threadpoolctl-3.1.0-py3-none-any.whl (14 kB)
Collecting joblib>=1.1.1
 Using cached joblib-1.2.0-py3-none-any.whl (297 kB)
Requirement already satisfied: colorama in c:\users\aorun\desktop\uni\year
4\text{-} 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from tqdm->cdlib)
Requirement already satisfied: zipp>=3.1.0 in c:\users\aorun\desktop\uni\year
4\text{term }2\text{-}social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from importlib-
resources>=3.2.0->matplotlib->cdlib) (3.15.0)
Requirement already satisfied: six>=1.5 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from python-
dateutil>=2.7->matplotlib->cdlib) (1.16.0)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
requests>=2.19.0->pooch->cdlib) (1.26.15)
Requirement already satisfied: certifi>=2017.4.17 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
requests>=2.19.0->pooch->cdlib) (2022.12.7)
Requirement already satisfied: idna<4,>=2.5 in c:\users\aorun\desktop\uni\year
4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
requests>=2.19.0->pooch->cdlib) (3.4)
Requirement already satisfied: charset-normalizer<4,>=2 in
c:\users\aorun\desktop\uni\year 4\term 2\ecmm447 - social networks and text
analysis\ca\redditcommunitydetection\.venv\lib\site-packages (from
requests>=2.19.0->pooch->cdlib) (3.1.0)
Building wheels for collected packages: future, pyclustering, python-igraph
 Building wheel for future (setup.py): started
 Building wheel for future (setup.py): finished with status 'done'
 Created wheel for future: filename=future-0.18.3-py3-none-any.whl size=492055
\verb|sha| 256 = 17 fa 293 dbc 19435c1 ddb 090b392fd 0f4e 90b4671366cb347a88f9af0eb4e0382|
 Stored in directory: c:\users\aorun\appdata\local\pip\cache\wheels\bf\5d\6a\2e
53874f7ec4e2bede522385439531fafec8fafe005b5c3d1b\\
 Building wheel for pyclustering (setup.py): started
 Building wheel for pyclustering (setup.py): finished with status 'done'
```

```
any.whl size=2395137
     \verb|sha| 256 = f939 ca 089513971b0b951471cfc8b05510d18b36a1de0da0dc83745df61b52ed|
       Stored in directory: c:\users\aorun\appdata\local\pip\cache\wheels\e0\56\c2\ab
     b6866a3fcd8a55862f1df8a18f57805c3a78fed9a9023cb9
       Building wheel for python-igraph (setup.py): started
       Building wheel for python-igraph (setup.py): finished with status 'done'
       Created wheel for python-igraph: filename=python_igraph-0.10.4-py3-none-
     any.whl size=9092
     sha256=daeefa2b7e9c9e7512277247089cbfb9103e52bad2362a56d4aca8ea73614267
       Stored in directory: c:\users\aorun\appdata\local\pip\cache\wheels\dc\07\ac\bf
     f79052fd6222d1239b228cd24a47222f227c2350f9c4df01
     Successfully built future pyclustering python-igraph
     Installing collected packages: pulp, tqdm, threadpoolctl, rapidfuzz, networkx,
     joblib, future, cython, thresholdclustering, scikit-learn, python-igraph, pooch,
     Levenshtein, eva-lcd, dynetx, demon, chinese-whispers, seaborn, python-
     Levenshtein, pyclustering, nf1, markov-clustering, bimlpa, angel-cd, cdlib
       Attempting uninstall: networkx
         Found existing installation: networkx 3.0
         Uninstalling networkx-3.0:
           Successfully uninstalled networkx-3.0
     Successfully installed Levenshtein-0.20.9 angel-cd-1.0.3 bimlpa-0.1.2
     cdlib-0.2.6 chinese-whispers-0.8.1 cython-0.29.34 demon-2.0.6 dynetx-0.3.1 eva-
     lcd-0.1.1 future-0.18.3 joblib-1.2.0 markov-clustering-0.0.6.dev0 networkx-2.8.8
     nf1-0.0.4 pooch-1.7.0 pulp-2.7.0 pyclustering-0.10.1.2 python-Levenshtein-0.20.9
     python-igraph-0.10.4 rapidfuzz-2.15.0 scikit-learn-1.2.2 seaborn-0.12.2
     threadpoolctl-3.1.0 thresholdclustering-1.1 tqdm-4.65.0
[47]: import numpy as np
      from collections import Counter
      import time
      # Pandas Imports
      import pandas
      import pandas as pd
      # Matplotlib Imports
      import matplotlib.pyplot as plt
      import matplotlib.colors as mcolors
      %matplotlib inline
      # NetworkX Imports
      import networkx as nx
      from networkx.algorithms import community as comm
      # Louvain Community Detection
      import community
```

Created wheel for pyclustering: filename=pyclustering-0.10.1.2-py3-none-

```
# Iteration Tools for Girvan-Newman
from itertools import islice, takewhile
# Imports for Downloading TSV
import urllib.request as req
import os.path as path
# CDLib Imports for Community Detection
from cdlib import NodeClustering
from cdlib.evaluation import triangle participation ratio, conductance,
 →newman_girvan_modularity
# Datashader Imports
import colorcet as cc
import datashader as ds
import datashader.transfer functions as tf
from datashader.layout import circular_layout, forceatlas2_layout
from datashader.bundling import directly_connect_edges, hammer_bundle
from datashader.utils import export_image
```

```
[4]: # Downloads the Reddit Hyperlink dataset TSV file and saves it to ./data/
     url = 'https://snap.stanford.edu/data/soc-redditHyperlinks-body.tsv'
     body_save = "./data/soc-redditHyperlinks-body.tsv"
     print("Checking for 'soc-redditHyperlinks-body.tsv'...")
     if not path.exists(body_save):
         print("File not Found.")
         print(f"Downloading 'soc-redditHyperlinks-body.tsv' from URL: '{url}'...")
         req.urlretrieve(url, body_save)
     else:
         print(f"Found at Path: '{body_save}'!")
     url = 'https://snap.stanford.edu/data/soc-redditHyperlinks-title.tsv'
     title_save = "./data/soc-redditHyperlinks-title.tsv"
     print("Checking for 'soc-redditHyperlinks-title.tsv'...")
     if not path.exists(title_save):
         print("File not Found.")
         print(f"Downloading 'soc-redditHyperlinks-title.tsv' from URL: '{url}'...")
         req.urlretrieve(url, title_save)
        print("Done!")
     else:
         print(f"Found at Path: '{title_save}'!")
     title_path = title_save
     body_path = body_save
     # Load the TSV files and convert to CSV files
```

```
title_df = pd.read_table(title_path, sep="\t")
title_csv = title_path[:-4] + ".csv"
print("Checking for 'soc-redditHyperlinks-title.csv'...")
if not path.exists(title_csv):
    print(f"Converting TSV to CSV file at: '{title_csv}'...")
    title_df.to_csv(title_csv)
    print("Converted!")
else:
    print(f"Found at Path: {title csv}")
body_df = pd.read_table(body_path, sep='\t')
body_csv = body_path[:-4] + ".csv"
print("Checking for 'soc-redditHyperlinks-body.csv'...")
if not path.exists(body_csv):
    print(f"Converting TSV to CSV file at: '{body_csv}'...")
    body_df.to_csv(body_csv)
    print("Converted!")
else:
    print(f"Found at Path: {body_csv}")
# Load CSV files into DataFrames, concatenate them and extract Source/Target_
 \rightarrownodes
body_df = pd.read_csv(body_csv)
title_df = pd.read_csv(title_csv)
print("Creating edgelist DataFrame...")
reddit_df = pd.concat([title_df, body_df]).reset_index(drop=True)
reddit_df = reddit_df[["SOURCE_SUBREDDIT", "TARGET_SUBREDDIT"]]
reddit_df = reddit_df.groupby(["SOURCE SUBREDDIT", "TARGET SUBREDDIT"]).size().
  →reset_index(name='WEIGHT')
print("Done!")
reddit path = "./data/redditHyperlinks-subredditsOnly.csv"
print(f"Saving edge details to CSV file at: '{reddit_path}'...")
reddit_df.to_csv(reddit_path, index=False)
print("Saved!")
Checking for 'soc-redditHyperlinks-body.tsv'...
Found at Path: './data/soc-redditHyperlinks-body.tsv'!
Checking for 'soc-redditHyperlinks-title.tsv'...
Found at Path: './data/soc-redditHyperlinks-title.tsv'!
Checking for 'soc-redditHyperlinks-title.csv'...
Found at Path: ./data/soc-redditHyperlinks-title.csv
Checking for 'soc-redditHyperlinks-body.csv'...
Found at Path: ./data/soc-redditHyperlinks-body.csv
Creating edgelist DataFrame...
```

Done!

```
Saving edge details to CSV file at: './data/redditHyperlinks-subredditsOnly.csv'...
Saved!
```

The following function loads a CSV file into a DataFrame.

```
[5]: def load_df(path):
    df = pd.read_csv(path)
    return df
```

Next, I load the subreddit edgelist into a dataframe. All graphs will be made from reddit_df or a subset of it.

```
[6]: reddit_csv = "./data/redditHyperlinks-subredditsOnly.csv"
reddit_df = load_df(reddit_csv)
print(reddit_df.columns)
```

```
Index(['SOURCE_SUBREDDIT', 'TARGET_SUBREDDIT', 'WEIGHT'], dtype='object')
```

The following function is a helper function that plots the nodes of a network via Datashader.

```
[7]: def nodes_plot(nodes, name=None, canvas=None, cat=None, cmap=["#FF3333"]):
    # Create Datashader Canvas
    canvas = ds.Canvas(**cvsopts) if canvas is None else canvas

# Create node aggregator
    aggregator = None if cat is None else ds.count_cat(cat)
    agg = canvas.points(nodes, 'x', 'y', aggregator)

# Return shaded nodes
    return tf.spread(tf.shade(agg, cmap=cmap), px=3, name=name)
```

The following function plots the edges of a network via Datashader

```
[8]: def edges_plot(edges, name=None, canvas=None):
    # Create Datashader Canvas
    canvas = ds.Canvas(**cvsopts) if canvas is None else canvas

# Return shaded edges
    return tf.shade(canvas.line(edges, 'x', 'y', agg=ds.count()), name=name)
```

The following function plots the nodes and edges combined

```
[9]: def graph_plot(nodes, edges, name="", canvas=None, cat=None):
    # Create the Canvas to draw on
    if canvas is None:
        xr = nodes.x.min(), nodes.x.max()
        yr = nodes.y.min(), nodes.y.max()
        canvas = ds.Canvas(x_range=xr, y_range=yr, **cvsopts)
```

```
# Plot the nodes
np = nodes_plot(nodes, name + " nodes", canvas, cat)

# Plot the edges
ep = edges_plot(edges, name + " edges", canvas)

# Return both nodes and edges shaded over one another
return tf.stack(ep, np, how="over", name=name)
```

Now, perform some formatting to allow the nodes and edges to be displayed by the graph_plot() function.

```
[10]: # Create a list of unique subreddits
      unique = pd.concat([reddit_df["SOURCE_SUBREDDIT"],__
       →reddit_df["TARGET_SUBREDDIT"]]).unique()
      # Create a list of nodes
      nodes = pd.DataFrame({'name': unique})
      # Create a dictionary of node name to index in nodes
      nodes_dict = {name: idx for idx, name in enumerate(nodes['name'])}
      # Create list of edges by mapping nodes_dict over reddit_df
      edges = reddit_df.copy()
      edges['source'] = edges["SOURCE_SUBREDDIT"].map(nodes_dict)
      edges['target'] = edges["TARGET_SUBREDDIT"].map(nodes_dict)
      edges['weight'] = edges['WEIGHT']
      edges.drop(columns=["SOURCE_SUBREDDIT", "TARGET_SUBREDDIT", "WEIGHT"], __
       →inplace=True)
      print(nodes.head)
      print(edges.head)
```

```
<bound method NDFrame.head of</pre>
                                                    name
0
                      007
                  07scape
1
2
            07thexpansion
3
         098f6bcd4621d373
               0____0
4
67175 radiotransmissions
67176
             islandparkny
67177 pogolithearthsound
67178
                ifukmydog
67179
                 zistopia
[67180 rows x 1 columns]>
```

```
<bound method NDFrame.head of</pre>
                                        source target weight
0
                  55863
                              1
1
             1
                  34580
                               2
2
             2
                  53031
                               1
3
             3
                  3289
                               1
4
                  49828
             4
339638
         55860
                  49449
                               1
339639
         55861
                  6316
                               1
                  17220
339640
         55861
                               1
339641
         55861
                  19663
339642
         55862
                  49953
                               1
```

[339643 rows x 3 columns]>

Next I plot the network in a Circular layout and using the ForceAtlas2 algorithm.

```
[11]: # Run this once to calculate the position of the nodes in both layouts.
      # The ForceAtlas2 Algorithm takes a long time for the entire network.
      # For example, when testing it took 28 minutes on a 32GB machine.
      cd_csv = "./data/circular.csv"
      fd_csv = "./data/force_directed.csv"
      print("Checking for saved Circular Layout CSV...")
      if not path.exists(cd_csv):
          print("File not found, generating layout...")
          %time cd = circular_layout(nodes, uniform=False)
      else:
          print(f"Found file at: {cd csv}")
          cd = pd.read_csv(cd_csv)
      print("Checking for saved Force-Directed Layout CSV...")
      if not path.exists(fd_csv):
          print("File not found, generating layout...")
          %time fd = forceatlas2_layout(nodes, edges)
      else:
          print(f"Found file at: {fd_csv}")
          fd = pd.read_csv(fd_csv)
```

Checking for saved Circular Layout CSV...
Found file at: ./data/circular.csv
Checking for saved Force-Directed Layout CSV...
Found file at: ./data/force_directed.csv

As this operation can take a long time, save the generated layouts to a CSV file.

```
[12]: # Save Circular Layout
if not path.exists(cd_csv):
    print(f"Saving file to: {cd_csv}")
```

```
cd.to_csv(cd_csv)

# Save Force-Directed Layout

if not path.exists(fd_csv):
    print(f"Saving file to: {fd_csv}")
    fd.to_csv(fd_csv)
```

Define some Canvas properties.

Wall time: 4min 32s

```
[13]: cvsopts = dict(plot_height=1200, plot_width=1200)
```

Create the plots. In this case, create a directly-connected graph for both Circular and Force-Directed as well as Bundled Edge graphs.

```
[12]: # Plot directly connected edge versions
      %time cd_n = graph_plot(cd, directly_connect_edges(cd, edges), "Circular_
       →Layout")
      %time fd_n = graph_plot(fd, directly_connect_edges(fd, edges), "Force-Directedu
       →Layout")
      # Also plot bundled edge versions
      %time cd_b = graph_plot(cd, hammer_bundle(cd, edges, weight='weight'), "Bundled"
       ⇔Circular Layout")
      %time fd b = graph plot(fd, hammer bundle(fd, edges, weight='weight'), "Bundled"
       →Force-Directed Layout")
     CPU times: total: 4.7 s
     Wall time: 4.74 s
     CPU times: total: 3.08 s
     Wall time: 3.07 s
     CPU times: total: 27min 55s
     Wall time: 12min 44s
     CPU times: total: 7min 3s
```

```
[13]: # Display images
tf.Images(cd_n, fd_n, cd_b, fd_b).cols(2)
```

[13]: <datashader.transfer_functions.Images at 0x1cf39dfbd00>

As can be seen form the Force-Directed layout, there are many disjoint components which do not connect to the main network of subreddits. These could be considered as very small communities in themselves. There is an evident, large central mass of subreddits, which are all interconnected.

In terms of the Circular layout, there are not any visible or clearly defined communities evident, although it is clear from the thick edges found in the Bundled representation that subreddits tend to connect to one another via key nodes.

As there are many disconnected components, the Louvain algorithm can be sped up by remove these disconnected components before computation so that they do not need to be evaluated.

```
[14]: # Create Weighted Graph
weighted = nx.from_pandas_edgelist(edges, 'source', 'target', ['weight'])
print(f"Number of Nodes: {nx.number_of_nodes(weighted)}")
```

Number of Nodes: 67180

To evaluate the amount of trimming necessary, create a histogram showing the size of all connected components in the dataset.

```
[15]: # Get all connected components
      %time connected_components = list(nx.connected_components(weighted))
      print(f"There are {len(connected_components)} connected components.")
      # Get sizes of connected components
      component_sizes = [len(c) for c in connected_components]
      size_counts = Counter(component_sizes)
      size_labels = size_counts.keys()
      # Plot histogram
      plt.xticks(range(len(size_counts.values())), size_labels)
      plt.xlabel('Size of Connected Components')
      plt.ylabel('Frequency')
      plt.title('Histogram of Connected Component Sizes')
      plt.bar(range(len(size_counts.values())), size_counts.values(),__

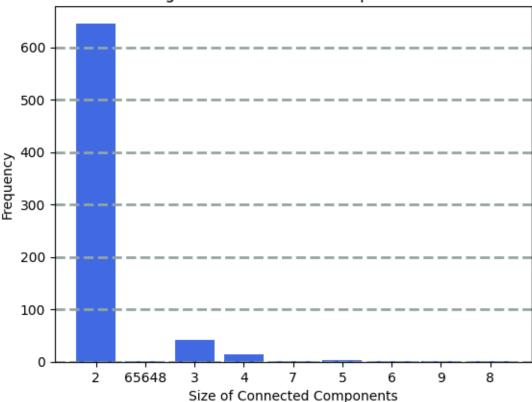
color='royalblue')

      plt.grid(color='#95a5a6', linestyle='--', linewidth=2, axis='y')
      plt.show()
      # Print list of actual numbers
      for key, value in size_counts.items():
          if value == 1:
              print(f"There is {value} component that contains {key} nodes.")
          else:
              print(f"There are {value} components that contain {key} nodes.")
```

CPU times: total: 93.8 ms Wall time: 90.3 ms

There are 712 connected components.





```
There are 646 components that contain 2 nodes. There is 1 component that contains 65648 nodes. There are 42 components that contain 3 nodes. There are 14 components that contain 4 nodes. There are 2 components that contain 7 nodes. There are 3 components that contain 5 nodes. There are 2 components that contain 6 nodes. There is 1 component that contains 9 nodes. There is 1 component that contains 8 nodes.
```

As can be seen from the histogram, there are 646 components of only 2 nodes, which can be immediately discarded. Going forwards, I will use the single connected component containing 65648 nodes, as this contains enough nodes to apply community detection to.

```
[16]: # Get largest connected component
largest = max(connected_components, key=len)
%time largest = weighted.subgraph(largest)
print(f"Number of Nodes: {nx.number_of_nodes(largest)}")
```

CPU times: total: 0 ns Wall time: 6.94 ms

```
Number of Nodes: 65648
```

Now plot the Circular and Force-Directed layouts again for the new, fully-connected network.

```
[17]: # Convert back into edgelist dataframe
      %time connected_edges = nx.to_pandas_edgelist(largest)
      print(f"Number of Edges: {len(connected_edges)}")
      print(connected_edges.columns)
     CPU times: total: 3.83 s
     Wall time: 3.83 s
     Number of Edges: 308839
     Index(['source', 'target', 'weight'], dtype='object')
[18]: # Create a new node list for the large component
      in_largest = nodes.index.isin(set(connected_edges['source']) |__
       set(connected_edges['target']))
      connected_nodes = nodes[in_largest]
      print(connected_nodes.head)
      print(connected_edges.head)
     <bound method NDFrame.head of</pre>
                                                          name
     1
                        07scape
                  07thexpansion
     2
     3
              098f6bcd4621d373
     4
                     0____0
     5
                        Omagick
     67175 radiotransmissions
                   islandparkny
     67176
     67177 pogolithearthsound
                      ifukmydog
     67178
     67179
                       zistopia
     [65648 rows x 1 columns]>
     <bound method NDFrame.head of</pre>
                                            source target weight
                       34580
     1
                   1
                         117
                                   3
     2
              34580
                         117
                                   4
     3
              34580
                       47740
                                  10
     4
                   2
                       53031
                                   1
     308834
              55729
                       67169
                                   1
     308835
              55729
                       67170
                                   1
                                   1
     308836
              55729
                       67171
     308837
              55796
                       55850
                                   1
     308838
              55798
                       55799
                                   1
```

```
[308839 rows x 3 columns]>
```

Calculate the positions for the single, connected component

```
[19]: # Calculate positions for the Circular and Force-Directed Layouts
      # The ForceAtlas2 Algorithm takes a long time for the entire network.
      # For example, when testing it took 31 minutes on a 32GB machine.
      connected_cd_csv = "./data/connected_cd.csv"
      connected_fd_csv = "./data/connected_fd.csv"
      print("Checking for saved Connected Circular layout CSV...")
      if not path.exists(connected_cd_csv):
          print("File not found, generating layout...")
          %time connected_cd = circular_layout(connected_nodes, uniform=False)
      else:
          print(f"File found at: {connected_cd_csv}")
          connected_cd = pd.read_csv(connected_cd_csv)
      print("Checking for saved Connected Force-Directed layout CSV...")
      if not path.exists(connected_fd_csv):
          print("File not found, generating layout...")
          %time connected_fd = forceatlas2_layout(connected_nodes, edges)
      else:
          print(f"File found at: {connected_fd_csv}")
          connected fd = pd.read csv(connected fd csv)
```

Checking for saved Connected Circular layout CSV...
File found at: ./data/connected_cd.csv
Checking for saved Connected Force-Directed layout CSV...
File found at: ./data/connected_fd.csv

Again, the above operation can take a long time, so save the DataFrame to CSV for later use.

```
[20]: # Save Circular Layout
if not path.exists(connected_cd_csv):
    print(f"Saving file to: {connected_cd_csv}")
    connected_cd.to_csv(connected_cd_csv)

# Save Force-Directed Layout
if not path.exists(connected_fd_csv):
    print(f"Saving file to: {connected_fd_csv}")
    connected_fd.to_csv(connected_fd_csv)
```

```
# Also plot bundled edge versions
      %time cd_b = graph_plot(connected_cd, hammer_bundle(connected_cd,__
       Gonnected_edges, weight='weight'), "Connected Bundled Circular Layout")
      %time fd_b = graph_plot(connected_fd, hammer_bundle(connected_fd,__
       -connected_edges, weight='weight'), "Connected Bundled Force-Directed Layout")
     CPU times: total: 2.33 s
     Wall time: 2.33 s
     CPU times: total: 1.59 s
     Wall time: 1.59 s
     CPU times: total: 26min 35s
     Wall time: 11min 36s
     CPU times: total: 8min 17s
     Wall time: 5min 33s
[21]: # Display images
      tf.Images(cd_n, fd_n, cd_b, fd_b).cols(2)
[21]: <datashader.transfer_functions.Images at 0x1cf415b4e50>
     Now that there is a single, connected network, community detection can be carried out. First, the
     Louvain algorithm:
[21]: # Make the new network
      connected = nx.from_pandas_edgelist(connected_edges)
      print(f"Number of Nodes: {nx.number of nodes(connected)}")
     Number of Nodes: 65648
[22]: # Make dictionary of community labels
      %time louvain_dict = community.best_partition(connected)
      # Unfold dict into a DataFrame
      %time louvain_labels = pd.DataFrame(list(louvain_dict.items()),__
       ⇔columns=['node', 'community'])
      print(f"There are {louvain labels['community'].nunique()} communities")
      print(louvain_labels.head)
     CPU times: total: 22.1 s
     Wall time: 22.1 s
     CPU times: total: 31.2 ms
     Wall time: 27.3 ms
     There are 75 communities
     <bound method NDFrame.head of</pre>
                                        node community
                           0
               1
     1
            34580
                            0
     2
              117
                            0
```

```
      3
      47740
      0

      4
      2
      0

      ...
      ...
      ...

      65643
      67167
      0

      65644
      67169
      15

      65645
      67170
      15

      65646
      67171
      15

      65647
      55798
      7
```

[65648 rows x 2 columns]>

Following function gets a set of colours equally spaced across a given colormap, for the display of communities.

Following function plots categorised nodes with the previous calculated Circular and Force-Directed node positions.

```
[24]: def plot_categories(nodes, edges, labels, name="", cat=None):
    n_comms = labels['community'].nunique()

    community_labels = np.sort(labels['community'].unique())

# Get colours for all communities
    colours = get_colours(n_comms, "hsv")

# Create canvas
    xr = nodes.x.min(), nodes.x.max()
    yr = nodes.y.min(), nodes.y.max()
    canvas = ds.Canvas(x_range=xr, y_range=yr, **cvsopts)

node_plots = []

for community in community_labels:
    # DataFrame of nodes in current community
    community_list = labels.loc[labels['community'] == community]
```

```
# Plot subset of nodes with specific colour
sub_nodes = nodes.loc[nodes.index.isin(community_list['node'])]

node_plot = nodes_plot(sub_nodes, f"Community {community}", canvas,u
cat=cat, cmap=colours[community])
node_plots.append(node_plot)

edge_plot = edges_plot(edges, name + " edges", canvas)

return tf.stack(edge_plot, *node_plots, how="over", name=name)
```

CPU times: total: 4.78 s Wall time: 4.79 s CPU times: total: 4.02 s Wall time: 4.02 s

[106]: <datashader.transfer_functions.Images at 0x168e48784f0>

These visualisations do not show evident communities as there are too many nodes, too closely clustered together. Next, create a grid of images showing each community by itself.

```
[25]: def plot_single_community(nodes, edges, labels, n, name="", cat=None):
    n_comms = labels['community'].nunique()

community_labels = np.sort(labels['community'].unique())

# Get colours for all communities
    colours = get_colours(n_comms, "hsv")
    col = colours[n]

# Create canvas
    xr = nodes.x.min(), nodes.x.max()
    yr = nodes.y.min(), nodes.y.max()
    canvas = ds.Canvas(x_range=xr, y_range=yr, **cvsopts)

# DataFrame of nodes in current community
    community_list = labels.loc[labels['community'] == n]
```

```
# Plot subset of nodes with specific colour
sub_nodes = nodes.loc[nodes.index.isin(community_list['node'])]
sub_node_names = set(sub_nodes['name'])
node_plot = nodes_plot(sub_nodes, f"Community {n}", canvas, cat=cat,u
cmap=col)
edges_pos = directly_connect_edges(sub_nodes, edges)
edge_plot = edges_plot(edges_pos, name + " edges_pos", canvas)
return tf.stack(edge_plot, node_plot, how="over", name=name)
```

```
imgs = []
for n in range(0, louvain_labels['community'].nunique()):
    single_cd = plot_single_community(connected_cd, connected_edges,
    louvain_labels, n, name=f"Circular, Community {n}")
    single_fd = plot_single_community(connected_fd, connected_edges,
    louvain_labels, n, name=f"Force-Directed, Community {n}")

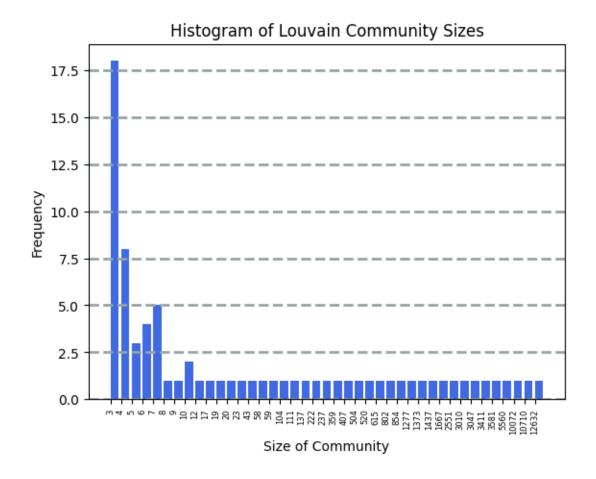
imgs.append(single_cd)
    imgs.append(single_fd)

tf.Images(*imgs).cols(2)
```

[29]: <datashader.transfer_functions.Images at 0x1d034a8a850>

```
plt.grid(color='#95a5a6', linestyle='--', linewidth=2, axis='y')
plt.show()
```

| <box></box> | method N | NDFrame.head | of | size | frequency |
|-------------|----------|--------------|----|------|-----------|
| 0 | 3 | 18 | | | |
| 1 | 4 | 8 | | | |
| 2 | 5 | 3 | | | |
| 3 | 6 | 4 | | | |
| 4 | 7 | 5 | | | |
| 5 | 8 | 1 | | | |
| 6 | 9 | 1 | | | |
| 7 | 10 | 2 | | | |
| 8 | 12 | 1 | | | |
| 9 | 17 | 1 | | | |
| 10 | 19 | 1 | | | |
| 11 | 20 | 1 | | | |
| 12 | 23 | 1 | | | |
| 13 | 43 | 1 | | | |
| 14 | 58 | 1 | | | |
| 15 | 59 | 1 | | | |
| 16 1 | 04 | 1 | | | |
| 17 1 | 11 | 1 | | | |
| 18 1 | 37 | 1 | | | |
| 19 2 | 22 | 1 | | | |
| 20 2 | 37 | 1 | | | |
| 21 3 | 59 | 1 | | | |
| 22 4 | 07 | 1 | | | |
| 23 5 | 04 | 1 | | | |
| 24 5 | 20 | 1 | | | |
| 25 6 | 15 | 1 | | | |
| 26 8 | 02 | 1 | | | |
| 27 8 | 54 | 1 | | | |
| 28 12 | 77 | 1 | | | |
| 29 13 | 73 | 1 | | | |
| 30 14 | 37 | 1 | | | |
| 31 16 | 67 | 1 | | | |
| 32 25 | 51 | 1 | | | |
| 33 30 | 10 | 1 | | | |
| 34 30 | 47 | 1 | | | |
| 35 34 | 11 | 1 | | | |
| 36 35 | 81 | 1 | | | |
| 37 55 | 60 | 1 | | | |
| 38 100 | 72 | 1 | | | |
| 39 107 | 10 | 1 | | | |
| 40 126 | 32 | 1> | | | |



Next, perform Asynchronous Label Propagation algorithm

```
[27]: %time lpa = comm.asyn_lpa_communities(connected, weight='weight')
    CPU times: total: 0 ns

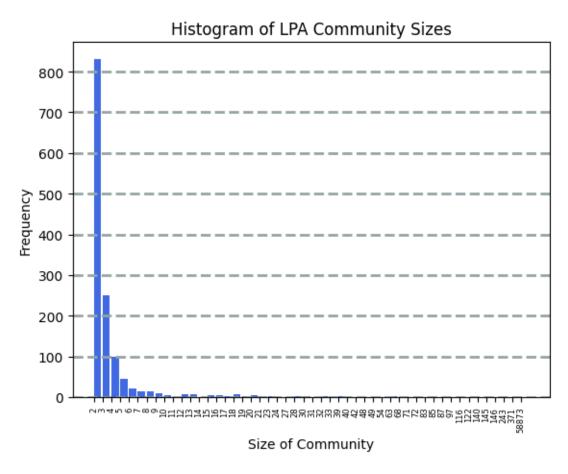
[28]: %time lpa_list = list(lpa)
    CPU times: total: 5.39 s
    Wall time: 5.4 s

[29]: nodes_to_add = []
    communities_to_add = []
    for i, sublist in enumerate(lpa_list):
        for node in sublist:
            nodes_to_add.append(node)
            communities_to_add.append(i)
```

```
to_add = {'node': nodes_to_add, 'community': communities_to_add}
       lpa_labels = pd.DataFrame.from_dict(to_add)
       print(lpa_labels.head)
      <bound method NDFrame.head of</pre>
                                             node community
                 1
      1
                 2
                             0
      2
                 3
                             0
                 4
      3
                             0
      4
                 5
                             0
      65643 46162
                         1364
                         1365
      65644 47216
      65645 65781
                         1365
      65646 65784
                          1366
      65647 47258
                         1366
      [65648 \text{ rows x 2 columns}] >
[113]: | # Plot the communities using the same layout as found previously
       %time cat cd = plot categories(connected cd,
       directly_connect_edges(connected_cd, connected_edges), lpa_labels,_
       ⇔name="Circular Layout Categorised")
       %time cat_fd = plot_categories(connected_fd,__
        directly_connect_edges(connected_fd, connected_edges), lpa_labels,
        →name="Force-Directed Layout Categorised")
       tf.Images(cat_cd, cat_fd).cols(2)
      CPU times: total: 1min 9s
      Wall time: 1min 9s
      CPU times: total: 1min
      Wall time: 1min
[113]: <datashader.transfer_functions.Images at 0x1688e70ab80>
[30]: hist = lpa_labels.groupby('community').size().reset_index(name='number')
       hist = hist.groupby('number').size().reset_index(name='frequency').
        ⇔rename(columns={'number': 'size'})
       print(hist.head)
       # Plot a histogram of the 'number' column
       plt.xticks(range(len(hist['frequency'])), hist['size'], rotation='vertical',_
        →fontsize=6)
       # Add axis labels and a title
```

```
<bound method NDFrame.head of</pre>
                                         size frequency
                    831
1
         3
                    251
2
         4
                     99
3
         5
                     45
4
         6
                     22
5
         7
                     14
6
         8
                     15
7
                      9
         9
8
                      5
        10
9
        11
                      1
        12
                      7
10
11
        13
                      6
12
        14
                      1
13
        15
                      4
14
        16
                      4
15
                      2
        17
16
        18
                      6
17
        19
                      2
18
        20
                      4
19
        21
                      3
                      2
20
        23
21
        24
                      1
22
        27
                      1
23
                      2
        28
24
        30
                      2
25
        31
                      1
                      2
26
        32
27
        33
                      1
                      2
28
        39
29
        40
                      1
30
        42
                      1
31
        48
                      1
32
        49
                      1
33
        54
                      1
34
        63
                      2
35
        68
                      1
36
        71
                      1
37
        72
                      1
38
        83
                      1
```

| 39 | 85 | 1 |
|----|-------|----|
| 40 | 87 | 1 |
| 41 | 97 | 1 |
| 42 | 116 | 1 |
| 43 | 122 | 1 |
| 44 | 140 | 1 |
| 45 | 145 | 1 |
| 46 | 146 | 1 |
| 47 | 243 | 1 |
| 48 | 371 | 1 |
| 49 | 58873 | 1> |



Calculate TPR, Conductance and Modularity of Louvain Labels and LPA Labels.

```
[59]: # Convert Louvain Labels into a NodeClustering object, so that I can
# perform evaluation on them
louvain_list = [[] for _ in range(max(louvain_dict.values()) + 1)]

for node, community in louvain_dict.items():
    louvain_list[community].append(node)
```

```
louvain_coms = NodeClustering(louvain_list, graph=connected,_
       →method_name='Louvain')
[60]: # Calculate Conductance, TPR and Modularity:
      louvain_conductance = conductance(connected, louvain_coms, summary=True)
      louvain_tpr = triangle_participation_ratio(connected, louvain_coms,_
       ⇔summary=True)
      louvain_modularity = newman_girvan_modularity(connected, louvain_coms,__
       ⇔summary=True)
      print(f"Conductance: {louvain_conductance}")
      print(f"TPR: {louvain tpr}")
      print(f"Modularity: {louvain_modularity}")
     Conductance: FitnessResult(min=0.0027247956403269754, max=0.6515609264853978,
     score=0.23753507740180022, std=0.13220832166822985)
     TPR: FitnessResult(min=0.0, max=1.0, score=0.24036557171981135,
     std=0.2877272691911959)
     Modularity: FitnessResult(min=None, max=None, score=0.41738497141682823,
     std=None)
[61]: | # Convert the LPA labels into a NodeClustering object, so that I can
      # perform evaluation on them
      lpa_coms = NodeClustering(lpa_list, graph=connected, method_name='LPA')
[62]: # Calculate Conductance, TPR and Modularity:
      lpa_conductance = conductance(connected, lpa_coms, summary=True)
      lpa tpr = triangle participation ratio(connected, lpa coms, summary=True)
      lpa_modularity = newman_girvan_modularity(connected, lpa_coms, summary=True)
      print(f"Conductance: {lpa_conductance}")
      print(f"TPR: {lpa_tpr}")
      print(f"Modularity: {lpa_modularity}")
     Conductance: FitnessResult(min=0.008459009567176759, max=0.7142857142857143,
     score=0.36287489337285733, std=0.10288671520150713)
     TPR: FitnessResult(min=0.0, max=1.0, score=0.03816279882849793,
     std=0.15726532973048263)
     Modularity: FitnessResult(min=None, max=None, score=0.05204184799923872,
     std=None)
```