```
pip install pyvis

→ Collecting pyvis

       Downloading pyvis-0.3.2-py3-none-any.whl.metadata (1.7 kB)
     Requirement already satisfied: ipython>=5.3.0 in /usr/local/lib/python3.10/dist-packages (from pyvis) (7.34.0)
     Requirement already satisfied: jinja2>=2.9.6 in /usr/local/lib/python3.10/dist-packages (from pyvis) (3.1.4)
     Requirement already satisfied: jsonpickle>=1.4.1 in /usr/local/lib/python3.10/dist-packages (from pyvis) (3.3.0)
     Requirement already satisfied: networkx>=1.11 in /usr/local/lib/python3.10/dist-packages (from pyvis) (3.4.2)
     Requirement already satisfied: setuptools>=18.5 in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (75.1.0)
     Collecting jedi>=0.16 (from ipython>=5.3.0->pyvis)
       Downloading jedi-0.19.1-py2.py3-none-any.whl.metadata (22 kB)
     Requirement already satisfied: decorator in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (4.4.2)
     Requirement already satisfied: pickleshare in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (0.7.5)
     Requirement already satisfied: traitlets>=4.2 in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (5.7.1)
     Requirement already satisfied: prompt-toolkit!=3.0.0,!=3.0.1,<3.1.0,>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from ipython>=5.
     Requirement already satisfied: pygments in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (2.18.0)
     Requirement already satisfied: backcall in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (0.2.0)
     Requirement already satisfied: matplotlib-inline in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (0.1.7)
     Requirement already satisfied: pexpect>4.3 in /usr/local/lib/python3.10/dist-packages (from ipython>=5.3.0->pyvis) (4.9.0)
     Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2>=2.9.6->pyvis) (3.0.2)
     Requirement already satisfied: parso<0.9.0,>=0.8.3 in /usr/local/lib/python3.10/dist-packages (from jedi>=0.16->ipython>=5.3.0->pyvis) (
     Requirement already satisfied: ptyprocess>=0.5 in /usr/local/lib/python3.10/dist-packages (from pexpect>4.3->ipython>=5.3.0->pyvis) (0.7
     Requirement already satisfied: wcwidth in /usr/local/lib/python3.10/dist-packages (from prompt-toolkit!=3.0.0,!=3.0.1,<3.1.0,>=2.0.0->ip
     Downloading pyvis-0.3.2-py3-none-any.whl (756 kB)
                                                 756.0/756.0 kB 19.3 MB/s eta 0:00:00
     Downloading jedi-0.19.1-py2.py3-none-any.whl (1.6 MB)
                                                 1.6/1.6 MB 38.0 MB/s eta 0:00:00
     Installing collected packages: jedi, pyvis
     Successfully installed jedi-0.19.1 pyvis-0.3.2
import csv
import os
import sys
import json
import pandas as pd
import networkx as nx
from operator import itemgetter
from networkx.algorithms import community
from pyvis.network import Network
from collections import Counter
import numpy as np
import matplotlib.pyplot as plt
Start coding or generate with AI.
```

Read the Data

```
from google.colab import drive
drive.mount('/content/drive')
→ Mounted at /content/drive
!ls '/content/drive/My Drive/Abo Akademi/Data science/Data-Miniproject 3'
→ artist_institution_network.graphml interactive_network.html
                                                                             network_visualization.png
     artists.csv
                                                                             relationships.csv
                                         lib
                                         network.graphml
                                                                             schools.csv
     fig1.jpg
     fig1.pdf
                                         network_visualization_high_res.png
     institutions.csv
                                         network_visualization_new.png
os.chdir("/content/drive/My Drive/Abo Akademi/Data science/Data-Miniproject 3")
artists = pd.read_csv('artists.csv')
artists = artists.dropna(subset=['artistUrl','id','image','nation','title','totalWorksTitle','year'])
artists
```

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	id	image	nation	title	totalWorksTitle	year
/en/ancient- egyptian- painting	5d230b31edc2c9fb74ae130a	https://uploads2.wikiart.org/00244/images/anci	Egyptians	Ancient Egypt	169 artworks	c.3150 BC - c.640 AD
/en/ancient- greek-pottery	5ced09b1edc2c993247c1ea0	https://uploads7.wikiart.org/00238/images/anci	Greeks	Ancient Greek Pottery	87 artworks	c.900 - 31 BC
/en/ancient- greek-painting	5ce3c54dedc2c921c4035b7d	https://uploads8.wikiart.org/00237/images/anci	Greeks	Ancient Greek Painting and Sculpture	76 artworks	c.700 - c.31 BC
/en/apelles	5ba232ccedc2c918cc654ea8	https://uploads2.wikiart.org/00202/images/apel	Greek	Apelles	8 artworks	born c.370 BC
/en/fayum- portrait	5beea1ededc2c915a01acdda	https://uploads7.wikiart.org/00210/images//01	Egyptians	Fayum portrait	76 artworks	c.50 - c.250
	egyptian- painting /en/ancient- greek-pottery /en/ancient- greek-painting /en/apelles /en/fayum-	egyptian- painting /en/ancient- greek-pottery /en/ancient- greek-painting 5ced09b1edc2c993247c1ea0 /en/ancient- greek-painting 5ce3c54dedc2c921c4035b7d /en/apelles 5ba232ccedc2c918cc654ea8 /en/fayum- 5beea1ededc2c915a01acdda	egyptian-painting /en/ancient-greek-pottery /en/ancient-greek-painting /en/ancient-greek-painting /en/ancient-greek-painting 5ce3c54dedc2c921c4035b7d https://uploads7.wikiart.org/00238/images/anci /en/apelles 5ba232ccedc2c918cc654ea8 https://uploads2.wikiart.org/00202/images/apel /en/fayum- /en/fayum- 5beea1ededc2c915a01acdda https://uploads7.wikiart.org/00210/images//01	egyptian- painting /en/ancient- greek-pottery /en/ancient- greek-painting /en/ancient- greek-painting fen/ancient- greek-painting /en/ancient- greek-painting fen/apelles fen/apelles fen/apelles fen/fayum- fen/fayum-	egyptian-painting fen/ancient-greek-pottery fen/ancient-greek-painting fen/ancient-greek-painting fen/ancient-greek-painting fen/apelles fen/apelles fegyptians fegypt	egyptian-painting /en/ancient-greek-pottery /en/ancient-greek-painting fen/ancient-greek-painting /en/apelles fen/apelles fen/apelles fen/apelles fen/apelles fen/fayum- fen/fayum- fen/fayum- fen/fayum- fen/fayum- fen/fayum- fen/fayum- fen/apelles fen/fayum- fen/fayum- fen/fayum- fen/apelles fen/fayum- fen/fayum-

relationships = pd.read_csv('relationships.csv')
relationships = relationships.dropna(subset=['artistUrl', 'friends', 'influenced_by','influenced_on','institution','movements','school','typrelationships.head()

_		artistUrl	friends	influenced_by	influenced_on	institution	movements
	/en/rogier- 75 van-der- weyden /en/jacques-daret /en/pieter- bruegel- the-elder /en/maarten-de- vos,/en/giulio-clovio		- /en/jacques-daret /en/jan-van-eyck,/en/stefan-lochner me		/en/antonello-da- messina,/en/hans- memling,/en/	/en/artists-by- art- institution/guild- of-saint	Northern Renaissance
			, - , , , , , , , , , , , , , , , , , ,	/en/hieronymus-hosch verhaecht /ei		/en/artists-by- art- institution/guild- of-saint	Northern Renaissance
	131	/en/jean- francois- millet	/en/paul- delaroche,/en/constant- troyon,/en/nar	/en/honore-daumier,/en/michelangelo,/en/nicola	/en/edward-mitchell- bannister,/en/andres- de-sa	/en/artists-by- art- institution/cole- des-beaux	Realism s
	/en/ford- /en/william- /en/w 169 madox- morris,/en/edward- brown burne-jones		madox- morris,/en/edward- /en/william-holman-hunt,/en/john- /en/dal		/en/dante-gabriel- rossetti	/en/artists-by- art- institution/royal- academy-o	Romanticism
4						•	

institutions = pd.read_csv('institutions.csv')

$\verb"institutions"$

	city	country	title	url
0	NaN	NaN	Guild of Saint Luke	/en/artists-by-art-institution/guild-of-saint
1	Antwerp	Belgium	Royal Academy of Fine Arts	/en/artists-by-art-institution/royal-academy-o
2	Munich	Germany	Akademie der Bildenden Künste München (Munich	/en/artists-by-art-institution/akademie-der
3	Milan	Italy	Accademia di Belle Arti di Brera (Brera Academy)	/en/artists-by-art-institution/accademia-di-be
4	Paris	France	Académie des Beaux-Arts	/en/artists-by-art-institution/acad-mie-des-beaux
68	East Sussex	UK	Hastings College of Arts and Technology (HCAT)	/en/artists-by-art-institution/hastings-colleg
69	Odessa	Ukraine	Grekov Odessa Art School	/en/artists-by-art-institution/grekov-odessa-a
70	Cambridge	MA	American Academy of Arts and Sciences	/en/artists-by-art-institution/american-academ
71	London	UK	Central Saint Martins (CSM)	/en/artists-by-art-institution/central-saint-m
72	Chicago	IL	School of the Art Institute of Chicago (SAIC)	/en/artists-by-art-institution/school-of-the-art
73 rc	ws × 4 column	S		
4				

```
schools = pd.read_csv('schools.csv')
schools
```

```
<del>_</del>
                       title
                                                                          ur1
       0
              Flemish School
                                  /en/artists-by-painting-school/flemish-school
       1
              Sienese School
                                 /en/artists-by-painting-school/sienese-school
       2
              Umbrian school
                                 /en/artists-by-painting-school/umbrian-school
       3
            Florentine School
                                /en/artists-by-painting-school/florentine-school
                Parma school
       4
                                  /en/artists-by-painting-school/parma-school
      215
                Cádiz School
                                    /en/artists-by-painting-school/c-diz-school
      216
                   Wu School
                                      /en/artists-by-painting-school/wu-school
      217
               Bruges School
                                  /en/artists-by-painting-school/bruges-school
      218
             Milanese school
                                /en/artists-by-painting-school/milanese-school
      219
               Danube school
                                  /en/artists-by-painting-school/danube-school
     220 rows × 2 columns
```

~ EDA

```
print("Artists Dataset Overview:")
print(artists.info())
print("Missing Values in Artists Dataset:", artists.isnull().sum())
print("\nRelationships Dataset Overview:")
print(relationships.info())
print("Missing Values in Relationships Dataset:", relationships.isnull().sum())
print("\nInstitutions Dataset Overview:")
print(institutions.info())
print("Missing Values in Institutions Dataset:", institutions.isnull().sum())
print("\nSchools Dataset Overview:")
print(schools.info())
print("Missing Values in Schools Dataset:", schools.isnull().sum())
# Distribution analysis
print("\nNation Distribution among Artists:")
print(artists['nation'].value_counts().head(10))
print("\nActive Year Intervals among Artists:")
print(artists['year'].value_counts().head(10))
print("\nNumber of Artworks Distribution:")
print(artists['totalWorksTitle'].describe())
→ Artists Dataset Overview:
     <class 'pandas.core.frame.DataFrame'>
     Index: 2963 entries, 0 to 2995
     Data columns (total 7 columns):
     # Column
                          Non-Null Count Dtype
     0
         artistUrl
                          2963 non-null object
                          2963 non-null
     1
         id
                                         object
     2 image
                          2963 non-null object
         nation
                          2963 non-null
                                          object
     4
                          2963 non-null object
         title
         totalWorksTitle 2963 non-null
                                          object
                           2963 non-null
                                          object
         year
     dtypes: object(7)
     memory usage: 185.2+ KB
     Missing Values in Artists Dataset: artistUrl
                       0
     id
     image
                       0
                       0
     nation
     title
                       0
     totalWorksTitle
                       0
     year
```

```
dtype: int64
     Relationships Dataset Overview:
     <class 'pandas.core.frame.DataFrame'>
Index: 58 entries, 75 to 2168
     Data columns (total 8 columns):
                    Non-Null Count Dtype
     # Column
     ---
                        _____
     0 artistUrl 58 non-null
1 friends 58 non-null
                                         object
                                         object
      2 influenced_by 58 non-null
                                         object
      3 influenced_on 58 non-null
4 institution 58 non-null
                                         object
                                         object
      5 movements
                        58 non-null
                                         object
                        58 non-null
      6 school
                                         object
         type
                         58 non-null
                                         object
     dtypes: object(8)
     memory usage: 4.1+ KB
     None
     Missing Values in Relationships Dataset: artistUrl
     friends
                     0
     influenced_by
                      0
     influenced_on
     institution
                      0
                      0
     movements
     school
                      0
     type
                      0
     dtype: int64
     Institutions Dataset Overview:
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 73 entries, 0 to 72
     Data columns (total 4 columns):
     # Column Non-Null Count Dtype
print("Artists Dataset Summary Statistics:")
display(artists.describe())
print('\n\n')
print("Relationships Dataset Summary Statistics:")
display(relationships.describe())
print('\n\n')
print("Institutions Dataset Summary Statistics:")
display(institutions.describe())
print('\n\n')
print("Schools Dataset Summary Statistics:")
display(schools.describe())
```

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→ Artists Dataset Summary Statistics:

	artistUrl	id	image	nation	title	totalWorksTitle	year
count	2963	2963	2963	2963	2963	2963	2963
unique	2963	2963	2963	106	2963	294	2181
top	/en/ferdinand- georg-waldm- ller	57726d96edc2cb3880b4b06a	https://uploads6.wikiart.org/00381/images/ferd	American	Ferdinand Georg Waldmüller	15 artworks	born 1937
freq	1	1	1	519	1	137	24

Relationships Dataset Summary Statistics:

	artistUrl	friends	influenced_by	influenced_on	institution	movements	school	type
count	58	58	58	58	58	58	58	58
unique	58	53	58	57	45	39	46	1
top	/en/rogier-van- der-weyden	/en/jackson- pollock	/en/jan-van- eyck,/en/stefan- lochner	/en/charles- gibbons	/en/artists-by-art- institution/art-students- le	Expressionism	/en/artists-by-painting- school/pre- raphaelite	Artists
freq	1	6	1	2	4	6	3	58

Institutions Dataset Summary Statistics:

url	title	country	city	
73	73	71	71	count
73	73	35	52	unique
/en/artists-by-art-institution/school-of-the-art	School of the Art Institute of Chicago (SAIC)	UK	Paris	top
1	1	8	7	freq

Schools Dataset Summary Statistics:

	title	url
count	220	220
unique	220	220

 $\mbox{\tt\#}$ Making the artist name based on artistUrl in relationships file from artists file relationships_art = relationships.merge(

 ${\tt artists}[['{\tt artistUrl'},\ '{\tt title'}]],\ \ {\tt\#}\ {\tt Select}\ \ {\tt the}\ \ '{\tt artistUrl'}\ \ {\tt and}\ \ '{\tt title'}\ \ {\tt columns}\ \ {\tt from}\ \ {\tt artists}$

on='artistUrl', # Join based on 'artistUrl'

how='left' # Use 'left' join to keep all relationships data

).rename(columns={'title': 'artist_name'})

relationships_art.head()

₹		artistUrl	f	riends		influenced_	_by	inf	luence	d_on	insti	itutior	1			mo	vement	s
	/en/rogier- 0 van-der- /en/jacques-daret /en/jan-van-eyck,/en/stefan-loweyden			.,/en/stefan-loch	ner r	messi	antonell na,/en/l emling,/	nans-	institutio		art- Northern Ren guild-			ern Ren	aissanc	e sc		
	/en/pieter- 1 bruegel- the-elder /en/maarten-de- /en/hieronymus-bos						/en/to ht,/en/prubens,	eter-	art-			Northern Renaissance				/ e sc		
	/en/jean- /en/paul- /en/honore- francois- delaroche,/en/constant- millet troyon,/en/nar daumier,/en/michelangelo,/en/nicola				ore-		ard-mit r,/en/an de		instituti	tists-by- art- on/cole- beaux	Realism sch				n sch			
	/en/ford- /en/william- 3 madox- morris,/en/edward- brown burne-jones		edward-	/en/william-holman-hunt,/en/john- everett-milla			/en/da	ante-ga ro		institutio	tists-by- art- on/royal- emy-o				Rom	nanticisn	/ n	
	•	/en/william-	/en/dante- ,		1	/en/jo	hn-	/en/	/ford-ma	adox-	/en/ar	tists-by- art			· ·			<i>/</i>
split_	_fr		column for sep ionships_art[
₹		0	1	2	3	4		5	6	7	8	9	10	11	12	13	14	15
	0	/en/jacques- daret	None	None	None	None	N	lone	None	None	None	None	None	None	None	None	None	None
	1	/en/maarten- de-vos	/en/giulio- clovio	None	None	None	N	lone	None	None	None	None	None	None	None	None	None	None
	2	/en/paul- delaroche	/en/constant- troyon	/en/narcisse- virgilio-diaz	/en/charles- jacque	/en/theodore- rousseau	/en/cam	nille- corot	None	None	None	None	None	None	None	None	None	None
	3	/en/william- morris	/en/edward-	None	None	None	N	lone	None	None	None	None	None	None	None	None	None	None

Rename columns with a naming pattern or specific names
split_fr_df.columns = [f'Fr_{i+1}' for i in range(split_fr_df.shape[1])]
split_fr_df.head()

Fr_2 Fr_3 Fr_5 Fr_7 Fr_8 Fr_9 Fr_10 Fr_11 Fr_12 Fr_13 Fr_14 Fr_15 F Fr_1 Fr_4 /en/jacques-None None daret /en/maarten-/en/giulio-None None de-vos clovio /en/paul-/en/constant- /en/narcisse- /en/charles- /en/theodore-/en/camille-None None None None None None None None None delaroche troyon virgilio-diaz jacque rousseau corot /en/william-/en/edward-None None morris burne-jones

art_fr = artists.rename(columns={'artistUrl':'Fr_1'})
art_fr.head()

morris burne-jones

₹

		Fr_1	id	image	nation	title	totalWorksTitle	year
	0	/en/ancient- egyptian- painting	5d230b31edc2c9fb74ae130a	https://uploads2.wikiart.org/00244/images/anci	Egyptians	Ancient Egypt	169 artworks	c.3150 BC - c.640 AD
	1	/en/ancient- reek-pottery	5ced09b1edc2c993247c1ea0	https://uploads7.wikiart.org/00238/images/anci	Greeks	Ancient Greek Pottery	87 artworks	c.900 - 31 BC
	· -	/en/ancient-	E0020E4d0d0200240402Eb7d	https://uploodo?.wikipt.org/00227/imagos/opsi	Crooks	Ancient Greek	76 ortworks	c.700 -

```
\verb|art_fr[['Fr_1', 'title']]|, # Select the 'artistUrl' and 'title' columns from artists| \\
   on='Fr 1',
                              # Join based on 'artistUrl'
                                # Use 'left' join to keep all relationships data
   how='left'
).rename(columns={'title': 'Fr_1_name'})
#for Fr2 column
art_fr = artists.rename(columns={'artistUrl':'Fr_2'})
relationships_fr = relationships_fr.merge(
   art_fr[['Fr_2', 'title']], # Select the 'artistUrl' and 'title' columns from artists
   on='Fr_2',
                             # Join based on 'artistUrl'
   how='left'
                               # Use 'left' join to keep all relationships data
).rename(columns={'title': 'Fr_2_name'})
# for fr3 column
art_fr = artists.rename(columns={'artistUrl':'Fr_3'})
relationships_fr = relationships_fr.merge(
   art_fr[['Fr_3', 'title']], # Select the 'artistUrl' and 'title' columns from artists
                       # Join based on 'artistUrl'
   on='Fr_3',
   how='left'
                                 # Use 'left' join to keep all relationships data
).rename(columns={'title': 'Fr_3_name'})
# for fr4 column
art_fr = artists.rename(columns={'artistUrl':'Fr_4'})
relationships_fr = relationships_fr.merge(
   on='Fr_4',
                             # Join based on 'artistUrl'
                                 # Use 'left' join to keep all relationships data
   how='left'
).rename(columns={'title': 'Fr_4_name'})
relationships_fr
```

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	Fr_1	Fr_2	Fr_3	Fr_4	Fr_5	Fr_6	Fr_7	Fr_8	Fr_9	Fr_10
0	/en/jacques- daret	None	None	None	None	None	None	None	None	None
1	/en/maarten- de-vos	/en/giulio- clovio	None	None	None	None	None	None	None	None
2	/en/paul- delaroche	/en/constant- troyon	/en/narcisse- virgilio-diaz	/en/charles- jacque	/en/theodore- rousseau	/en/camille- corot	None	None	None	None
3	/en/william- morris	/en/edward- burne-jones	None	None	None	None	None	None	None	None
4	/en/dante- gabriel-rossetti	/en/john- everett-millais	/en/edward- robert-hughes	/en/herbert- gustave- schmalz	None	None	None	None	None	None
5	/en/william- holman-hunt	None	None	None	None	None	None	None	None	None
6	/en/henri- fantin-latour	/en/edouard- manet	None	None	None	None	None	None	None	None
7	/en/ilya-repin	/en/ivan- kramskoy	/en/grigoriy- myasoyedov	/en/vladimir- makovsky	/en/mikhail- vrubel	/en/taras- shevchenko	/en/nikolai-ge	/en/petro- kholodny- elder	None	None
8	/en/vasily- polenov	/en/alexandre- benois	/en/mykola- murashko	None	None	None	None	None	None	None
9	/en/paul- cezanne	/en/vincent- van-gogh	/en/stanislaw- wyspianski	/en/emile- bernard	/en/meijer-de- haan	/en/charles- laval	/en/roderic- oconor	/en/ferdinand- du- puigaudeau	None	None
10	/en/julian- alden-weir	/en/henry- tonks	/en/martin-rico	/en/frank- omeara	/en/william- logsdail	None	None	None	None	None
11	/en/mykola- murashko	/en/valentin- serov	/en/konstantin- korovin	None	None	None	None	None	None	None
12	/en/edward-e- simmons	None	None	None	None	None	None	None	None	None
13	/en/paul-klee	/en/franz-marc	/en/alexej-von- jawlensky	/en/vadym- meller	None	None	None	None	None	None
14	/en/xavier- martinez	/en/emilie- charmy	/en/henri- manguin	None	None	None	None	None	None	None
15	/en/albert- julius-olsson	None	None	None	None	None	None	None	None	None
16	/en/marc- chagall	/en/fernand- leger	/en/jan-toorop	/en/theo-van- doesburg	/en/roman- selsky	/en/margit- selska	None	None	None	None
17	/en/olga- rozanova	/en/lyubov- popova	/en/aleksandra- ekster	/en/marc- chagall	/en/laszlo- moholy-nagy	/en/vadym- meller	/en/pavel- filonov	None	None	None
18	/en/josef- albers	/en/wassily- kandinsky	/en/august- macke	/en/vadym- meller	None	None	None	None	None	None
19	/en/georges- braque	/en/constantin- brancusi	/en/man-ray	/en/max-jacob	/en/amedeo- modigliani	/en/christopher- wood	/en/mykhailo- boychuk	/en/c-r-w- nevinson	/en/valentine- hugo	/en/lee- miller
20	/en/vsevolod- maksymovych	None	None	None	None	None	None	None	None	None
21	/en/george- bellows	/en/rockwell- kent	/en/charles-e- burchfield	None	None	None	None	None	None	None
22	/en/marc- chagall	/en/piet- mondrian	/en/le-corbusier	/en/robert- delaunay	/en/constantin- brancusi	/en/roman- selsky	/en/alice-bailly	None	None	None
23	/en/edvard- munch	/en/august- macke	/en/wassily- kandinsky	/en/erich- heckel	None	None	None	None	None	None
24	/en/oleksandr- murashko	/en/pablo- picasso	/en/diego- rivera	/en/victor- palmov	/en/vasyl- yermylov	/en/oleksandr- bogomazov	/en/alexander- khvostenko- khvostov	None	None	None
25	/en/robert- delaunay	/en/alice-bailly	None	None	None	None	None	None	None	None
26	/en/diego-	/en/david- alfaro-	None	None	None	None	None	None	None	None

IIVOIU	siqueiros

27	/en/constantin- brancusi	/en/diego- rivera	/en/chaim- soutine	/en/moise- kisling	/en/jacques- lipchitz	/en/max-jacob	/en/pablo- picasso	/en/jacob- epstein	/en/juan-gris	/en/gino- severini
28	/en/egon- schiele	None	None	None	None	None	None	None	None	None
29	/en/franz-marc	/en/paul-klee	None	None	None	None	None	None	None	None
30	/en/amedeo- modigliani	/en/saturnino- herran	/en/robert- montenegro	/en/jose- clemente- orozco	/en/david- alfaro- siqueiros	/en/mykhailo- boychuk	/en/alice- rahon	/en/jacqueline- lamba	None	None
31	/en/robert- delaunay	/en/fernand- leger	/en/piet- mondrian	/en/kazimir- malevich	None	None	None	None	None	None
32	/en/alfred- stieglitz	/en/arthur- dove	/en/charles- demuth	/en/beauford- delaney	None	None	None	None	None	None
33	/en/amedeo- modigliani	/en/maruja- mallo	None	None	None	None	None	None	None	None
34	/en/paul-klee	None	None	None	None	None	None	None	None	None
35	/en/gustav- klimt	/en/oskar- kokoschka	None	None	None	None	None	None	None	None
36	/en/constantin- brancusi	/en/salvador- dali	/en/marcel- duchamp	/en/francis- picabia	/en/pablo- picasso	/en/alfred- stieglitz	/en/maurice- tabard	/en/lee-miller	None	None
37	/en/amedeo- modigliani	/en/juan-gris	None	None	None	None	None	None	None	None
38	/en/amedeo- modigliani	/en/michel- kikoine	/en/pinchus- kremegne	None	None	None	None	None	None	None
39	/en/frederic- edwin-church	/en/john- trumbull	/en/asher- brown-durand	None	None	None	None	None	None	None
40	/en/jackson- pollock	None	None	None	None	None	None	None	None	None
41	/en/jackson- pollock	None	None	None	None	None	None	None	None	None
42	/en/jackson- pollock	None	None	None	None	None	None	None	None	None
43	/en/jackson- pollock	None	None	None	None	None	None	None	None	None
44	/en/lee- krasner	/en/robert- motherwell	/en/mark- rothko	/en/willem-de- kooning	/en/barnett- newman	/en/philip- guston	/en/ad- reinhardt	None	None	None
45	/en/kenneth- noland	/en/jack- tworkov	/en/arshile- gorky	/en/david- alfaro- siqueiros	None	None	None	None	None	None
46	/en/jackson- pollock	/en/rosemarie- beck	None	None	None	None	None	None	None	None
47	/en/jackson- pollock	None	None	None	None	None	None	None	None	None
48	/en/jackson- pollock	None	None	None	None	None	None	None	None	None
49	/en/boris- kustodiev	/en/filipp- malyavin	/en/oleksandr- bogomazov	/en/heorhiy- narbut	/en/mykhailo- boychuk	None	None	None	None	None
50	/en/francis- bacon	None	None	None	None	None	None	None	None	None
51	/en/morris- louis	None	None	None	None	None	None	None	None	None
52	/en/morris- louis	/en/jose- clemente- orozco	/en/diego- rivera	/en/antonio- berni	None	None	None	None	None	None
53	/en/jasper- johns	/en/robert- rauschenberg	None	None	None	None	None	None	None	None
54	/en/larry-rivers	/en/philip- pearlstein	/en/jean- michel- basquiat	/en/francesco- clemente	/en/salvador- dali	/en/steve- kaufman	/en/jack- armstrong	None	None	None

5	twombly	None	None	None	None	None	None	None	None	None
5	6 /en/blinky- palermo	/en/sigmar- polke	/en/joseph- beuys	/en/georg- baselitz	/en/manfred- kuttner	/en/konrad-lueg	None	None	None	None
5	7 /en/gerhard- richter	/en/manfred- kuttner	/en/konrad- lueg	None	None	None	None	None	None	None
58	rows × 21 columns									

```
relationships_art[['Fr1','Fr2','Fr3','Fr4']]=relationships_fr[['Fr_1_name','Fr_2_name','Fr_3_name','Fr_4_name']]
relationships_art.head()
```

_		artistUrl	friends	influenced_by	influenced_on	institution	movements	
	0	/en/rogier- van-der- weyden	/en/jacques-daret	/en/jan-van-eyck,/en/stefan-lochner	/en/antonello-da- messina,/en/hans- memling,/en/	/en/artists-by- art- institution/guild- of-saint	Northern Renaissance	/ sc
	1	/en/pieter- bruegel- the-elder	/en/maarten-de- vos,/en/giulio-clovio	/en/hieronymus-bosch	/en/tobias- verhaecht,/en/peter- paul-rubens,/en	/en/artists-by- art- institution/guild- of-saint	Northern Renaissance	/ sc
	2	/en/jean- francois- millet	/en/paul- delaroche,/en/constant- troyon,/en/nar	/en/honore-daumier,/en/michelangelo,/en/nicola	/en/edward-mitchell- bannister,/en/andres- de-sa	/en/artists-by- art- institution/cole- des-beaux	Realism	/ sch
	3	/en/ford- madox- brown	/en/william- morris,/en/edward- burne-jones	/en/william-holman-hunt,/en/john- everett-milla	/en/dante-gabriel- rossetti	/en/artists-by- art- institution/royal- academy-o	Romanticism	1
	4	/en/william- holman- hunt	/en/dante-gabriel- rossetti,/en/john- everett-mi	/en/john- ruskin,/en/giotto,/en/ambrogio- lorenz	/en/ford-madox- brown	/en/artists-by- art- institution/royal- academy-of	Romanticism,Symbolism,Orientalism	1

Start coding or generate with AI.

).rename(columns={'title': 'fr_name'})

```
# art = artists_red.rename(columns={'artistUrl':'friends'})
# art

# relationships_fr = relationships_art.merge(
# art[['friends', 'title']], # Select the 'artistUrl' and 'title' columns from artists
# on='friends', # Join based on 'artistUrl'
# how='left' # Use 'left' join to keep all relationships data
```

relationships_fr

relationships_art.head()

₹	artistUrl	friends	influenced_by	influenced_on	institution	movements	
	/en/rogier- 0 van-der- weyden	/en/jacques-daret	/en/jan-van-eyck,/en/stefan-lochner	/en/antonello-da- messina,/en/hans- memling,/en/	/en/artists-by- art- institution/guild- of-saint	Northern Renaissance	/ sc
	/en/pieter- 1 bruegel- the-elder	/en/maarten-de- vos,/en/giulio-clovio	/en/hieronymus-bosch	/en/tobias- verhaecht,/en/peter- paul-rubens,/en	/en/artists-by- art- institution/guild- of-saint	Northern Renaissance	/ sc
	/en/jean- 2 francois- millet	/en/paul- delaroche,/en/constant- troyon,/en/nar	/en/honore-daumier,/en/michelangelo,/en/nicola	/en/edward-mitchell- bannister,/en/andres- de-sa	/en/artists-by- art- institution/cole- des-beaux	Realism	/ sch
	/en/ford- 3 madox- brown	/en/william- morris,/en/edward- burne-jones	/en/william-holman-hunt,/en/john- everett-milla	/en/dante-gabriel- rossetti	/en/artists-by- art- institution/royal- academy-o	Romanticism	1
	/en/william- 4 holman- hunt	/en/dante-gabriel- rossetti,/en/john- everett-mi	/en/john- ruskin,/en/giotto,/en/ambrogio- lorenz	/en/ford-madox- brown	/en/artists-by- art- institution/royal- academy-of	Romanticism,Symbolism,Orientalism	/

```
<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 58 entries, 0 to 57 \,
    Data columns (total 13 columns):
                     Non-Null Count Dtype
    # Column
    ---
                       -----
     0 artistUrl 58 non-null
1 friends 58 non-null
                                       object
                                       object
     2 influenced_by 58 non-null
3 influenced_on 58 non-null
                                       object
                                       object
     4 institution 58 non-null
                                       object
    4 11352-1
5 movements 58 non-null
58 non-null
                       58 non-null
                                       object
                                       object
     7 type
                      58 non-null
                                       object
     8 artist_name 58 non-null
                                       object
                       55 non-null
     9 Fr1
                                       object
     10 Fr2
                      38 non-null
                                       object
     11 Fr3
                       31 non-null
                                       object
     12 Fr4
                      24 non-null
                                       object
    dtypes: object(13)
    memory usage: 6.0+ KB
```

reduce the artists's node for better showing the graph
artists_red = artists.drop(artists.index[80:])
artists_red

}	artistUrl	id	image	nation	title	totalWorksTitle	year
	/en/ancient- egyptian- painting	5d230b31edc2c9fb74ae130a	https://uploads2.wikiart.org/00244/images/anci	Egyptians	Ancient Egypt	169 artworks	c.3150 BC - c.640 AD
,	/en/ancient- greek-pottery	5ced09b1edc2c993247c1ea0	https://uploads7.wikiart.org/00238/images/anci	Greeks	Ancient Greek Pottery	87 artworks	c.900 - 31 BC
:	/en/ancient- greek-painting	5ce3c54dedc2c921c4035b7d	https://uploads8.wikiart.org/00237/images/anci	Greeks	Ancient Greek Painting and Sculpture	76 artworks	c.700 - c.31 BC
;	3 /en/apelles	5ba232ccedc2c918cc654ea8	https://uploads2.wikiart.org/00202/images/apel	Greek	Apelles	8 artworks	born c.370 BC
	/en/fayum- portrait	5beea1ededc2c915a01acdda	https://uploads7.wikiart.org/00210/images//01	Egyptians	Fayum portrait	76 artworks	c.50 - c.250
4	/en/hans-						

pip install nltk

```
# from nltk.corpus import stopwords
# import nltk

# # Download stopwords if you haven't yet
# nltk.download('stopwords')

# # Load the English stopwords list
# stop_words = set(stopwords.words('english'))

# # Remove stopwords from the specified column
# artists_red['title'] = artists_red['title'].apply(lambda x: ' '.join([word for word in x.split() if word.lower() not in stop_words]))

# # Strip whitespace and remove any potential unwanted characters
# artists_red['title'] = artists_red['title'].str.strip()

# artists_red
```

```
# # Check for duplicate titles
# print(artists_red['title'].duplicated().sum())
# # Add only unique titles as nodes
# unique_titles = artists_red['title'].unique()
```

```
# artists_red['title'] = artists_red['title'].astype(str)
```

Network

```
# Initialize the graph
G = nx.Graph()
```

Artists and their freinds network

In the relationships dataframe, friends column has urls. That's why I firstly trying to find the exact url's name from the artist column. In the freinds column, there are multiple values seperated by commas and I have to seperate the column and then doing naming. Also, there are

```
more rows approximately 3000. I reduce this node number for better showing the network in colab small interface.
# Add each title as a single node, ensuring no accidental splits occur
for title in artists_red['title']:
    # print("Adding node:", title) # Check output to confirm full title
    G.add_node(title, type='artist')
print("Total nodes after addition:", len(G.nodes))

→ Total nodes after addition: 80
print(G.nodes)
🚁 ['Ancient Egypt', 'Ancient Greek Pottery', 'Ancient Greek Painting and Sculpture', 'Apelles', 'Fayum portrait', 'Orthodox Icons', 'Wu Da
for _, row in relationships_art.iterrows():
    artist = row['artist_name']
    # Check if 'Fr1' column has valid (non-NaN) data
    if pd.notna(row['Fr1']):
        # Split the string into individual friends if comma-separated
        friends = row['Fr1'].split(',')
        for friend in friends:
           G.add_edge(artist, friend.strip(), relationship='friend')
    if pd.notna(row['Fr2']):
        # Split the string into individual friends if comma-separated
        friends = row['Fr2'].split(',')
        for friend in friends:
           G.add_edge(artist, friend.strip(), relationship='friend')
    if pd.notna(row['Fr3']):
        # Split the string into individual friends if comma-separated
        friends = row['Fr3'].split(',')
        for friend in friends:
           G.add_edge(artist, friend.strip(), relationship='friend')
    if pd.notna(row['Fr4']):
        # Split the string into individual friends if comma-separated
        friends = row['Fr4'].split(',')
        for friend in friends:
            G.add_edge(artist, friend.strip(), relationship='friend')
print(G.edges)
🔁 [('Rogier van der Weyden', 'Jacques Daret'), ('Giulio Clovio', 'Pieter Bruegel the Elder'), ('Pieter Bruegel the Elder', 'Maarten de Vos
     4
print("Total edges after addition:", len(G.edges))
→ Total edges after addition: 124
degree_seq = [d for n,d in G.degree()]
aver_deg = sum(degree_seq)/len(degree_seq)
```

```
print("Average_degree: ",aver_deg)

Average_degree: 1.1221719457013575

# art_dic = {}
# inst_dic = {}
# school_dic = {}
# or node in G.nodes:
# art_dic[node[0]]=node[1]
# inst_dic[node[0]]=node[2]
# school_dic[node[0]]=node[3]

density = nx.density(G)
density

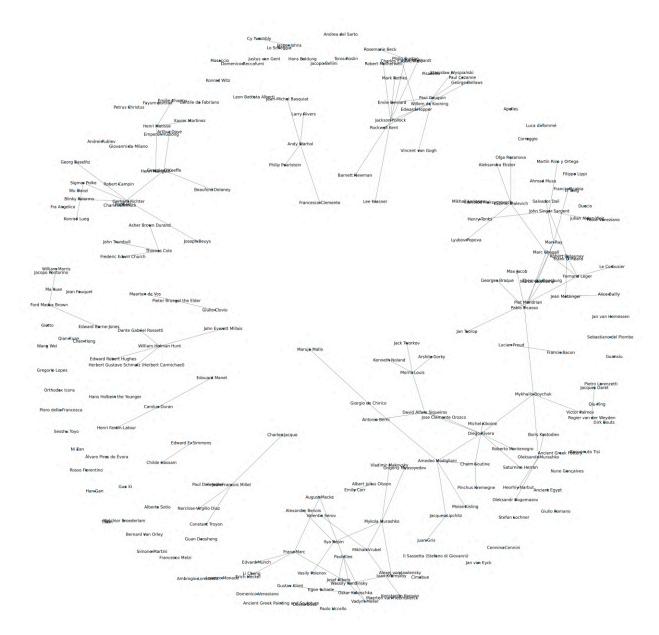
Provided in the school of the
```

nx.draw(G, pos, node_size=180, with_labels=True, font_size=30)

plt.savefig("network_visualization_high_res.png", format="PNG", dpi=300)

plt.title("Basic Network Visualization")

plt.show()



The goal is to answer the following questions:

- 1. Which were the most influential artists?
- 2. Which were the most influential movements?
- 3. Which were the most influential institutions?
- 4. Which nationalities concentrate the majority of artists?
- 5. Which are the biggest communities in the network?

```
# Initialize the graph again
G1 = nx.Graph()
# Add artist nodes
for _, row in artists.iterrows():
    G1.add_node(row['title'], type='artist', nation=row['nation'], active_years=row['year'], artworks=row['totalWorksTitle'])
# Add institution nodes
for _, row in institutions.iterrows():
    G1.add_node(row['url'], type='institution', country=row['country'], city=row['city'])
# Add school nodes
for _, row in schools.iterrows():
    G1.add_node(row['url'], type='school')
# Add nation nodes
for _, row in artists.iterrows():
    G1.add_node(row['nation'], type='nation')
#add movements nodes
for _, row in relationships.iterrows():
    G1.add_node(row['movements'], type='movement')
print("Total nodes after addition:", len(G1.nodes))
→ Total nodes after addition: 3401
print(G1.nodes)
🚁 ['Ancient Egypt', 'Ancient Greek Pottery', 'Ancient Greek Painting and Sculpture', 'Apelles', 'Fayum portrait', 'Orthodox Icons', 'Wu Da
# Add edges based on relationships
for _, row in relationships_art.iterrows():
    artist = row['artist_name']
    if pd.notna(row['friends']):
        for friend in row['friends'].split(','):
           G1.add_edge(artist, friend.strip(), relationship='friend')
    if pd.notna(row['influenced_by']):
        for influencer in row['influenced_by'].split(','):
            G1.add_edge(artist, influencer.strip(), relationship='influenced_by'
    if pd.notna(row['influenced_on']):
        for influence in row['influenced_on'].split(','):
           G1.add_edge(artist, influence.strip(), relationship='influences_on')
    if pd.notna(row['institution']):
        for institution in row['institution'].split(','):
            G1.add_edge(artist, institution.strip(), relationship='studied_at')
    if pd.notna(row['school']):
        for institution in row['school'].split(','):
            G1.add_edge(artist, institution.strip(), relationship='studied_at')
    if pd.notna(row['movements']):
        for mov in row['movements'].split(','):
           G1.add_edge(artist, mov.strip(), relationship='movement')
#add nation edges
for _, row in artists.iterrows():
    artist = row['title']
```

```
if pd.notna(row['nation']):
        for nat in row['nation'].split(','):
           G1.add edge(artist, nat.strip(), relationship='nation')
print("Total edges after addition:", len(G1.edges))
→ Total edges after addition: 3973
print(G1.edges)
🚁 [('Ancient Egypt', 'Egyptians'), ('Ancient Greek Pottery', 'Greeks'), ('Ancient Greek Painting and Sculpture', 'Greeks'), ('Apelles'
degree_seq = [d for n,d in G1.degree()]
aver_deg = sum(degree_seq)/len(degree_seq)
print("Average_degree: ",aver_deg)
Average_degree: 2.08392341987936
density = nx.density(G1)
density
3. 0.0005466745592548164
# Remove nodes of a certain type, e.g., nodes without edges (isolated)
isolated_nodes = [node for node, degree in G1.degree() if degree == 0]
G1.remove_nodes_from(isolated_nodes)
# Visualization using networkx
plt.figure(figsize=(80, 80))
pos = nx.spring_layout(G1, k=.1)
nx.draw(G1, pos, node_size=180, with_labels=True, font_size=30)
# plt.savefig("network_visualization_new.png", format="PNG", dpi=300)
plt.title("Network Visualization")
plt.show()
```



