

```

!git clone https://github.com/pbloom/kgbench-loader.git
%cd kgbench-loader
!pip install .

→ Downloading nvidia_cusparse_cu12-12.3.1.170-py3-none-manylinux2014_x86_64.whl (207.5 MB)
    207.5/207.5 MB 5.8 MB/s eta 0:00:00
→ Downloading nvidia_nvjitlink_cu12-12.4.127-py3-none-manylinux2014_x86_64.whl (21.1 MB)
    21.1/21.1 MB 81.7 MB/s eta 0:00:00
Building wheels for collected packages: kgbench, fire, wget
  Building wheel for kgbench (setup.py) ... done
    Created wheel for kgbench: filename=kgbench-0.2-py3-none-any.whl size=14130 sha256=cce60d3d3376ae5a4ba64ce436c7e48ca7c9c315c93b5a88b072196d34a2d922
  Stored in directory: /root/.cache/pip/wheels/0a/19/59/262e8e140b9b0256a1c0d914f1d588d08ed7e7f9714a9d647a
  Building wheel for fire (setup.py) ... done
    Created wheel for fire: filename=fire-0.7.0-py3-none-any.whl size=114249 sha256=c0460ee9f772ed329ca4dec5e324f757b01336e33f0fd90c4575e96d3cc42c5
  Stored in directory: /root/.cache/pip/wheels/46/54/24/1624fd5b8674eb1188623f7e8e17cf7c0f6c24b609dfb8a89
  Building wheel for wget (setup.py) ... done
    Created wheel for wget: filename=wget-3.2-py3-none-any.whl size=9656 sha256=e9fbe72c75ccc23e9dabe0bdf618ea975237ffd7e0c1ecc3819f2eba0abcd6a2
  Stored in directory: /root/.cache/pip/wheels/40/b3/0f/a40dbd1c6861731779f62cc4babcb234387e11d697df70ee97
Successfully built kgbench fire wget
Installing collected packages: wget, rdflib, nvidia-nvjitlink-cu12, nvidia-curand-cu12, nvidia-cufft-cu12, nvidia-cuda-runtime-cu12, nvidia-cuda-nvrtc-cu12, nvidia-cuda-cupti-cu12, nvidia-cublas-cu12,
  Attempting uninstall: nvidia-nvjitlink-cu12
    Found existing installation: nvidia-nvjitlink-cu12 12.5.82
    Uninstalling nvidia-nvjitlink-cu12-12.5.82:
      Successfully uninstalled nvidia-nvjitlink-cu12-12.5.82
  Attempting uninstall: nvidia-curand-cu12
    Found existing installation: nvidia-curand-cu12 10.3.6.82
    Uninstalling nvidia-curand-cu12-10.3.6.82:
      Successfully uninstalled nvidia-curand-cu12-10.3.6.82
  Attempting uninstall: nvidia-cufft-cu12
    Found existing installation: nvidia-cufft-cu12 11.2.3.61
    Uninstalling nvidia-cufft-cu12-11.2.3.61:
      Successfully uninstalled nvidia-cufft-cu12-11.2.3.61
  Attempting uninstall: nvidia-cuda-runtime-cu12
    Found existing installation: nvidia-cuda-runtime-cu12 12.5.82
    Uninstalling nvidia-cuda-runtime-cu12-12.5.82:
      Successfully uninstalled nvidia-cuda-runtime-cu12-12.5.82
  Attempting uninstall: nvidia-cuda-nvrtc-cu12
    Found existing installation: nvidia-cuda-nvrtc-cu12 12.5.82
    Uninstalling nvidia-cuda-nvrtc-cu12-12.5.82:
      Successfully uninstalled nvidia-cuda-nvrtc-cu12-12.5.82
  Attempting uninstall: nvidia-cuda-cupti-cu12
    Found existing installation: nvidia-cuda-cupti-cu12 12.5.82
    Uninstalling nvidia-cuda-cupti-cu12-12.5.82:
      Successfully uninstalled nvidia-cuda-cupti-cu12-12.5.82
  Attempting uninstall: nvidia-cublas-cu12
    Found existing installation: nvidia-cublas-cu12 12.5.3.2
    Uninstalling nvidia-cublas-cu12-12.5.3.2:
      Successfully uninstalled nvidia-cublas-cu12-12.5.3.2
  Attempting uninstall: nvidia-cusparse-cu12
    Found existing installation: nvidia-cusparse-cu12 12.5.1.3
    Uninstalling nvidia-cusparse-cu12-12.5.1.3:
      Successfully uninstalled nvidia-cusparse-cu12-12.5.1.3
  Attempting uninstall: nvidia-cudnn-cu12
    Found existing installation: nvidia-cudnn-cu12 9.3.0.75
    Uninstalling nvidia-cudnn-cu12-9.3.0.75:
      Successfully uninstalled nvidia-cudnn-cu12-9.3.0.75
  Attempting uninstall: nvidia-cusolver-cu12
    Found existing installation: nvidia-cusolver-cu12 11.6.3.83
    Uninstalling nvidia-cusolver-cu12-11.6.3.83:
      Successfully uninstalled nvidia-cusolver-cu12-11.6.3.83
Successfully installed fire-0.7.0 kgbench-0.2 nvidia-cublas-cu12-12.4.5.8 nvidia-cuda-cupti-cu12-12.4.127 nvidia-cuda-nvrtc-cu12-12.4.127 nvidia-cuda-runtime-cu12-12.4.127 nvidia-cudnn-cu12-9.1.0.70 nv

```

```

from kgbench import load
dataset = load('dmg777k',torch=True)

→ Downloading dmg777k dataset.
loaded data dmg777k (64.19s).

```

```

print(f"Dataset type: {type(dataset)}")
print(f"Available attributes: {dir(dataset)}")

```

```

→ Dataset type: <class 'kgbench.load.Data'>
Available attributes: ['__class__', '__delattr__', '__dict__', '__dir__', '__doc__', '__eq__', '__format__', '__ge__', '__getattribute__', '__getstate__', '__gt__', '__hash__', '__init__', '__init_subclass__', '__reduce__', '__reduce_ex__', '__repr__', '__setattr__', '__sizeof__', '__str__', '__subclasshook__']
<----->

# Import necessary libraries
import kgbench
import torch
import networkx as nx
import matplotlib.pyplot as plt
from PIL import Image

# Load dataset (Change 'dmg777k' to your dataset if needed)

# Print basic dataset info
print(f"Dataset Name: {dataset.name}")
print(f"Number of Entities (Nodes): {dataset.num_entities}")
print(f"Number of Relations (Edges): {dataset.num_relations}")
print(f"Number of Triples: {len(dataset.triples)}\n")

# Check Available Attributes
print("Dataset Attributes:")
print(dir(dataset))

# Explore Triples (Knowledge Graph structure)
print("\nSample Triples (subject, predicate, object):")
for triple in dataset.triples[:5]:
    print(triple)

# Entity-to-Index and Index-to-Entity mappings
print("\nEntity-to-Index Mapping (First 5):", list(dataset.e2i.items())[:5])
print("Index-to-Entity Mapping (First 5):", list(dataset.i2e)[:5])
print("Index-to-Relation Mapping (First 5):", list(dataset.i2r)[:5])

```

```

# Check available data types
print("\nAvailable Data Types:", dataset.datatypes())

# Check if there are images or text
if "image" in dataset.datatypes():
    print("Dataset contains image data.")
    img = dataset.get_images(0)
    if img is not None:
        plt.imshow(img)
        plt.axis('off')
        plt.title("Sample Image from Dataset")
        plt.show()
    else:
        print("No image available for this node.")

if "text" in dataset.datatypes():
    print("\nSample text from dataset:", dataset.get_strings(0))

# Convert dataset to PyG format (for Graph Neural Networks)
try:
    graph_pyg = dataset.pyg()
    print("\nPyG Graph Structure:")
    print(graph_pyg)
except Exception as e:
    print("\nPyG Conversion Error:", e)

# Convert dataset to DGL format (for DGL-based GNNs)
try:
    graph_dgl = dataset.dgl()
    print("\nDGL Graph Structure:")
    print(graph_dgl)
except Exception as e:
    print("\nDGL Conversion Error:", e)

# Visualize a small subgraph using Network

```

→ Dataset Name: dm777k
Number of Entities (Nodes): 341270
Number of Relations (Edges): 60
Number of Triples: 777124

Dataset Attributes:
['__class__', '__delattr__', '__dict__', '__dir__', '__doc__', '__eq__', '__format__', '__ge__', '__getattribute__', '__getstate__', '__gt__', '__hash__', '__init__', '__init_subclass__', '__le__', '__lt__']

Sample Triples (subject, predicate, object):
tensor([130685, 28, 54795])
tensor([130685, 31, 201822])
tensor([130690, 28, 58948])
tensor([130690, 31, 201822])
tensor([130691, 28, 63024])

Entity-to-Index Mapping (First 5): [((0, ('', 'none')), 341269), ((1, ('"\t Hof Alle Morgen Nieuwe Zorgen". Bakstenen woonhuis met pannen zadeldak. Tuittopgevels met vlechtingen en jaarrankers 1796. Bakgoot op klossen. DichtIndex-to-Entity Mapping (First 5): [('', 'none'), ('\t Hof Alle Morgen Nieuwe Zorgen". Bakstenen woonhuis met pannen zadeldak. Tuittopgevels met vlechtingen en jaarrankers 1796. Bakgoot op klossen. DichtIndex-to-Relation Mapping (First 5): ['<http://data.pdok.nl/def/pdok#asWKT-RD>', '<http://dbpedia.org/ontology/city>', '<http://dbpedia.org/ontology/codeNationalMonument>', '<http://dbpedia.org/ontology/location>']

Available Data Types: ['iri', 'none', '@es', '@fy', '@nl', '@nl-nl', '@pt', '@ru', '<http://kgbench.info/dt#base64Image>', '<http://www.opengis.net/ont/geosparql#wktLiteral>', '<http://www.w3.org/2001/XMLSchema>']

PyG Conversion Error: Pytorch geometric does not appear to be installed. Try `pip install pyg` to install it.

DGL Conversion Error: DGL does not appear to be installed. Try `pip install dgl` to install it. See <http://dgl.ai> for more information.

< ━━━━ >

```

len(dataset.triples)

```

→ 777124

```

import matplotlib.pyplot as plt
import random

# Get the images (you may adjust this based on how your dataset returns them)
images = dataset.get_images()

# Randomly select 20 indices (ensure there are enough images)
num_images_to_show = 10
random_indices = random.sample(range(len(images)), min(num_images_to_show, len(images)))

# Display the selected random images
plt.figure(figsize=(15, 15)) # Increase the figure size to make room for multiple images
for i, idx in enumerate(random_indices):
    plt.subplot(4, 5, i+1) # Create a 4x5 grid (4 rows, 5 columns)
    plt.imshow(images[idx])
    plt.axis('off') # Hide axes for better visualization
    plt.title(f"Image {idx+1}")

plt.show()

```

→

Image 43909



Image 8867



Image 12688



Image 9549



Image 13546



Image 36626



Image 43058



Image 39909



Image 31840



Image 6571



```
print(len(images))
```

→ 46061

```
import networkx as nx
import matplotlib.pyplot as plt

# Create a graph
G = nx.Graph()
n=5
# Add edges from dataset (taking a subset for visualization)
for (s, p, o) in dataset.triples[:n]: # First 50 triples
    G.add_edge(dataset.i2e[s], dataset.i2e[o], label=dataset.i2r[p])

# Use a better layout
pos = nx.spring_layout(G, seed=42) # For better spacing

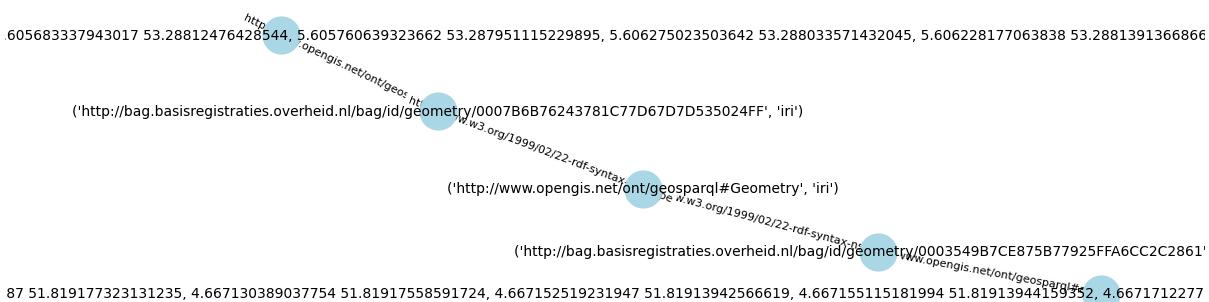
# Plot the graph with enhancements
plt.figure(figsize=(12, 8))
nx.draw(G, pos, with_labels=True, node_size=700, font_size=10, edge_color="gray", node_color="lightblue")

# Draw edge labels
edge_labels = {(dataset.i2e[s], dataset.i2e[o]): dataset.i2r[p] for (s, p, o) in dataset.triples[:n]}
nx.draw_networkx_edge_labels(G, pos, edge_labels=edge_labels, font_size=8)

plt.title("Graph Visualization", fontsize=14)
plt.show()
```

→

Graph Visualization (First 50 Edges)



#asWKT
ties.overheid.nl/bag/id/geometry/000B23D2133D5AE6A0CA771ECB2368B9', 'iri')
337608214358, 6.71417723756294 52.75847847165101, 6.714193425658023 52.75847733856224, 6.714214102235433 52.758586777916214, 6.7140030623806455 52.
http://www.

```

unique_relations = set() # Store unique predicates

for _, p, _ in dataset.triples: # Loop through triples
    unique_relations.add(dataset.i2r[p]) # Convert index to relation string

# Print all unique relationships
print("Unique Relationships (Predicates) in the Graph:")
for rel in sorted(unique_relations): # Sorting for easier readability
    print(rel)

# Count of unique relations
print("\nTotal Unique Relationships:", len(unique_relations))

```

→ Unique Relationships (Predicates) in the Graph:

```

http://data.pdok.nl/def/pdok#asWKT-RD
http://dbpedia.org/ontology/city
http://dbpedia.org/ontology/codeNationalMonument
http://dbpedia.org/ontology/location
http://dbpedia.org/ontology/name
http://dbpedia.org/ontology/neighbourhood
http://dbpedia.org/ontology/thumbail
http://purl.org/dc/terms/created
http://purl.org/dc/terms/creator
http://purl.org/dc/terms/description
http://purl.org/dc/terms/isPartOf
http://purl.org/dc/terms/spatial
http://schema.org/dateCreated
http://schema.org/roleName
http://www.geonames.org/ontology#alternateName
http://www.geonames.org/ontology#countryCode
http://www.geonames.org/ontology#featureClass
http://www.geonames.org/ontology#featureCode
http://www.geonames.org/ontology#locationMap
http://www.geonames.org/ontology#name
http://www.geonames.org/ontology#nearbyFeatures
http://www.geonames.org/ontology#officialName
http://www.geonames.org/ontology#parentADM1
http://www.geonames.org/ontology#parentADM2
http://www.geonames.org/ontology#parentCountry
http://www.geonames.org/ontology#parentFeature
http://www.geonames.org/ontology#population
http://www.geonames.org/ontology#wikipediaArticle
http://www.opengis.net/ont/geosparql#asWKT
http://www.opengis.net/ont/geosparql#hasGeometry
http://www.opengis.net/ont/geosparql#sfWithin
http://www.w3.org/1999/02/22-rdf-syntax-ns#type
http://www.w3.org/2000/01/rdf-schema#isDefinedBy
http://www.w3.org/2000/01/rdf-schema#label
http://www.w3.org/2000/01/rdf-schema#seeAlso
http://www.w3.org/2002/07/owl#sameAs
http://www.w3.org/2003/01/geo/wgs84_pos#alt
http://www.w3.org/2006/vcard/ns#hasStreetAddress
http://www.w3.org/2006/vcard/ns#postal-code
http://www.w3.org/2006/vcard/ns#street-address
http://xmlns.com/foaf/0.1/depiction
http://xmlns.com/foaf/0.1/depicts
http://xmlns.com/foaf/0.1/name
https://data.labs.pdok.nl/rce/def/bouwjaar
https://data.labs.pdok.nl/rce/def/complexnummer
https://data.labs.pdok.nl/rce/def/fotograaf
https://data.labs.pdok.nl/rce/def/graveur
https://data.labs.pdok.nl/rce/def/huisnummer
https://data.labs.pdok.nl/rce/def/huisnummerCompleet
https://data.labs.pdok.nl/rce/def/huisnummerToevoeging
https://data.labs.pdok.nl/rce/def/internComplexNummers
https://data.labs.pdok.nl/rce/def/isFree
https://data.labs.pdok.nl/rce/def/locator
https://data.labs.pdok.nl/rce/def/ontwerper
https://data.labs.pdok.nl/rce/def/reprofotograaf
https://data.labs.pdok.nl/rce/def/rnasSubject
https://data.labs.pdok.nl/rce/def/schilder

```

Start coding or generate with AI.

→ Number of Unique Nodes: 341270

```

print(dataset.i2e[130685]) # Get subject name
print(dataset.i2r[28]) # Get relation name
print(dataset.i2e[54795]) # Get object name

```

→ ('<http://bag.basisregistraties.overheid.nl/bag/id/geometry/0003549B7CE875B77925FFA6CC2C2861>', 'iri')

'POLYGON ((4.667227191836703 51.81920171648348, 4.667236120685068 51.819203883584755, 4.6672263678506125 51.81920790458488, 4.667215533196375 51.81921064256385, 4.667204024085515 51.81921201912295, 4.667227191836703 51.81920171648348))'

Start coding or generate with AI.

```

# Get the first two nodes
nodes_list = list(G.nodes()) # Convert nodes to a list

if len(nodes_list) >= 2:
    node1, node2 = nodes_list[:2] # Extract first two nodes
    print("First node:", node1)
    print("Second node:", node2)
elif len(nodes_list) == 1:
    print("Only one node present:", nodes_list[0])
else:
    print("Graph has no nodes.")

# Get one example edge with label (if available)
if len(G.edges) > 0:
    example_edge = next(iter(G.edges(data=True)))
    u, v, data = example_edge
    relation_label = data.get("label", "Unknown")
    print(f"Example edge: {u} --{relation_label}--> {v}")
else:

```

```
print("Graph has no edges.")
```

First node: ('<http://bag.basisregistraties.overheid.nl/bag/id/geometry/0003549B7CE87B577925FA66CC2C2861>', 'iri')
Second node: ('POLYGON ((4.66727191836769 51.81920171648384, 4.667236120685068 51.81920383584575, 4.6672263678506125 51.81920790458488, 4.667215533196375 51.81921064256385, 4.667204024085515 51.81921201836769, 4.66727191836769 51.81920171648384))', 'POLYGON ((4.66727191836769 51.81920383584575, 4.6672263678506125 51.81920790458488, 4.667215533196375 51.81921064256385, 4.667204024085515 51.81921201836769, 4.66727191836769 51.81920171648384))')
Example edges: ('<https://bag.basisregistraties.overheid.nl/bag/id/geometry/0003549B7CE87B577925FA66CC2C2861>', 'iri'), ('<https://www.eenopniet.net/cat/geocontrol#GKUT>', 'iri')

Example 8-17. [\(link\)](#)

```
!pip install adjustText
```

→ Collecting adjustText

```
Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (from adjustText) (2.0.2)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-packages (from adjustText) (3.10.0)
Requirement already satisfied: scipy in /usr/local/lib/python3.11/dist-packages (from adjustText) (1.14.1)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (1.3.1)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (0.12.1)
Requirement already satisfied: fonttools<=4.22.0 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (4.56.0)
Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (1.4.8)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (24.2)
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (11.1.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (3.2.1)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.11/dist-packages (from matplotlib->adjustText) (2.8.2)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.7->matplotlib->adjustText) (1.17.0)
Downloading adjustText-1.3.0-py3-none-any.whl (13 kB)
Installing collected packages: adjustText
Successfully installed adjustText-1.3.0
```

```

import networkx as nx
import matplotlib.pyplot as plt
from shapely.wkt import loads # To parse WKT polygon

# Create a directed graph
G = nx.DiGraph()

# Example node identifiers
geometry_node = "Geometry 1"
polygon_node = "Polygon Shape"
edge_label = "asWKT"

# Add nodes with metadata
G.add_node(geometry_node, full_url='http://bag.basisregistraties.overheid.nl/bag/id/geometry/0003549B7CE875B77925FFA6CC2C2861')
G.add_node(polygon_node, full_url='http://www.opengis.net/ont/geosparql#wktLiteral')

# Add edge
G.add_edge(geometry_node, polygon_node, label=edge_label)

# Compute positions using a layout algorithm
pos = nx.spring_layout(G, seed=42)

# Draw graph nodes and edges
plt.figure(figsize=(8, 6))
nx.draw(G, pos, with_labels=True, node_color='lightblue', edge_color='gray', node_size=2000, font_size=10, font_weight="bold")

# Draw edge labels
edge_labels = {(geometry_node, polygon_node): edge_label}
nx.draw_networkx_edge_labels(G, pos, edge_labels=edge_labels, font_size=9)

# POLYGON DATA
wkt_polygon = "POLYGON ((4.667227191836703 51.81920171648348, 4.667236120685068 51.819203883584755, 4.6672263678506125 51.819203883584755, 4.667227191836703 51.81920171648348))"

# Parse the WKT polygon
polygon = loads(wkt_polygon)

# Extract polygon coordinates
polygon_coords = list(polygon.exterior.coords)
x_vals, y_vals = zip(*polygon_coords)

# Position the polygon inside the "Polygon Shape" node
polygon_center = pos[polygon_node]

# Adjust polygon coordinates to fit within the node
scale_factor = 1000 # Adjust this for better fit
x_scaled = [polygon_center[0] + scale_factor * (x - min(x_vals)) for x in x_vals]
y_scaled = [polygon_center[1] + scale_factor * (y - min(y_vals)) for y in y_vals]

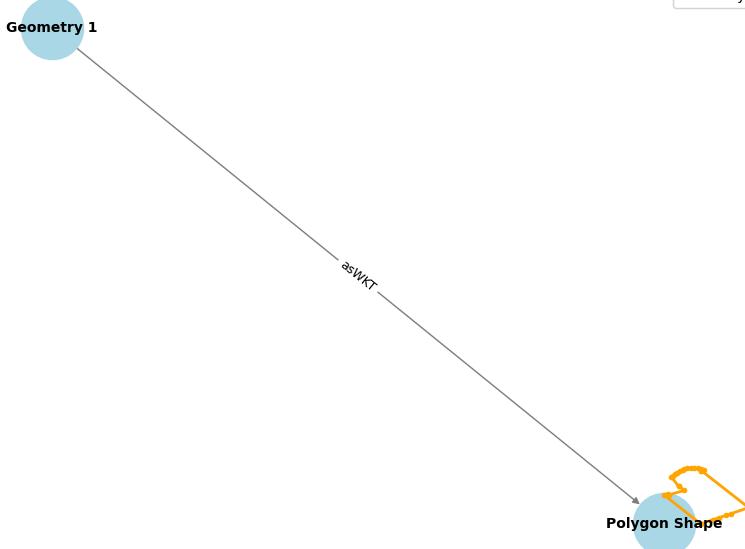
# Plot the polygon INSIDE the node
plt.plot(x_scaled, y_scaled, color='orange', linewidth=2, marker="o", markersize=3, label="Polygon")

# Show title and legend
plt.title("Graph Representation of Geometry Data with Embedded Polygon", fontsize=12)

```

Graph Representation of Geometry Data with Embedded Polygon

Polygon



```

import networkx as nx
import matplotlib.pyplot as plt
from shapely.wkt import loads
from shapely.geometry import Polygon
import textwrap

def shorten_label(label, max_length=50):
    """Shorten long node labels for better visualization."""
    label = str(label)

    return label[:max_length] + "..." if len(label) > max_length else label

# Create a graph
G = nx.Graph()
n = 5 # Number of triples to visualize
polygons = [] # Store polygons separately for plotting

# Add edges and detect polygons
for (s, p, o) in dataset.triples[:n]:
    node1 = dataset.i2e[s]
    node2 = dataset.i2e[o]
    relation = dataset.i2r[p]

    # Handle WKT polygons separately
    if is_wkt_polygon(node2):
        polygons.append(loads(node2)) # Convert WKT to geometry for plotting
        node2 = "[POLYGON]" # Replace with a simple label

    # Add edges with shorter relations (e.g., show "asWKT" instead of long URIs)
    relation = "asWKT" if relation == "http://www.opengis.net/ont/geosparql#asWKT" else relation
    G.add_edge(node1, node2, label=relation)

# Use a better layout
pos = nx.spring_layout(G, seed=42)

# Create figure
fig, ax = plt.subplots(figsize=(12, 8))

# Draw the graph
nx.draw(G, pos, with_labels=True, node_size=1000, font_size=10,
        edge_color="gray", node_color="lightblue", ax=ax)

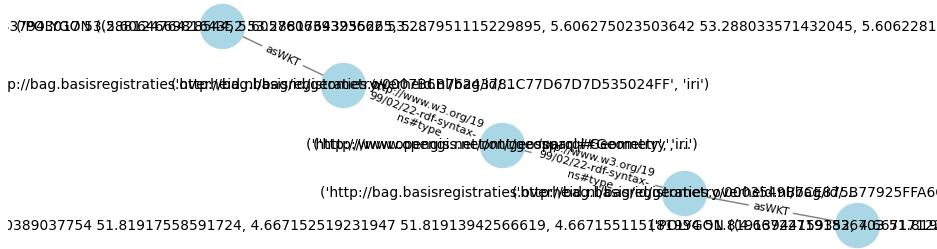
# Format edge labels
edge_labels = {(u, v): textwrap.fill(data["label"], width=20) for u, v, data in G.edges(data=True)}
nx.draw_networkx_edge_labels(G, pos, edge_labels=edge_labels, font_size=8, ax=ax)

# Shorten node labels for clarity
node_labels = {node: shorten_label(node) for node in G.nodes()}
nx.draw_networkx_labels(G, pos, labels=node_labels, font_size=10, ax=ax)

# Plot the polygons separately
for poly in polygons:
    x, y = poly.exterior.xy # Extract polygon coordinates
    ax.fill(x, y, alpha=0.4, fc="lightgreen", edgecolor="green")

# Set title
plt.title("Graph Visualization with WKT Polygons", fontsize=14)

# Show the plot
plt.show()
  
```



http://diag1.wiki.dreamhosters.com/00023D1bagD0AE6A0CA771ECB2368B9?_iri

```

import networkx as nx
import matplotlib.pyplot as plt
from shapely.wkt import loads
from shapely.geometry import Polygon
import textwrap

# Function to check if a node contains a WKT POLYGON
def is_wkt_polygon(node):
    return isinstance(node, str) and node.startswith("POLYGON")

# Function to shorten long URIs or WKT literals
def shorten_label(label, max_length=10):
    label = str(label)
    if is_wkt_polygon(label):
        return "[POLYGON]" # Replace polygons with a generic label
    label=label[:max_length] + "..." if len(label) > max_length else label
    return textwrap.fill(label, width=10) # Wrap text for better visualization

# Create a graph
G = nx.Graph()
n = 100 # Limit the number of triples to visualize
polygons = [] # Store polygons separately for plotting

# Add edges and detect polygons
for (s, p, o) in dataset.triples[:n]:
    node1 = dataset.i2e[s]
    node2 = dataset.i2e[o]
    relation = dataset.i2r[p]

    # Rename relation if it's "asWKT"
    G.add_edge(node1, node2, label=relation)

# Use a better layout
pos = nx.kamada_kawai_layout(G)

# Create figure
fig, ax = plt.subplots(figsize=(100, 100))

# Draw the graph
nx.draw(G, pos, with_labels=True, node_size=1500, font_size=10,
        edge_color="gray", node_color="skyblue", ax=ax)

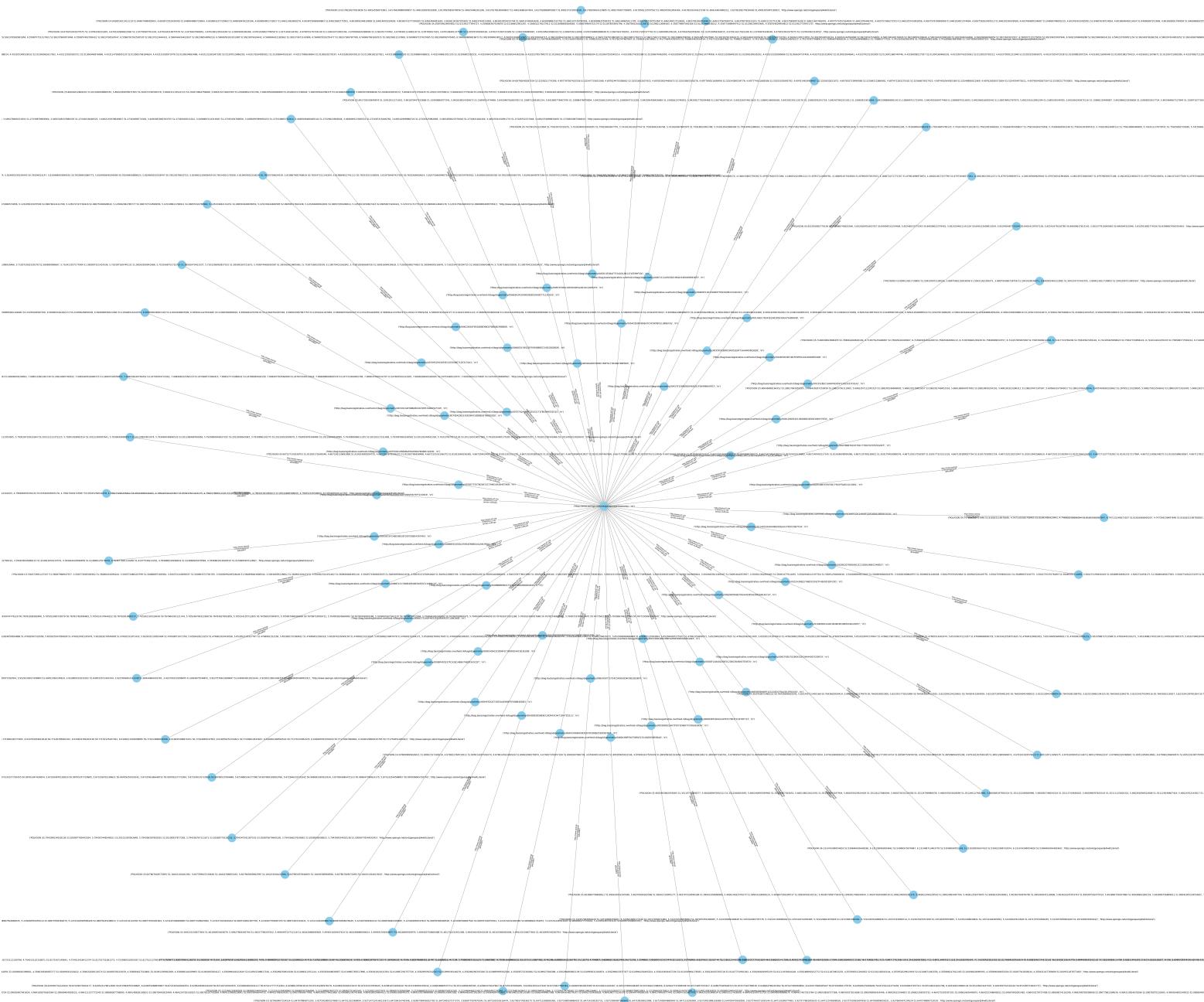
# Format edge labels with better wrapping
edge_labels = {(u, v): textwrap.fill(data["label"], width=17) for u, v, data in
nx.draw_networkx_edge_labels(G, pos, edge_labels=edge_labels, font_size=8, ax=ax)

# Plot the polygons separately
for poly in polygons:
    x, y = poly.exterior.xy # Extract polygon coordinates
    ax.fill(x, y, alpha=0.4, fc="lightgreen", edgecolor="green", linewidth=1)

# Set title
plt.title("Graph Visualization with WKT Polygons", fontsize=14)
plt.savefig("graph_visualization.png", dpi=300, bbox_inches="tight") # Saving

# Show the plot
plt.show()

```



Start coding or generate with AI.

→<Figure size 640x480 with 0 Axes>

```
for u, v, data in G.edges(data=True):
    print(v)
```

Entity Frequency Analysis

```

from collections import Counter

# Extract entities (converting tensor to integer)
entities = [int(triple[0]) for triple in dataset.triples] + [int(triple[2]) for triple in dataset.triples]

# Count occurrences
entity_counts = Counter(entities)

# Display actual most frequent entities
print("Most common entities:", entity_counts.most_common(10))

# Check unique and rare entities again
rare_entities = [e for e, c in entity_counts.items() if c == 1]
print(f"Number of rare entities (appear once): {len(rare_entities)}")


```

Start coding or generate with AT

```
import numpy as np
sorted_entities = entity_counts.most_common()

# Compute cumulative sum
frequencies = np.array([count for _, count in sorted_entities])
cumulative_freq = np.cumsum(frequencies) / sum(frequencies) # Normalize to [0,1]

# Find 80% cutoff point
cutoff_index = np.searchsorted(cumulative_freq, 0.8)
top_entities = sorted_entities[:cutoff_index]

print(f"Number of entities covering 80% of occurrences: {len(top_entities)}")
print("Example top entities:", top_entities[:10])

→ Number of entities covering 80% of occurrences: 68865
```

```
# Get the top 10 most common entity indices
```

```
# Print entity names using i2e mapping
for entity_id, count in top_entities[:10]:
    entity_name = dataset.i2e[entity_id] # Convert ID to actual entity
    print(f"Entity: {entity_name} (ID: {entity_id}) appears {count} times.")

→ Entity: ('http://www.opengis.net/ont/geosparql#Geometry', 'iri') (ID: 201822) appears 65576 times.
Entity: ('https://data.labs.pdk.nl/rce/def/Afheeling', 'iri') (ID: 211084) appears 60996 times.
Entity: ('zwart wit negatief', '@nl-nl') (ID: 341077) appears 23232 times.
Entity: ('Fotocollectie', '@nl-nl') (ID: 28996) appears 18146 times.
Entity: ('Collectie gebouwd', '@nl-nl') (ID: 23569) appears 16868 times.
Entity: ('Dukker, G.J.', 'none') (ID: 25943) appears 13544 times.
Entity: ('Wal, A.J. van der', 'none') (ID: 79498) appears 9397 times.
Entity: ('true', 'http://www.w3.org/2001/XMLSchema#boolean') (ID: 337400) appears 8230 times.
Entity: ('Agrarisch Erfgoed - Fotocollectie SHBO', '@nl-nl') (ID: 19920) appears 8090 times.
Entity: ('Digitale opname', '@nl-nl') (ID: 25196) appears 6023 times.
```

```

# Get statistics
total_entities = len(entity_counts)
rare_entity_count = len(rare_entities)
rare_entity_ratio = rare_entity_count / total_entities

print(f"Total Entities: {total_entities}")
print(f"Rare Entities (appear once): {rare_entity_count}")
print(f"Rare Entity Ratio: {rare_entity_ratio:.2%}")

→ Total Entities: 341270
Rare Entities (appear once): 250658
Rare Entity Ratio: 73.45%

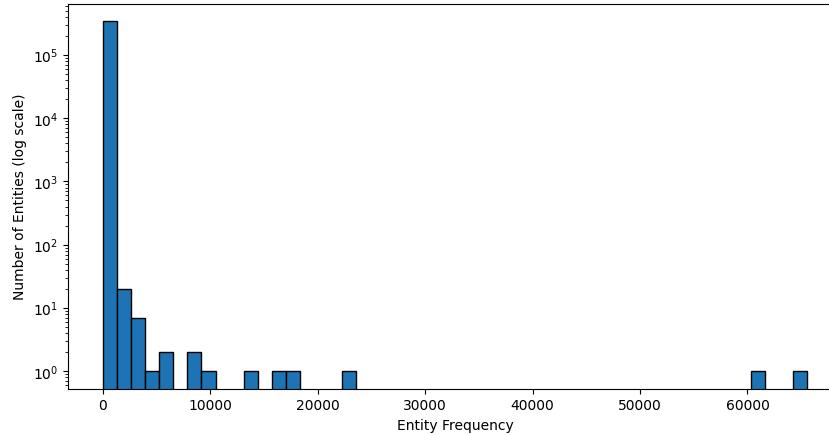
```

```
import matplotlib.pyplot as plt
```

```
# Get frequencies of all entities
entity_frequencies = list(entity_counts.values())
```

```
# Plot histogram (log-scale for better visualization)
plt.figure(figsize=(10, 5))
plt.hist(entity_frequencies, bins=50, log=True, edgecolor="black")
plt.xlabel("Entity Frequency")
plt.ylabel("Number of Entities (log scale)")
plt.title("Entity Frequency Distribution")
plt.show()
```

Entity Frequency Distribution



```
import networkx as nx
```

```
# Create a graph where entities are nodes and relations are edges
G = nx.Graph()
G.add_edges_from([(t[0], t[2]) for t in dataset.triples])
```

```
# Compute connectivity statistics
num_components = nx.number_connected_components(G)
largest_component = max(nx.connected_components(G), key=len)
largest_component_ratio = len(largest_component) / total_entities
```

```
print(f"Number of Connected Components: {num_components}")
```

```
print(f"Largest Component Size: {len(largest_component)} ({largest_component_ratio:.2%} of all nodes}")
print(f"Average Node Degree: {sum(dict(G.degree()).values()) / len(G.nodes())}")
```

```
→ Number of Connected Components: 777124
Largest Component Size: 2 (0.00% of all nodes)
```

```
# Count relation occurrences
relation_counts = Counter([int(triple[1]) for triple in dataset.triples])
```

```
# Get most common relations
most_common_relations = relation_counts.most_common(10)
```

```
least_common_relations = sorted(relation_counts.items(), key=lambda x: x[1])[:10]
```

```
# Display results
print("Most common relations:")
for relation_id, count in most_common_relations:
    print(f"Relation: {dataset.i2r[relation_id]} (ID: {relation_id}) appears {count} times.")
```

```
print("\nLeast common relations:")
for relation_id, count in least_common_relations:
    print(f"Relation: {dataset.i2r[relation_id]} (ID: {relation_id}) appears {count} times.")
```

```
→ Most common relations:
Relation: http://www.w3.org/1999/02/22-rdf-syntax-ns#type (ID: 31) appears 130126 times.
Relation: http://purl.org/dc/terms/description (ID: 9) appears 57042 times.
Relation: http://xmlns.com/foaf/0.1.depicts (ID: 41) appears 47872 times.
Relation: http://xmlns.com/foaf/0.1/depiction (ID: 40) appears 47872 times.
Relation: http://purl.org/dc/terms/isPartOf (ID: 10) appears 47384 times.
Relation: https://data.labs.pdok.nl/rce/def/locator (ID: 52) appears 46123 times.
Relation: http://dbpedia.org/ontology/thumbnail (ID: 6) appears 46108 times.
Relation: http://purl.org/dc/terms/creator (ID: 8) appears 45111 times.
Relation: http://purl.org/dc/terms/created (ID: 7) appears 44216 times.
Relation: https://data.labs.pdok.nl/rce/def/fotograaf (ID: 45) appears 43964 times.
```

```
Least common relations:
Relation: https://data.labs.pdok.nl/rce/def/graveur (ID: 46) appears 1 times.
Relation: http://www.w3.org/2003/01/geo/wgs84\_pos#alt (ID: 36) appears 3 times.
```

```

Relation: https://data.labs.pdok.nl/rce/def/ontwerper (ID: 53) appears 7 times.
Relation: https://data.labs.pdok.nl/rce/def/reprofotograaf (ID: 54) appears 8 times.
Relation: https://data.labs.pdok.nl/rce/def/schijder (ID: 56) appears 36 times.
Relation: http://www.geonames.org/ontology#officialName (ID: 21) appears 55 times.
Relation: http://dbpedia.org/ontology/location (ID: 3) appears 156 times.
Relation: https://data.labs.pdok.nl/rce/def/huisnummerToevoeging (ID: 49) appears 212 times.
Relation: http://dbpedia.org/ontology/neighbourhood (ID: 5) appears 258 times.
Relation: http://purl.org/dc/terms/spatial (ID: 11) appears 260 times.

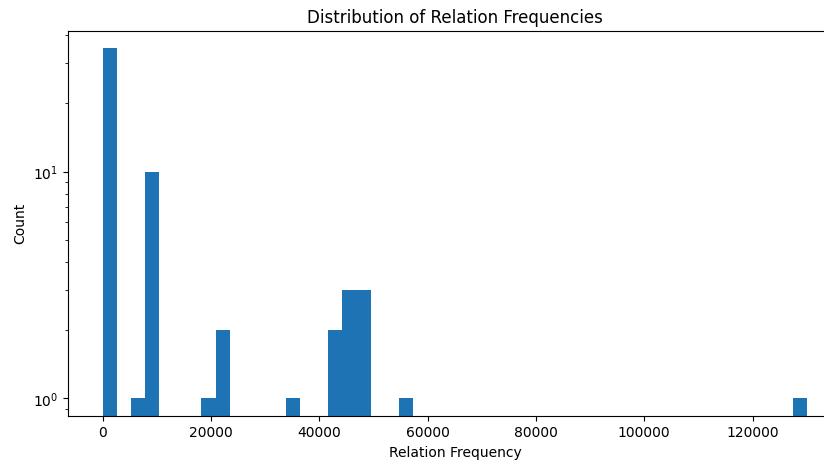
```

```

# Get relation frequencies
relation_frequencies = list(relation_counts.values())

# Plot histogram
plt.figure(figsize=(10,5))
plt.hist(relation_frequencies, bins=50, log=True) # Log scale for better visualization
plt.xlabel("Relation Frequency")
plt.ylabel("Count")
plt.title("Distribution of Relation Frequencies")
plt.show()

```



resembles a Zipfian distribution

```

relation_usage = {int(relation): set() for relation in relation_counts.keys()} # Ensure keys are integers

for head, rel, tail in dataset.triples:
    rel = int(rel) # Convert tensor to integer
    head = int(head)
    tail = int(tail)

    relation_usage[rel].add(head)
    relation_usage[rel].add(tail)

relation_connectivity = {rel: len(entities) for rel, entities in relation_usage.items()}
sorted_connectivity = sorted(relation_connectivity.items(), key=lambda x: x[1], reverse=True)

# Print top 10 most connected relations
print("Top 10 most connected relations:")
for rel, count in sorted_connectivity[:10]:
    print(f"Relation: {dataset.i2r[rel]} connects {count} unique entities")

```

→ Top 10 most connected relations:

```

Relation: http://www.w3.org/1999/02/22-rdf-syntax-ns#type connects 129374 unique entities
Relation: https://data.labs.pdok.nl/rce/def/locator connects 92214 unique entities
Relation: http://dbpedia.org/ontology/thumbnail connects 92169 unique entities
Relation: http://purl.org/dc/terms/description connects 89927 unique entities
Relation: http://xmlns.com/foaf/0.1.depicts connects 56268 unique entities
Relation: http://xmlns.com/foaf/0.1/depiction connects 56268 unique entities
Relation: http://purl.org/dc/terms/isPartOf connects 47422 unique entities
Relation: http://purl.org/dc/terms/creator connects 45738 unique entities
Relation: http://purl.org/dc/terms/created connects 45298 unique entities
Relation: https://data.labs.pdok.nl/rce/def/fotograaf connects 44428 unique entities

```

```

rare_relations = [rel for rel, count in relation_counts.items() if count == 1]
print(f"Number of relations that appear only once: {len(rare_relations)}")
print(f"Percentage of sparse relations: {len(rare_relations) / len(relation_counts) * 100:.2f}%")

```

→ Number of relations that appear only once: 1
Percentage of sparse relations: 1.67%

```

from collections import defaultdict
import itertools

entity_relation_map = defaultdict(set)

for head, rel, tail in dataset.triples:
    rel = int(rel) # Convert tensor to integer
    head = int(head)
    tail = int(tail)
    entity_relation_map[head].add(rel)

co_occurrence_counts = Counter()

for relations in entity_relation_map.values():
    for rel1, rel2 in itertools.combinations(relations, 2):
        co_occurrence_counts[(rel1, rel2)] += 1

```

```
# Print most common relation pairs
print("Top 10 most co-occurring relation pairs:")
for (rel1, rel2), count in co_occurrence_counts.most_common(10):
    print(f'Relations: {dataset.i2r[rel1]} & {dataset.i2r[rel2]} appear together {count} times.')

→ Top 10 most co-occurring relation pairs:
Relations: http://purl.org/dc/terms/description & http://xmlns.com/foaf/0.1.depicts appear together 47872 times.
Relations: http://purl.org/dc/terms/description & http://www.w3.org/1999/02/22-rdf-syntax-ns#type appear together 47872 times.
Relations: http://xmlns.com/foaf/0.1.depicts & http://www.w3.org/1999/02/22-rdf-syntax-ns#type appear together 47872 times.
Relations: http://purl.org/dc/terms/description & http://purl.org/dc/terms/isPartOf appear together 47384 times.
Relations: http://purl.org/dc/terms/isPartOf & http://xmlns.com/foaf/0.1.depicts appear together 47384 times.
Relations: http://purl.org/dc/terms/isPartOf & http://www.w3.org/1999/02/22-rdf-syntax-ns#type appear together 47384 times.
Relations: http://purl.org/dc/terms/description & https://data.labs.pdk.nl/rce/def/locator appear together 46123 times.
Relations: http://xmlns.com/foaf/0.1.depicts & https://data.labs.pdk.nl/rce/def/locator appear together 46123 times.
Relations: https://data.labs.pdk.nl/rce/def/locator & http://www.w3.org/1999/02/22-rdf-syntax-ns#type appear together 46123 times.
Relations: http://dbpedia.org/ontology.thumbnail & http://purl.org/dc/terms/description appear together 46108 times.
```

```

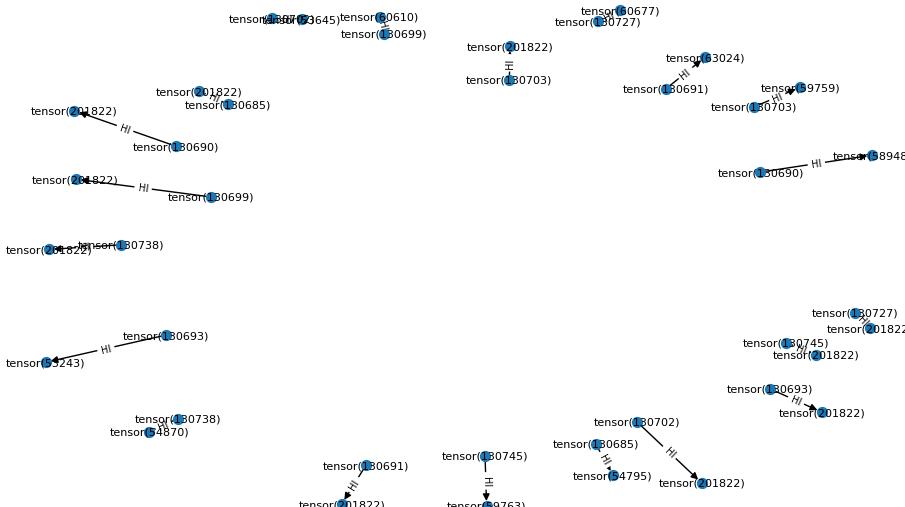
# Create a directed graph
G = nx.DiGraph()
G.add_edges_from([(tuple[0], tuple[2]) for triple in dataset.triples])

# Compute graph properties
print(f"Number of Connected Components: {nx.number_weakly_connected_components(G)}")
print(f"Average Node Degree: {sum(dict(G.degree()).values()) / len(G.nodes())}")

```

```
    Number of Connected Components: 7771
    Average Node Degree: 1.0

# Draw a small subgraph for visualization
plt.figure(figsize=(10,6))
pos = nx.spring_layout(G)
nx.draw(G, pos, with_labels=True, node_size=1000)
edge_labels = { (u, v): "HI" for u, v, d in G.edges}
nx.draw_networkx_edge_labels(G, pos, edge_labels)
```



```
# Convert to a set to remove duplicates
unique_triples = set(dataset.triples)

# Check how many duplicates were present
num_duplicates = len(dataset.triples) - len(unique_triples)
print(f"Number of duplicate triples: {num_duplicates}")

# Number of duplicate triples: 0

type(dataset.triples)

# torch.Tensor

import torch

# Check for missing values in dataset.triples
missing_triples = [triple for triple in dataset.triples if torch.isnan(triple).any() or torch.isinf(triple).any()]
print(f"Number of missing triples: {len(missing_triples)}")
```

```

→ Number of missing triples: 0
-----
RuntimeError                               Traceback (most recent call last)
<ipython-input-55-fd59ec1a98de> in <cell line: 0>()
      4
      5 # If needed, remove them
----> 6 dataset.triples = [triple for triple in dataset.triples if None not in triple]
    7 print(f"Dataset size after removing missing triples: {len(dataset.triples)}")

```

```

----- ▾ 1 frames -----
/usr/local/lib/python3.11/dist-packages/torch/_tensor.py in __contains__(self, element)
 1223     return bool((element == self).any().item()) # type: ignore[union-attr]
 1224
-> 1225     raise RuntimeError(
 1226         f"Tensor.__contains__ only supports Tensor or scalar, but you passed in a {type(element)}." 
 1227     )

```

RuntimeError: Tensor.__contains__ only supports Tensor or scalar, but you passed in a <class 'NoneType'>.

```
type(dataset.triples[1])
```

→ torch.Tensor

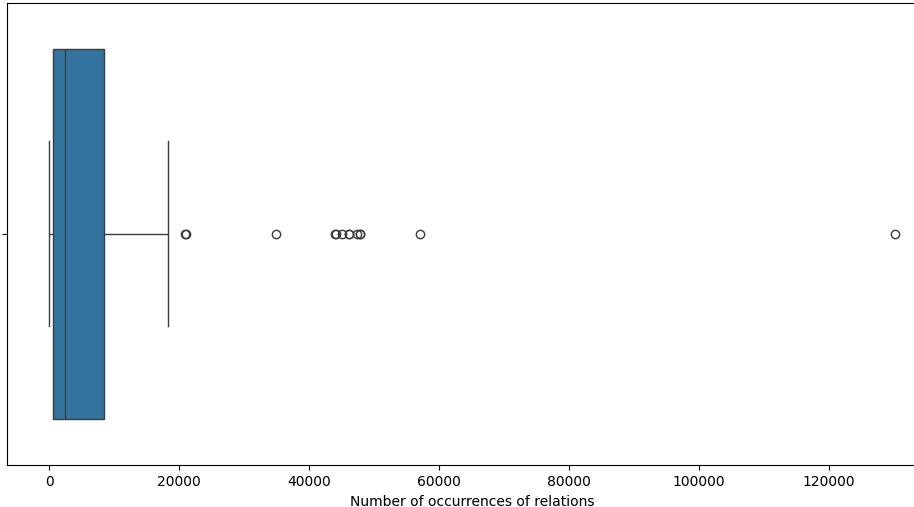
```
# Count occurrences of each relation
relation_counts = Counter(int(triple[1].item()) for triple in dataset.triples)
```

```
# Plot distribution
plt.figure(figsize=(12,6))
sns.boxplot(x=list(relation_counts.values()))
plt.xlabel("Number of occurrences of relations")
plt.title("Outlier Detection in Relation Frequency")
plt.show()
```

```
# Find relations appearing significantly more or less
q1 = np.percentile(list(relation_counts.values()), 25)
q3 = np.percentile(list(relation_counts.values()), 75)
iqr = q3 - q1
```

```
# Define outliers as relations occurring way outside normal range
outlier_threshold = q3 + 1.5 * iqr
outlier_relations = {rel: count for rel, count in relation_counts.items() if count > outlier_threshold}
```

Outlier Detection in Relation Frequency



```

-----
TypeError                                Traceback (most recent call last)
<ipython-input-60-4fd82e9b9687> in <cell line: 0>()
  18 outlier_relations = {rel: count for rel, count in relation_counts.items() if count > outlier_threshold}
  19
---> 20 print(f"Outlier relations (very high frequency): {dataset.i2r[outlier_relations]}")

```

TypeError: list indices must be integers or slices, not dict

```
outlier_keys = list(outlier_relations.keys()) # Extract relation indices
outlier_names = [dataset.i2r[key] for key in outlier_keys] # Convert indices to names

print(f"Outlier relations (very high frequency): {outlier_names}")
```

→ Outlier relations (very high frequency): [<http://www.opengis.net/ont/geosparql#asWKT>, <http://www.w3.org/1999/02/22-rdf-syntax-ns#type>, <http://www.opengis.net/ont/geosparql#hasGeometry>, <http://dbpedia.org/resource/Geography>]

```
outlier_keys = list(outlier_relations.keys()) # Extract relation indices

# Extract only the relation keys from the least common ones
low_freq_relations = [relation for relation, _ in relation_counts.most_common()[-10:]]
```

```
# Convert indices to relation names
outlier_names = [dataset.i2r[key] for key in low_freq_relations]
```

```
print(f"Outlier relations (very low frequency): {outlier_names}")
```

Outlier relations (very low frequency): ['<http://purl.org/dc/terms/spatial>' , '<http://dbpedia.org/ontology/neighbourhood>' , '<https://data.labs.pdok.nl/rce/def/huisnummerToevoeging>' , '<http://dbpedia.org/ont>']

```

11: 260 occurrences
5: 258 occurrences
49: 212 occurrences
3: 156 occurrences
21: 55 occurrences
56: 36 occurrences
54: 8 occurrences
53: 7 occurrences
36: 3 occurrences
46: 1 occurrences

```

```

import networkx as nx

# Create a directed graph
G = nx.DiGraph()

# Add edges with relations as edge attributes
G.add_edges_from([(int(triple[0].item()), int(triple[2].item()), {"relation": triple[1]}) for triple in dataset.triples])

# Find all entities appearing in triples (convert tensors to integers)
all_entities = set(int(triple[0].item()) for triple in dataset.triples) | set(int(triple[2].item()) for triple in dataset.triples)

# Add all entities to the graph explicitly
G.add_nodes_from(all_entities)

# Now find isolated nodes
connected_entities = set(G.nodes())
isolated_nodes = all_entities - connected_entities

print(f"Number of isolated nodes: {len(isolated_nodes)}") # Should now give correct result
if G.number_of_nodes() > 0:
    node_max_in = max(G.in_degree(), key=lambda x: x[1], default=(None, 0)) # Node with max in-degree
    node_max_out = max(G.out_degree(), key=lambda x: x[1], default=(None, 0)) # Node with max out-degree
    node_max_total = max(G.degree(), key=lambda x: x[1], default=(None, 0)) # Node with max total degree
else:
    print("Graph has no nodes.")

-----
```

KeyboardInterrupt

```

<ipython-input-114-d894f41a09fc> in <cell line: 0>()
      5
      6 # Add edges with relations as edge attributes
----> 7 G.add_edges_from([(int(triple[0].item()), int(triple[2].item()), {"relation": triple[1]}) for triple in dataset.triples])
      8
      9 # Find all entities appearing in triples (convert tensors to integers)

<ipython-input-114-d894f41a09fc> in <listcomp>(.0)
      5
      6 # Add edges with relations as edge attributes
----> 7 G.add_edges_from([(int(triple[0].item()), int(triple[2].item()), {"relation": triple[1]}) for triple in dataset.triples])
      8
      9 # Find all entities appearing in triples (convert tensors to integers)

KeyboardInterrupt:
```

```

import networkx as nx
import matplotlib.pyplot as plt

# Create a directed graph
G = nx.DiGraph()

# Add edges with relations as edge attributes
G.add_edges_from([(int(triple[0].item()), int(triple[2].item()), {"relation": triple[1]}) for triple in dataset.triples])

# Compute degrees
in_degrees = [G.in_degree(n) for n in G.nodes()]
out_degrees = [G.out_degree(n) for n in G.nodes()]
total_degrees = [G.degree(n) for n in G.nodes()]

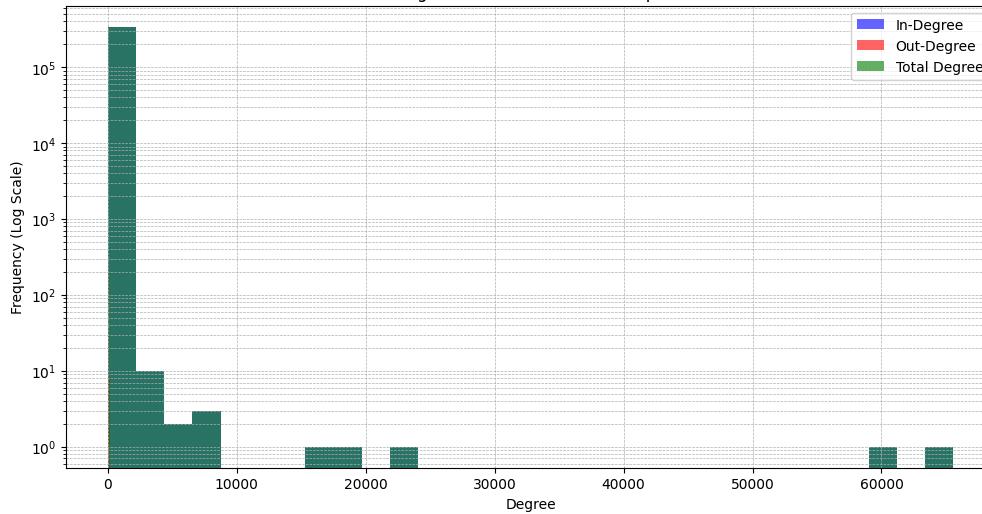
# Plot degree distributions
plt.figure(figsize=(12, 6))

plt.hist(in_degrees, bins=30, alpha=0.6, label="In-Degree", color="blue", log=True)
plt.hist(out_degrees, bins=30, alpha=0.6, label="Out-Degree", color="red", log=True)
plt.hist(total_degrees, bins=30, alpha=0.6, label="Total Degree", color="green", log=True)

plt.xlabel("Degree")
plt.ylabel("Frequency (Log Scale)")
plt.title("Degree Distribution of the Graph")
plt.legend()
plt.grid(True, which="both", linestyle="--", linewidth=0.5)

plt.show()
```

Degree Distribution of the Graph



```

import networkx as nx
import matplotlib.pyplot as plt

# Compute degrees
node_indices = list(G.nodes()) # Get the node indices
in_degrees = [G.in_degree(n) for n in node_indices]
out_degrees = [G.out_degree(n) for n in node_indices]
total_degrees = [G.degree(n) for n in node_indices]

# Plot scatter graph
plt.figure(figsize=(12, 6))

scatter_in = plt.scatter(node_indices, in_degrees, label="In-Degree", color="blue", alpha=0.6)
scatter_out = plt.scatter(node_indices, out_degrees, label="Out-Degree", color="red", alpha=0.6)
scatter_total = plt.scatter(node_indices, total_degrees, label="Total Degree", color="green", alpha=0.6)

# Annotate points where the total degree > 10
for i, (node, degree) in enumerate(zip(node_indices, total_degrees)):
    if degree > 10:
        plt.text(node, degree, f"({node}, {degree})", fontsize=9, color="black", ha="right")

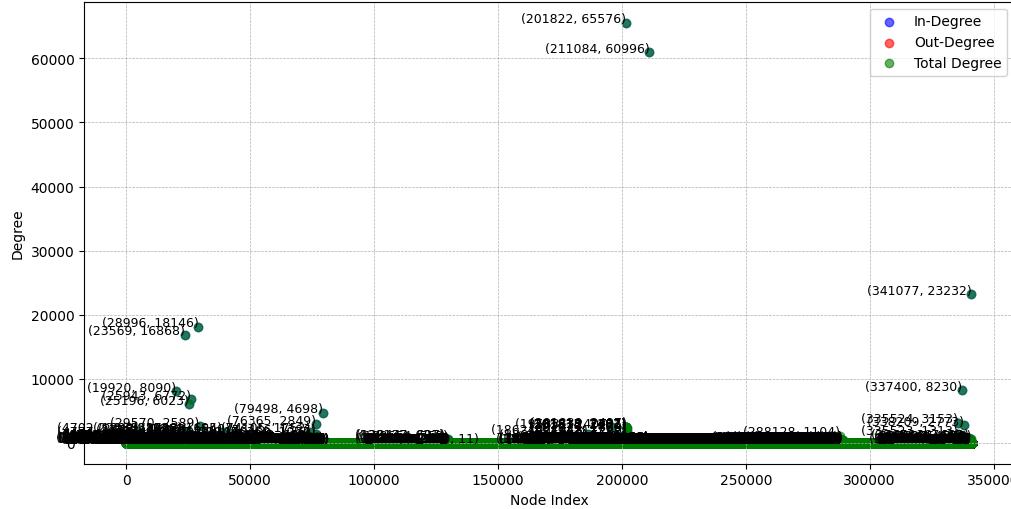
plt.xlabel("Node Index")
plt.ylabel("Degree")
plt.title("Node Degree Distribution")
plt.legend()
plt.grid(True, linestyle="--", linewidth=0.5)

plt.show()

```

→ /usr/local/lib/python3.11/dist-packages/IPython/core/pylabtools.py:151: UserWarning: Creating legend with loc="best" can be slow with large amounts of data.
fig.canvas.print_figure(bytes_io, **kw)

Node Degree Distribution



```

# Annotate points where the total degree > 10
for i, (node, degree) in enumerate(zip(node_indices, total_degrees)):
    if degree > 10000:
        plt.text(node, degree, f"({node}, {degree})", fontsize=9, color="black", ha="right")

plt.xlabel("Node Index")
plt.ylabel("Degree")
plt.title("Node Degree Distribution")
plt.legend()
plt.grid(True, linestyle="--", linewidth=0.5)

plt.show()

```

```
ipython-input-7-74c48fd29b45>9: UserWarning: No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.
plt.legend()
-----
ValueError                                 Traceback (most recent call last)
/usr/local/lib/python3.11/dist-packages/IPython/core/formatters.py in __call__(self, obj)
    339         pass
    340     else:
--> 341         return printer(obj)
    342     # Finally look for special method names
    343     method = get_real_method(obj, self.print_method)

[7 frames]
/usr/local/lib/python3.11/dist-packages/matplotlib/backends/backend_agg.py in __init__(self, width, height, dpi)
    67     self.width = width
    68     self.height = height
---> 69     self._renderer = _RendererAgg(int(width), int(height), dpi)
    70     self._filter_renderers = []
    71

ValueError: Image size of 169174263x24236962 pixels is too large. It must be less than 2^23 in each direction.

<Figure size 640x480 with 1 Axes>

print(dataset.i2e[201822 ],dataset.i2e[288128 ],dataset.i2e[201822 ])
→ ('http://www.opengis.net/ont/geosparql#Geometry', 'iri') ('https://data.labs.pdok.nl/rce/id/monument/8336', 'iri') ('http://www.opengis.net/ont/geosparql#Geometry', 'iri')

relations = set(int(triple[1].item()) for triple in dataset.triples)
print(f"Number of unique relations: {len(relations)}")

→ Number of unique relations: 60

relation_names = [dataset.i2r[rel] for rel in relations]
print("Relations:", relation_names)
→ Relations: [http://data.pdok.nl/def/pdok#asWKT-RD, http://dbpedia.org/ontology/city, http://dbpedia.org/ontology/codeNationalMonument, http://dbpedia.org/ontology/location, http://dbpedia.org/ontology/monument]

relation_dict = {
    "http://data.pdok.nl/def/pdok#asWKT-RD": "geometry",
    "http://dbpedia.org/ontology/city": "city",
    "http://dbpedia.org/ontology/codeNationalMonument": "monument_code",
    "http://dbpedia.org/ontology/location": "location",
    "http://dbpedia.org/ontology/name": "name",
    "http://dbpedia.org/ontology/neighbourhood": "neighbourhood",
    "http://dbpedia.org/ontology/thumbnail": "thumbnail",
    "http://purl.org/dc/terms/created": "created_date",
    "http://purl.org/dc/terms/creator": "creator",
    "http://purl.org/dc/terms/description": "description",
    "http://purl.org/dc/terms/isPartOf": "part_of",
    "http://purl.org/dc/terms/spatial": "spatial",
    "http://schema.org/dateCreated": "date_created",
    "http://schema.org/roleName": "role_name",
    "http://www.geonames.org/ontology#alternateName": "alt_name",
    "http://www.geonames.org/ontology#countryCode": "country_code",
    "http://www.geonames.org/ontology#featureClass": "feature_class",
    "http://www.geonames.org/ontology#featureCode": "feature_code",
    "http://www.geonames.org/ontology#locationMap": "location_map",
    "http://www.geonames.org/ontology#name": "name",
    "http://www.geonames.org/ontology#nearbyFeatures": "nearby_features",
    "http://www.geonames.org/ontology#officialName": "official_name",
    "http://www.geonames.org/ontology#parentADM1": "admin1_parent",
    "http://www.geonames.org/ontology#parentADM2": "admin2_parent",
    "http://www.geonames.org/ontology#parentCountry": "parent_country",
    "http://www.geonames.org/ontology#parentFeature": "parent_feature",
    "http://www.geonames.org/ontology#population": "population",
    "http://www.geonames.org/ontology#wikipediaArticle": "wikipedia",
    "http://www.opengis.net/ont/geosparql#asWKT": "geometry",
    "http://www.opengis.net/ont/geosparql#hasGeometry": "has_geometry",
    "http://www.opengis.net/ont/geosparql#sfWithin": "within",
    "http://www.w3.org/1999/02/22-rdf-syntax-ns#type": "type",
    "http://www.w3.org/2000/01/rdf-schema#isDefinedBy": "defined_by",
    "http://www.w3.org/2000/01/rdf-schema#label": "label",
    "http://www.w3.org/2000/01/rdf-schema#seeAlso": "see_also",
    "http://www.w3.org/2002/07/owl#sameAs": "same_as",
    "http://www.w3.org/2003/01/geo/wgs84_pos#alt": "altitude",
    "http://www.w3.org/2006/vcard/ns#hasStreetAddress": "street_address",
    "http://www.w3.org/2006/vcard/ns#postal-code": "postal_code",
    "http://www.w3.org/2006/vcard/ns#street-address": "street_address",
    "http://xmlns.com/foaf/0.1/depiction": "depiction",
    "http://xmlns.com/foaf/0.1/depicts": "depicts",
    "http://xmlns.com/foaf/0.1/name": "name",
    "https://data.labs.pdok.nl/rce/bouwjaar": "construction_year",
    "https://data.labs.pdok.nl/rce/def/complexnummer": "complex_number",
    "https://data.labs.pdok.nl/rce/def/fotograaf": "photographer",
    "https://data.labs.pdok.nl/rce/def/graveur": "engraver",
    "https://data.labs.pdok.nl/rce/def/huisnummer": "house_number",
    "https://data.labs.pdok.nl/rce/def/huisnummerCompleet": "full_house_number",
    "https://data.labs.pdok.nl/rce/def/huisnummerToevoeging": "house_number_suffix",
    "https://data.labs.pdok.nl/rce/def/internComplexNumbers": "internal_complex_numbers",
    "https://data.labs.pdok.nl/rce/def/isFree": "is_free",
    "https://data.labs.pdok.nl/rce/def/locator": "locator",
    "https://data.labs.pdok.nl/rce/def/ontwerper": "designer",
    "https://data.labs.pdok.nl/rce/def/reprofotograaf": "reproduction_photographer",
    "https://data.labs.pdok.nl/rce/def/rnaSubject": "rna_subject",
    "https://data.labs.pdok.nl/rce/def/schilder": "painter",
    "https://data.labs.pdok.nl/rce/def/techniek": "technique",
    "https://data.labs.pdok.nl/rce/def/tekenaar": "drafter",
    "https://data.pldn.nl/cbs/wijken-buurten/def/cbs#regiocode": "region_code"
}
head_dict = {
    "https://data.labs.pdok.nl/.well-known/genid/": "genid",
}
```

```

"https://data.pldn.nl/cbs/wijken-buurten/regions/2016/id/land-geografisch/": "geo_region",
"http://sws.geonames.org/": "geonames",
"https://data.labs.pdok.nl/cbs/id/gemeente/": "municipality",
"https://data.pldn.nl/cbs/wijken-buurten/regions/2016/id/geometry/": "geometry",
"https://data.labs.pdok.nl/rce/id/image/": "image_by_rce",
"http://bag.basisregistraties.overheid.nl/bag/id/geometry/": "bag_geometry",
"https://data.labs.pdok.nl/rce/id/monument/": "monument",
"http://www.opengis.net/ont/geosparql#": "GeoSparql",
"http://data.labs.pdok.nl/rce/def/Afbeelding": "Afbeelding",
"https://images.memorix.nl/rce/download/fullsize/": "fullsize_image_download",
"http://data.cultureelerfgoed.nl/semnet/": "cultureel_ergoed",
"http://nl.wikipedia.org/wiki/": "wikipedia",
"http://www.geonames.org/ontology#feature": "OntologyFeature",
"https://data.pldn.nl/cbs/wijken-buurten/def/": "wijken_buurten_def",
"https://images.memorix.nl/rce/download/fullsize/": "memorial_images",
"http://www.opengis.net/ont/": "open_gis_ontology",
"http://sws.geonames.org/": "geonames_sws",
"http://data.cultureelerfgoed.nl/semnet/": "cultureel_ergoed_semnet",
"http://www.geonames.org/": "geonames",
"http://www.rnaproject.org/data/": "rna_project_data",
"https://data.labs.pdok.nl/cbs/id/gemeente/": "gemeente",
"https://data.pldn.nl/cbs/wijken-buurten/regions/2016/id/geometry/": "geometry",
"http://dbpedia.org/resource/": "dbpedia_resource",
"https://data.labs.pdok.nl/rce/id/image/": "rce_image_id",
"://": "unknown_protocol",
"https://data.labs.pdok.nl/rce/def/": "rce_def",
"http://nl.wikipedia.org/": "wikipedia",
"http://bag.basisregistraties.overheid.nl/bag/id/geometry/": "bag_geometry",
"https://data.labs.pdok.nl/rce/def/monument/": "monument"
}

```

```

import networkx as nx
import matplotlib.pyplot as plt
import torch
import numpy as np
n = 1000 # Number of random triples you want to select
indices = torch.randperm(dataset.triples.size(0))[:n]

# Sample random triples from the tensor using these indices
random_triples = dataset.triples[211000:211100]
# Randomly select n triples from dataset

def shorten(name, length=70):
    """Ensure name is a string and shorten it intelligently."""
    if isinstance(name, tuple):
        url=name[0]
        for base_url, short_label in head_dict.items():
            if url.startswith(base_url): # Check if URL starts with a known base URL
                return short_label # Return the corresponding short label

    # If tuple contains 'POLYGON', extract and shorten the polygon description
    for part in name:
        if "POLYGON" in part:
            return "POLYGON(...)"
        if "POINT" in part:
            return "Point()" # Shortened for readability
        if 'http://kgbench.info/dt#base64Image' in part:
            return "image_base_64"
        if 'wiki' in part:
            return "image_base_64"

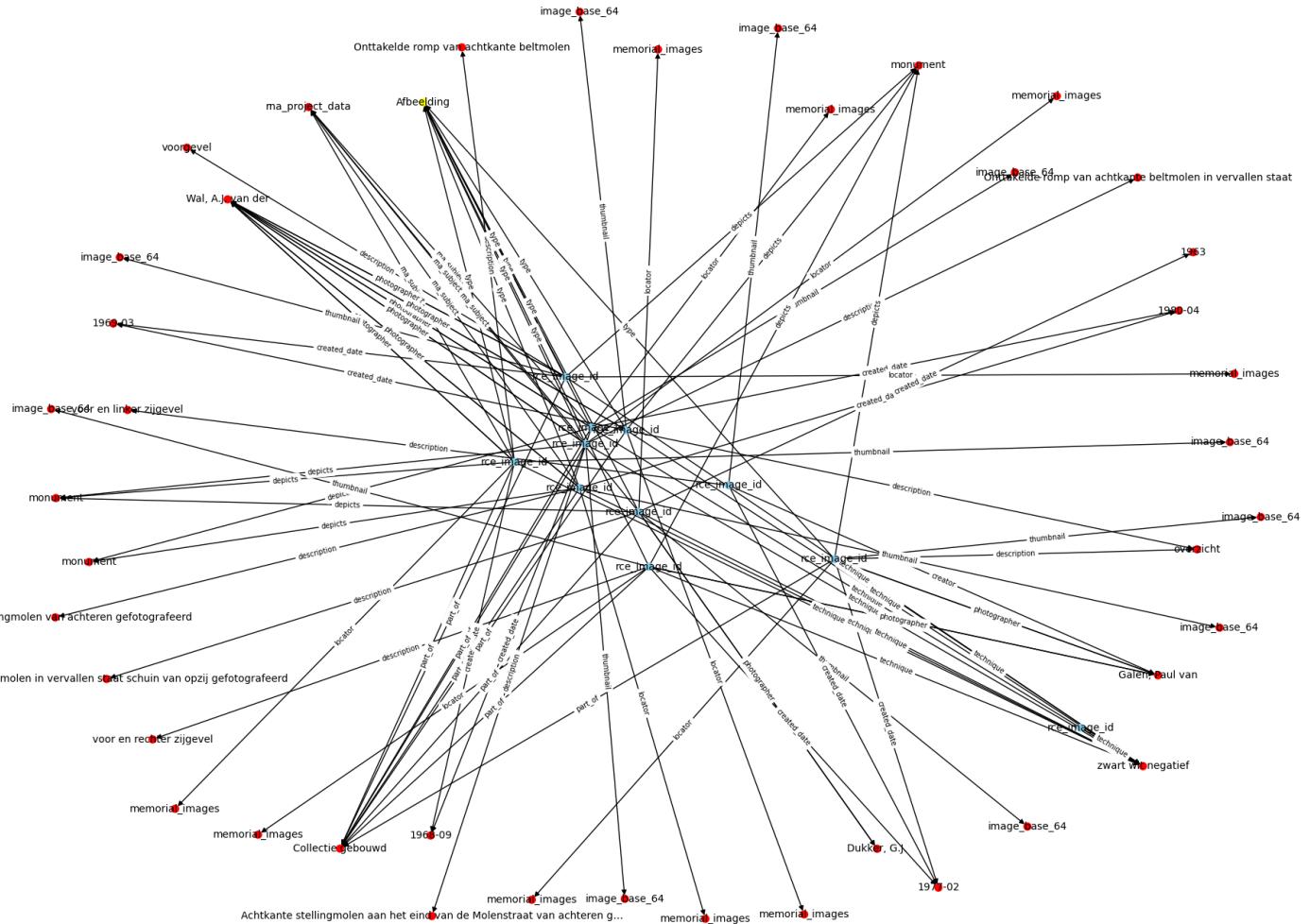
    # Default: Convert entire tuple to string if no specific pattern found
    name = name[0]
    if "http" in name :
        print(name)
    # Truncate if necessary
    return name[:length] + "..." if len(name) > length else name

# Create a MultiDiGraph where edges have labels (relations)
G = nx.MultiDiGraph()
for head, rel, tail in random_triples:
    # G.add_edge(dataset.i2e[head], dataset.i2e[tail], label=dataset.i2r[rel])
    G.add_edge(int(head),int(tail), label=dataset.i2r[rel])
# Set different colors for Head and Tail nodes
head_nodes = {int(head) for head, _, _ in random_triples}
tail_nodes = {int(tail) for _, _, tail in random_triples}

# Assign colors
node_colors = []
for node in G.nodes():
    if int(node)==211084:#to fidn degree for some reason its shown seperately
        print("Found")
        node_colors.append("yellow")
    elif int(node) in head_nodes:
        node_colors.append("skyblue") # Head nodes - Blue
    elif int(node) in tail_nodes:
        node_colors.append("red") # Tail nodes - Red
    else:
        node_colors.append("red") # Other nodes (if any)

# Draw a small subgraph for visualization
plt.figure(figsize=(20,15))
pos = nx.spring_layout(G, k=3/np.sqrt(n), scale=20) # Increased k and scale for better spacing
nx.draw(G, pos, with_labels=False, node_size=50, font_size=8, node_color=node_colors)
edge_labels = {(u, v): relation_dict.get(d['label'], "unkown") for u, v, d in G.edges(data=True)}
nx.draw_networkx_edge_labels(G, pos, edge_labels=edge_labels, font_size=7)
node_labels = {node: shorten(dataset.i2e[node]) for node in G.nodes()}
nx.draw_networkx_labels(G, pos, labels=node_labels, font_size=10)
plt.show()

```



```
# Print nodes with degree > 1 along with their mapped value from dataset.i2e
print("Nodes with Degree > 1:")

for node in G.nodes():
    degree = G.degree(node)
    if degree > 1:
        node_value = dataset.i2e[node] # Handle missing mappings
        print(f"Node: {node}, Degree: {degree}, Mapped Value: {node_value}")
```

```
Nodes with Degree > 1:
Node: 341077, Degree: 10, Mapped Value: ('zwart wit negatief', '@nl-nl')
Node: 228653, Degree: 11, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215140', 'iri')
Node: 4328, Degree: 2, Mapped Value: ('1980-04', 'none')
Node: 79498, Degree: 12, Mapped Value: ('Wal, A.J. van den', 'none')
Node: 23569, Degree: 9, Mapped Value: ('Collectie gebouwd', '@nl-nl')
Node: 211084, Degree: 9, Mapped Value: ('https://data.labs.pdok.nl/rce/def/Afbeelding', 'iri')
Node: 283366, Degree: 2, Mapped Value: ('https://data.labs.pdok.nl/rce/id/monument/39032', 'iri')
Node: 201839, Degree: 5, Mapped Value: ('http://www.rnaproject.org/data/0d4bd09f-e7e6-44b1-8939-e6f65d3cf47a', 'http://www.w3.org/2001/XMLSchema#anyURI')
Node: 228654, Degree: 11, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215142', 'iri')
Node: 228656, Degree: 11, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215428', 'iri')
Node: 25943, Degree: 2, Mapped Value: ('Dukker, G.J.', 'none')
Node: 286852, Degree: 3, Mapped Value: ('https://data.labs.pdok.nl/rce/id/monument/526018', 'iri')
Node: 228657, Degree: 11, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215429', 'iri')
Node: 4160, Degree: 2, Mapped Value: ('1968-09', 'none')
Node: 228658, Degree: 11, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215430', 'iri')
Node: 228659, Degree: 10, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215498', 'iri')
Node: 4167, Degree: 2, Mapped Value: ('1969-03', 'none')
Node: 287106, Degree: 4, Mapped Value: ('https://data.labs.pdok.nl/rce/id/monument/529104', 'iri')
Node: 228660, Degree: 2, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215499', 'iri')
Node: 336504, Degree: 2, Mapped Value: ('overzicht', '@nl-nl')
Node: 228661, Degree: 10, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215507', 'iri')
Node: 4284, Degree: 3, Mapped Value: ('1977-02', 'none')
Node: 29570, Degree: 5, Mapped Value: ('Galen, Paul van', 'none')
Node: 228662, Degree: 10, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215508', 'iri')
Node: 228663, Degree: 4, Mapped Value: ('https://data.labs.pdok.nl/rce/id/image/20215509', 'iri')
```

```
for head, _, tail in random_triples:
    if int(head.item()) == 211084 or int(tail) == 211084:
        print(f" head {head.item()} {(type(head))} and tail {tail} ")

→ head 228653 <class 'torch.Tensor'> and tail 211084
head 228654 <class 'torch.Tensor'> and tail 211084
head 228656 <class 'torch.Tensor'> and tail 211084
head 228657 <class 'torch.Tensor'> and tail 211084
head 228658 <class 'torch.Tensor'> and tail 211084
head 228659 <class 'torch.Tensor'> and tail 211084
head 228660 <class 'torch.Tensor'> and tail 211084
head 228661 <class 'torch.Tensor'> and tail 211084
head 228662 <class 'torch.Tensor'> and tail 211084

node_degrees = dict(G.degree()) # Total degree (in-degree + out-degree)
filtered_nodes = {node for node, degree in node_degrees.items() if degree >= 2} # Keep nodes with degree ≥ 2

# Create a subgraph with only the filtered nodes
G_filtered = G.subgraph(filtered_nodes).copy()

# Assign colors
node_colors = ["skyblue" if node in filtered_nodes else "red" for node in G_filtered.nodes()]

# Draw the filtered graph
plt.figure(figsize=(20, 15))
pos = nx.spring_layout(G_filtered, k=3/np.sqrt(n), scale=20)
nx.draw(G_filtered, pos, with_labels=False, node_size=50, font_size=8, node_color=node_colors)
edge_labels = {(u, v): relation_dict.get(d['label'], "unknown") for u, v, d in G_filtered.edges(data=True)}
nx.draw_networkx_edge_labels(G_filtered, pos, edge_labels=edge_labels, font_size=7)
node_labels = {node: shorten(dataset.ize[node]) for node in G_filtered.nodes()}
nx.draw_networkx_labels(G_filtered, pos, labels=node_labels, font_size=10)
plt.show()

print(f"Original graph nodes: {G.number_of_nodes()}, edges: {G.number_of_edges()}")
print(f"Filtered graph nodes: {G_filtered.number_of_nodes()}, edges: {G_filtered.number_of_edges()}"
```

⤵

Original graph nodes: 2000, edges: 1000
Filtered graph nodes: 0, edges: 0

import random

```
random_triples = random.sample(dataset.triples.tolist(), 70) # Select 120 random triples
urls=[]
for head, rel, tail in random_triples:
    head_value = dataset.i2e[head] # Get the actual head value
    tail_value = dataset.i2e[tail]
    urls.append(head_value)
    if "@nl-nl" == tail_value[1] or "none" == tail_value[1]:
        print(tail_value)
```

⤵ ('Nieuwstraat', 'none')
('Achterkant stellingmolen in de winter schuin van opzij gefotografeerd achterloods', '@nl-nl')
('48', 'none')
('Dukker, G.J.', 'none')
('POLYGON ((156818.473816 462205.330627, 156820.317017 462208.863037, 156828.575989 462224.725037, 156829.343018 462226.197998, 156834.826416 462236.93103, 156841.749023 462250.481018, 156842.267029 462267.11103, 156843.85103 462274.55103, 156844.44103 462281.24103, 156845.03103 462287.93103, 156846.62103 462294.62103, 156847.21103 462301.31103, 156848.80103 462308.00103, 156849.39103 462314.69103, 156850.98103 462321.38103, 156851.57103 462328.07103, 156852.16103 462334.76103, 156853.75103 462341.45103, 156854.34103 462348.14103, 156855.93103 462354.83103, 156856.52103 462361.52103, 156857.11103 462368.21103, 156858.70103 462374.90103, 156859.29103 462381.59103, 156860.88103 462388.28103, 156861.47103 462394.97103, 156862.06103 462401.66103, 156863.65103 462408.35103, 156864.24103 462415.04103, 156865.83103 462421.73103, 156866.42103 462428.42103, 156867.01103 462435.11103, 156868.60103 462441.80103, 156869.19103 462448.49103, 156870.78103 462455.18103, 156871.37103 462461.87103, 156872.96103 462468.56103, 156873.55103 462475.25103, 156874.14103 462481.94103, 156875.73103 462488.63103, 156876.32103 462495.32103, 156877.91103 462502.01103, 156878.50103 462508.70103, 156879.09103 462515.39103, 156880.68103 462522.08103, 156881.27103 462528.77103, 156882.86103 462535.46103, 156883.45103 462542.15103, 156884.04103 462548.84103, 156885.63103 462555.53103, 156886.22103 462562.22103, 156887.81103 462568.91103, 156888.40103 462575.60103, 156889.99103 462582.29103, 156890.58103 462588.98103, 156891.17103 462595.67103, 156892.76103 462602.36103, 156893.35103 462609.05103, 156894.94103 462615.74103, 156895.53103 462622.43103, 156896.12103 462629.12103, 156897.71103 462635.81103, 156898.30103 462642.50103, 156899.89103 462649.19103, 156900.48103 462655.88103, 156901.07103 462662.57103, 156902.66103 462669.26103, 156903.25103 462675.95103, 156904.84103 462682.64103, 156905.43103 462689.33103, 156906.02103 462695.02103, 156907.61103 462701.71103, 156908.20103 462708.40103, 156909.79103 462715.09103, 156910.38103 462721.78103, 156911.97103 462728.47103, 156912.56103 462735.16103, 156913.15103 462741.85103, 156914.74103 462748.54103, 156915.33103 462755.23103, 156916.92103 462761.92103, 156917.51103 462768.61103, 156918.10103 462775.30103, 156919.69103 462782.09103, 156920.28103 462788.78103, 156921.87103 462795.47103, 156922.46103 462802.16103, 156923.05103 462808.85103, 156924.64103 462815.54103, 156925.23103 462822.23103, 156926.82103 462828.92103, 156927.41103 462835.61103, 156928.00103 462842.30103, 156929.59103 462849.09103, 156930.18103 462855.78103, 156931.77103 462862.47103, 156932.36103 462869.16103, 156933.95103 462875.85103, 156934.54103 462882.54103, 156935.13103 462889.23103, 156936.72103 462895.92103, 156937.31103 462902.61103, 156938.90103 462909.30103, 156939.49103 462915.99103, 156940.08103 462922.68103, 156941.67103 462929.37103, 156942.26103 462936.06103, 156943.85103 462942.75103, 156944.44103 462949.44103, 156945.03103 462956.13103, 156946.62103 462962.82103, 156947.21103 462969.51103, 156948.80103 462976.20103, 156949.39103 462982.89103, 156950.98103 462989.58103, 156951.57103 462996.27103, 156952.16103 463002.96103, 156953.75103 463009.65103, 156954.34103 463016.34103, 156955.93103 463023.03103, 156956.52103 463029.72103, 156957.11103 463036.41103, 156958.70103 463043.10103, 156959.29103 463049.79103, 156960.88103 463056.48103, 156961.47103 463063.17103, 156962.06103 463069.86103, 156963.65103 463076.55103, 156964.24103 463083.24103, 156965.83103 463089.93103, 156966.42103 463096.62103, 156967.01103 463103.31103, 156968.60103 463109.99103, 156969.19103 463116.68103, 156970.78103 463123.37103, 156971.37103 463129.96103, 156972.96103 463136.65103, 156973.55103 463143.34103, 156974.14103 463149.93103, 156975.73103 463156.62103, 156976.32103 463163.31103, 156977.91103 463169.99103, 156978.50103 463176.68103, 156979.09103 463183.37103, 156980.68103 463189.96103, 156981.27103 463196.65103, 156982.86103 463203.34103, 156983.45103 463209.93103, 156984.04103 463216.62103, 156985.63103 463223.31103, 156986.22103 463229.99103, 156987.81103 463236.68103, 156988.40103 463243.37103, 156989.99103 463249.96103, 156990.58103 463256.65103, 156991.17103 463263.34103, 156992.76103 463269.93103, 156993.35103 463276.62103, 156994.94103 463283.31103, 156995.53103 463289.99103, 156996.12103 463296.68103, 156997.71103 463303.37103, 156998.30103 463310.06103, 156999.89103 463316.65103, 157000.48103 463323.34103, 157001.07103 463329.93103, 157002.66103 463336.62103, 157003.25103 463343.31103, 157004.84103 463350.09103, 157005.43103 463356.78103, 157006.02103 463363.47103, 157007.61103 463370.16103, 157008.20103 463376.85103, 157009.79103 463383.54103, 157010.38103 463390.23103, 157011.97103 463396.92103, 157012.56103 463403.61103, 157013.15103 463410.30103, 157014.74103 463416.99103, 157015.33103 463423.68103, 157016.92103 463430.37103, 157017.51103 463437.06103, 157018.10103 463443.75103, 157019.69103 463450.44103, 157020.28103 463457.13103, 157021.87103 463463.82103, 157022.46103 463470.51103, 157023.05103 463477.20103, 157024.64103 463483.89103, 157025.23103 463490.58103, 157026.82103 463497.27103, 157027.41103 463503.96103, 157028.00103 463510.65103, 157029.59103 463517.34103, 157030.18103 463524.03103, 157031.77103 463530.72103, 157032.36103 463537.41103, 157033.95103 463544.10103, 157034.54103 463550.79103, 157035.13103 463557.48103, 157036.72103 463564.17103, 157037.31103 463570.86103, 157038.90103 463577.55103, 157039.49103 463584.24103, 157040.08103 463590.93103, 157041.67103 463597.62103, 157042.26103 463604.31103, 157043.85103 463610.99103, 157044.44103 463617.68103, 157045.03103 463624.37103, 157046.62103 463631.06103, 157047.21103 463637.75103, 157048.80103 463644.44103, 157049.39103 463651.13103, 157050.98103 463657.82103, 157051.57103 463664.51103, 157052.16103 463671.20103, 157053.75103 463677.89103, 157054.34103 463684.58103, 157055.93103 463691.27103, 157056.52103 463697.96103, 157057.11103 463704.65103, 157058.70103 463711.34103, 157059.29103 463717.99103, 157060.88103 463724.68103, 157061.47103 463731.37103, 157062.06103 463738.06103, 157063.65103 463744.75103, 157064.24103 463751.44103, 157065.83103 463758.13103, 157066.42103 463764.82103, 157067.01103 463771.51103, 157068.60103 463778.20103, 157069.19103 463784.89103, 157070.78103 463791.58103, 157071.37103 463798.27103, 157072.96103 463804.96103, 157073.55103 463811.65103, 157074.14103 463818.34103, 157075.73103 463824.99103, 157076.32103 463831.68103, 157077.91103 463838.37103, 157078.50103 463845.06103, 157079.09103 463851.75103, 157080.68103 463858.44103, 157081.27103 463865.13103, 157082.86103 463871.82103, 157083.45103 463878.51103, 157084.04103 463885.20103, 157085.63103 463891.89103, 157086.22103 463898.58103, 157087.81103 463905.27103, 157088.40103 463911.96103, 157089.99103 463918.65103, 157090.58103 463925.34103, 157091.17103 463931.99103, 157092.76103 463938.68103, 157093.35103 463945.37103, 157094.94103 463952.06103, 157095.53103 463958.75103, 157096.12103 463965.44103, 157097.71103 463972.13103, 157098.30103 463978.82103, 157099.89103 463985.51103, 157100.48103 463992.20103, 157101.07103 464008.89103, 157102.66103 464015.58103, 157103.25103 464022.27103, 157104.84103 464028.96103, 157105.43103 464035.65103, 157106.02103 464042.34103, 157107.61103 464049.03103, 157108.20103 464055.72103, 157109.79103 464062.41103, 157110.38103 464069.10103, 157111.97103 464075.79103, 157112.56103 464082.48103, 157113.15103 464089.17103, 157114.74103 464095.86103, 157115.33103 464102.55103, 157116.92103 464109.24103, 157117.51103 464115.93103, 157118.10103 464122.62103, 157119.69103 464129.31103, 157120.28103 464135.99103, 157121.87103 464142.68103, 157122.46103 464149.37103, 157123.05103 464156.06103, 157124.64103 464162.75103, 157125.23103 464169.44103, 157126.82103 464176.13103, 157127.41103 464182.82103, 157128.00103 464189.51103, 157129.59103 464196.20103, 157130.18103 464202.89103, 157131.77103 464209.58103, 157132.36103 464216.27103, 157133.95103 464222.96103, 157134.54103 464229.65103, 157135.13103 464236.34103, 157136.72103 464243.03103, 157137.31103 464249.72103, 157138.90103 464256.41103, 157139.49103 464263.10103, 157140.08103 464269.79103, 157141.67103 464276.48103, 157142.26103 464283.17103, 157143.85103 464289.86103, 157144.44103 464296.55103, 157145.03103 464303.24103, 157146.62103 464309.93103, 157147.21103 464316.62103, 157148.80103 464323.31103, 157149.39103 464329.99103, 157150.98103 464336.68103, 157151.57103 464343.37103, 157152.16103 464349.96103, 157153.75103 464356.65103, 157154.34103 464363.34103, 157155.93103 464370.03103, 157156.52103 464376.72103, 157157.11103 464383.41103, 157158.70103 464389.99103, 157159.29103 464396.68103, 157160.88103 464403.37103, 157161.47103 464410.06103, 157162.06103 464416.75103, 157163.65103 464423.44103, 157164.24103 464430.13103, 157165.83103 464436.82103, 157166.42103 464443.51103, 157167.01103 464450.20103, 157168.60103 464456.89103, 157169.19103 464463.58103, 157170.78103 464470.27103, 157171.37103 464476.96103, 157172.96103 464483.65103, 157173.55103 464490.34103, 157174.14103 464497.03103, 157175.73103 464503.72103, 157176.32103 464510.41103, 157177.91103 464517.10103, 157178.50103 464523.79103, 157179.09103 464530.48103, 157180.68103 464537.17103, 157181.27103 464543.86103, 157182.86103 464550.55103, 157183.45103 464557.24103, 157184.04103 464563.93103, 157185.63103 464570.62103, 157186.22103 464577.31103, 157187.81103 464583.99103, 157188.40103 464590.68103, 157189.99103 464597.37103, 157190.58103 464604.06103, 157191.17103 464610.75103, 157192.76103 464617.44103, 157193.35103 464624.13103, 157194.94103 464630.82103, 157195.53103 464637.51103, 157196.12103 464644.20103, 157197.71103 464650.89103, 157198.30103 464657.58103, 157199.89103 464664.27103, 157200.48103 464670.96103, 157201.07103 464677.65103, 157202.66103 464684.34103, 157203.25103 464690.99103, 157204.84103 464697.68103, 157205.43103 464704.37103, 157206.02103 464711.06103, 157207.61103 464717.75103, 157208.20103 464724.44103, 157209.79103 464731.13103, 157210.38103 464737.82103, 157211.97103 464744.51103, 157212.56103 464751.20103, 157213.15103 464757.89103, 157214.74103 464764.58103, 157215.33103 464771.27103, 157216.92103 464777.96103, 157217.51103 464784.65103, 157218.10103 464791.34103, 157219.69103 464797.99103, 157220.28103 464804.68103, 157221.87103 464811.37103, 157222.46103 464818.06103, 157223.05103 464824.75103, 157224.64103 464831.44103, 157225.23103 464838.13103, 157226.82103 464844.82103, 157227.41103 464851.51103, 157228.00103 464858.20103, 157229.59103 464864.89103, 157230.18103 464871.58103, 157231.77103 464878.27103, 157232.36103 464884.96103, 157233.95103 464891.65103, 157234.54103 464898.34103, 157235.13103 464905.03103, 157236.72103 464911.72103, 157237.31103 464918.41103, 157238.90103 464925.10103, 157239.49103 464931.79103, 157240.08103 464938.48103, 157241.67103 464945.17103, 157242.26103 464951.86103, 157243.85103 464958.55103, 1

```
('1965', 'none')
('Gevelsteen', '@nl-nl')
('Papenvoort 2A te Nuenen', 'none')
('Overzicht van de rechter zijgevel en de achtergevel', '@nl-nl')
('Collectie gebouwd', '@nl-nl')
```

```
from urllib.parse import urlparse
filtered_uris = []
uniqueurlset()
for head,_,_ in dataset.triples:
    url=dataset.i2e[head][0]
    parsed_url = urlparse(url)
    base_path = "/".join(parsed_url.path.strip('/').split('/')[ :-1]) # Remove last segment
    new_url = f'{parsed_url.scheme}://{parsed_url.netloc}/{base_path}' # Reconstruct the URL
    uniqueurl.add(new_url)
    filtered_uris.append(new_url)
print(len(uniqueurl))
# Print results
for url in filtered_uris:
    print(len(url))
```

Start coding or generate with AI.

```
→ {http://sws.geonames.org/2744199/, http://sws.geonames.org/2756359/, http://sws.geonames.org/2755243/, http://sws.geonames.org/2745154/, http://sws.geonames.org/2749689/, http://sws.geonames.o
```

finding unqire heads

```
import random
from urllib.parse import urlparse
from tqdm import tqdm

unique_urls = set()

for head, rel, tail in tqdm(dataset.triples, desc="Processing triples"):
    head_value = dataset.i2e[head] # Get the actual head value
    if not isinstance(head_value, (list, tuple)): # Ensure it's iterable
        print(f"Unexpected head_value: {head_value}")
        continue

    parsed_url = urlparse(head_value[0])
    base_path = "/".join(parsed_url.path.strip('/').split('/')[ :-1]) # Remove last segment
    new_url = f'{parsed_url.scheme}://{parsed_url.netloc}/{base_path}' # Reconstruct the URL
    unique_urls.add(new_url)

# Save unique URLs to a text file
output_file = "unique_urls.txt"
with open(output_file, "w") as f:
    for url in unique_urls:
        f.write(url + "\n")

print(f"Total unique base URLs: {len(unique_urls)})")
```

```
→ Processing triples: 100%|██████████| 777124/777124 [00:14<00:00, 54190.27it/s]Total unique base URLs: 3200
```

```
unique_urls2 = {url for url in unique_urls if ".geonames.org" not in url and "wiki" not in url }
unique_urls2.add("http://sws.geonames.org/")
unique_urls2.add("http://www.geonames.org/")
unique_urls2.add("http://wikipedia.org/")

# Print the cleaned set
print(len(unique_urls2),(unique_urls2))
```

```
→ 18 {https://data.labs.pdok.nl/.well-known/genid/, https://data.pldn.nl/cbs/wijken-buurten/regios/2016/id/land-geografisch/, https://data.pldn.nl/cbs/wijken-buurten/def/, https://images.memorix.nl/
```

```
import requests
from rdflib import Graph
```

```
geonames_id = "2755348"
rdf_url = f"http://sws.geonames.org/{geonames_id}/about.rdf"

g = Graph()
g.parse(rdf_url)

for s, p, o in g:
    print(s, p, o)
```

```
→ https://sws.geonames.org/2755348/ http://www.w3.org/2003/01/geo/wgs84\_pos#long 6.12083
https://sws.geonames.org/2755348/about.rdf http://creativecommons.org/ns#attributionName GeoNames
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#parentADM2 https://sws.geonames.org/6544256/
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#nearbyFeatures https://sws.geonames.org/2755348/nearby.rdf
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#locationMap https://www.geonames.org/2755348/greffelkamp.html
https://sws.geonames.org/2755348/ http://purl.org/dc/terms/modified 2007-06-03
https://sws.geonames.org/2755348/ http://www.w3.org/2003/01/geo/wgs84\_pos#lat 51.9525
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#countryCode NL
https://sws.geonames.org/2755348/ http://creativecommons.org/ns#license https://creativecommons.org/licenses/by/4.0/
https://sws.geonames.org/2755348/ http://www.w3.org/1999/02/22-rdf-syntax-ns#type http://xmlns.com/foaf/0.1/Document
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#featureClass https://www.geonames.org/ontology#P
https://sws.geonames.org/2755348/ http://creativecommons.org/ns#attributionURL https://www.geonames.org/
https://sws.geonames.org/2755348/ http://www.geonames.org/2755348/about.rdf http://www.geonames.org/2755348/
https://sws.geonames.org/2755348/ http://www.w3.org/1999/02/22-rdf-syntax-ns#type http://www.geonames.org/ontology#Feature
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#parentCountry https://sws.geonames.org/2758045/
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#featureCode https://www.geonames.org/ontology#PPL
https://sws.geonames.org/2755348/ http://purl.org/dc/terms/created 2006-01-15
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#parentFeature https://sws.geonames.org/6544256/
https://sws.geonames.org/2755348/ http://www.geonames.org/ontology#name Greffelkamp
```

<https://sws.geonames.org/2755348/> <http://www.geonames.org/ontology#parentADM1> <https://sws.geonames.org/2755634/>
<https://sws.geonames.org/2755348/> <http://www.w3.org/2000/01/rdf-schema#isDefinedBy> <https://sws.geonames.org/2755348/about.rdf>

unique tails

```
import random
from urllib.parse import urlparse
from tqdm import tqdm

tails=[]
for head, rel, tail in tqdm(dataset.triples, desc="Processing triples"):
    tail_value = dataset.i2e[tail] # Get the actual head value
    tails.append(tail_value)

# Save unique URLs to a text file
output_file = "tails.txt"
with open(output_file, "w") as f:
    for tail in tails:
        f.write(tail[0] + "\n")

print(f"Total tails: {len(tails)}")

# Save unique URLs to a text file
output_file = "tails.txt"
with open(output_file, "w") as f:
    for tail in tails:
        f.write(tail[0] + "\n")

print(f"Total tails: {len(tails)}")

→ Total tails: 777124

torch.version.cuda
→ '12.4'

!pip install -q torch-geometric
!pip install open_clip_torch
!pip install shapely # For handling geometry
!pip install PILLOW
→ 63.1/63.1 kB 3.6 MB/s eta 0:00:00
→ 1.1/1.1 MB 33.8 MB/s eta 0:00:00
Collecting open_clip_torch
  Downloading open_clip_torch-2.31.0-py3-none-any.whl.metadata (31 kB)
Requirement already satisfied: torch>=1.9.0 in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (2.6.0+cu124)
Requirement already satisfied: torchvision in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (0.21.0+cu124)
Requirement already satisfied: regex in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (2024.11.6)
Collecting ftfy (from open_clip_torch)
  Downloading ftfy-6.3.1-py3-none-any.whl.metadata (7.3 kB)
Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (4.67.1)
Requirement already satisfied: huggingface-hub in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (0.29.3)
Requirement already satisfied: safetensors in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (0.5.3)
Requirement already satisfied: timm in /usr/local/lib/python3.11/dist-packages (from open_clip_torch) (1.0.15)
Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (3.18.0)
Requirement already satisfied: typing-extensions=>4.10.0 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (4.12.2)
Requirement already satisfied: networkx in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (3.4.2)
Requirement already satisfied: jinja2 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (3.1.6)
Requirement already satisfied: fsspec in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (2025.3.0)
Requirement already satisfied: nvidia-cuda-nvrtc-cu12=>12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.4.127)
Requirement already satisfied: nvidia-cuda-runtime-cu12=>12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.4.127)
Requirement already satisfied: nvidia-cuda-cupti-cu12=>12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.4.127)
Requirement already satisfied: nvidia-cudnn-cu12=>9.1.0.70 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (9.1.0.70)
Requirement already satisfied: nvidia-cublas-cu12=>12.4.5.8 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.4.5.8)
Requirement already satisfied: nvidia-cufft-cu12=>11.2.1.3 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (11.2.1.3)
Requirement already satisfied: nvidia-curand-cu12=>10.3.5.147 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (10.3.5.147)
Requirement already satisfied: nvidia-cusolver-cu12=>11.6.1.9 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (11.6.1.9)
Requirement already satisfied: nvidia-cusparse-cu12=>12.3.1.170 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.3.1.170)
Requirement already satisfied: nvidia-cusparseelt-cu12=>0.6.2 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (0.6.2)
Requirement already satisfied: nvidia-nccl-cu12=>2.21.5 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (2.21.5)
Requirement already satisfied: nvidia-nvtx-cu12=>12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.4.127)
Requirement already satisfied: nvidia-nvjitlink-cu12=>12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (12.4.127)
Requirement already satisfied: triton=>3.2.0 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (3.2.0)
Requirement already satisfied: sympy=>1.13.1 in /usr/local/lib/python3.11/dist-packages (from torch>1.9.0->open_clip_torch) (1.13.1)
Requirement already satisfied: mpmath<1.4,>=1.1.0 in /usr/local/lib/python3.11/dist-packages (from sympy=>1.13.1->torch>=1.9.0->open_clip_torch) (1.3.0)
Requirement already satisfied: wctwid in /usr/local/lib/python3.11/dist-packages (from ftfy->open_clip_torch) (0.2.13)
Requirement already satisfied: packaging=>20.9 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub->open_clip_torch) (24.2)
Requirement already satisfied: pyyaml=>5.1 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub->open_clip_torch) (6.0.2)
Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from huggingface-hub->open_clip_torch) (2.32.3)
Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (from torchvision->open_clip_torch) (2.0.2)
Requirement already satisfied: pillow=>8.3.*,>=5.3.0 in /usr/local/lib/python3.11/dist-packages (from torchvision->open_clip_torch) (11.1.0)
Requirement already satisfied: MarkupSafe=>2.0 in /usr/local/lib/python3.11/dist-packages (from jinja2->torch>1.9.0->open_clip_torch) (3.0.2)
Requirement already satisfied: charset-normalizer<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub->open_clip_torch) (3.4.1)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub->open_clip_torch) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub->open_clip_torch) (2.3.0)
Requirement already satisfied: certifi=>2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub->open_clip_torch) (2025.1.31)
Downloading open_clip_torch-2.31.0-py3-none-any.whl (1.5 MB)
→ 1.5/1.5 kB 28.3 MB/s eta 0:00:00
Downloading ftfy-6.3.1-py3-none-any.whl (44 kB)
→ 44.8/44.8 kB 1.7 MB/s eta 0:00:00
Installing collected packages: ftfy, open_clip_torch
Successfully installed ftfy-6.3.1 open_clip_torch-2.31.0
Requirement already satisfied: shapely in /usr/local/lib/python3.11/dist-packages (2.0.7)
Requirement already satisfied: numpy<3,>=1.14 in /usr/local/lib/python3.11/dist-packages (from shapely) (2.0.2)
Requirement already satisfied: PILLOW in /usr/local/lib/python3.11/dist-packages (11.1.0)
```

