

▼ Demonstration of networkKit

```
1 !git clone https://github.com/networkit/networkit.git
```

```
Cloning into 'networkit'...
remote: Enumerating objects: 78077, done.
remote: Counting objects: 100% (399/399), done.
remote: Compressing objects: 100% (190/190), done.
remote: Total 78077 (delta 221), reused 298 (delta 190), pack-reused 77678
Receiving objects: 100% (78077/78077), 259.62 MiB | 30.03 MiB/s, done.
Resolving deltas: 100% (56668/56668), done.
```

```
1 !pip install networkit
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public
Collecting networkit
  Downloading networkit-10.1-cp39-cp39-manylinux_2_12_x86_64.manylinux2010_x86_64.whl (9.9 MB)
    9.9/9.9 MB 87.5 MB/s eta 0:00:00
Requirement already satisfied: scipy in /usr/local/lib/python3.9/dist-packages (from networkit)
Requirement already satisfied: numpy in /usr/local/lib/python3.9/dist-packages (from networkit)
Installing collected packages: networkit
Successfully installed networkit-10.1
```

```
1 import networkit as nk
```

```
1 # create a small undirected graph with 5 nodes and 5 edges
```

```
2 G = nk.Graph(5)
```

```
3 G.addEdge(0, 1)
```

```
4 G.addEdge(0, 2)
```

```
5 G.addEdge(1, 3)
```

```
6 G.addEdge(2, 3)
```

```
7 G.addEdge(3, 4)
```

```
8
```

```
True
```

```
1 import networkx as nx
```

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

```
3
```

```
4 # create a graph
```

```
5 G = nx.Graph()
```

```
6 G.add_edges_from([(0, 1), (0, 2), (1, 3), (2, 3), (3, 4)])
```

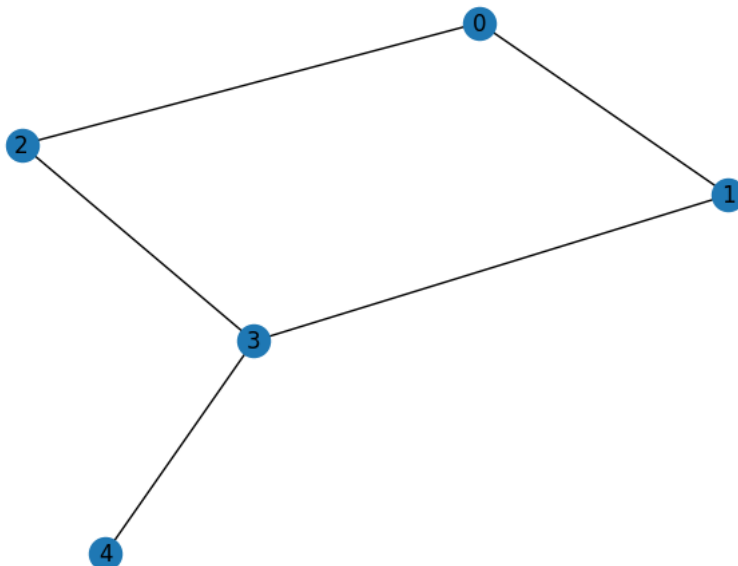
```
7
```

```
8 # draw the graph
```

```
9 nx.draw(G, with_labels=True)
```

```
10 plt.show()
```

```
11
```



▼ calculate degree centrality for each node

```
1 dc = nk centrality .DegreeCentrality(G)
2 dc.run()
3 print("Degree centrality:", dc.ranking())
4
Degree centrality: [(3, 3.0), (0, 2.0), (1, 2.0), (2, 2.0), (4, 1.0)]
```

▼ calculate betweenness centrality for each node

```
1
2 bc = nk centrality .Betweenness(G)
3 bc.run()
4 print("Betweenness centrality:", bc.ranking())
5
Betweenness centrality: [(3, 7.0), (1, 2.0), (2, 2.0), (0, 1.0), (4, 0.0)]
```

▼ calculate shortest paths between all pairs of nodes

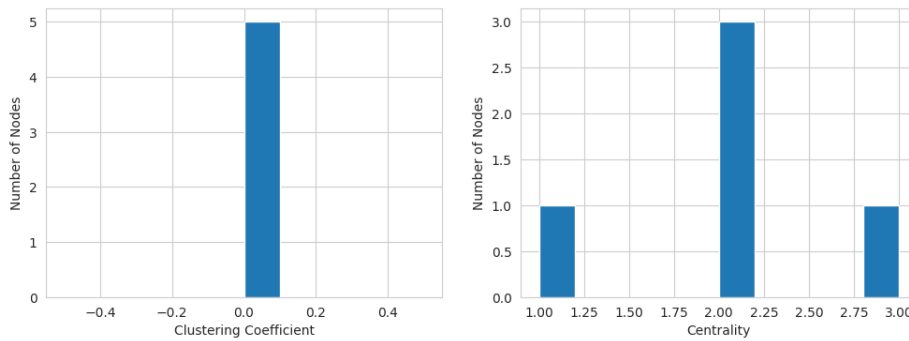
```
1 G = nk .Graph(5)
2
3 # add edges to the graph
4 G.addEdge(0, 1)
5 G.addEdge(0, 3)
6 G.addEdge(1, 2)
7 G.addEdge(1, 3)
8 G.addEdge(2, 3)
9 G.addEdge(2, 4)
10 apsp = nk .distance .APSP(G)
11 apsp.run()
12
13 # get the matrix of shortest path lengths
14 shortest_paths = apsp.getDistances()
15
16 # print the shortest path between node 0 and all other nodes
17 for i, distance in enumerate(shortest_paths[0]):
18     print("Shortest path between nodes 0 and {}: {}".format(i, distance))
19
[ ] Shortest path between nodes 0 and 0: 0.0
Shortest path between nodes 0 and 1: 1.0
Shortest path between nodes 0 and 2: 2.0
Shortest path between nodes 0 and 3: 1.0
Shortest path between nodes 0 and 4: 3.0
```

```
1 import networkkit as nk
2 import matplotlib.pyplot as plt
3
4 # create a small undirected graph with 5 nodes and 5 edges
5 G = nk .Graph(5)
6 G.addEdge(0, 1)
7 G.addEdge(0, 2)
8 G.addEdge(1, 3)
9 G.addEdge(2, 3)
10 G.addEdge(3, 4)
11
12 # calculate the clustering coefficient of each node
13 cc = nk .centrality .LocalClusteringCoefficient(G)
14 cc_values = cc.run().scores()
15
16 # calculate the centrality of each node
17 centrality = nk .centrality .DegreeCentrality(G)
18 centrality_values = centrality.run().scores()
19
20
21
22 # plot the clustering coefficient, centrality, and degree distribution
23 fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(12,4))
24
25 ax1.hist(cc_values, bins=10)
26 ax1.set_xlabel("Clustering Coefficient")
27 ax1.set_ylabel("Number of Nodes")
```

```

28
29 ax2.hist(centrality_values, bins=10)
30 ax2.set_xlabel("Centrality")
31 ax2.set_ylabel("Number of Nodes")
32
33
34
35 plt.show()
36

```



▼ Community Detection

```

1 import networkkit as nk
2 import matplotlib.pyplot as plt
3
4 # create a small undirected graph with 5 nodes and 5 edges
5 G = nk.Graph(5)
6 G.addEdge(0, 1)
7 G.addEdge(0, 2)
8 G.addEdge(1, 3)
9 G.addEdge(2, 3)
10 G.addEdge(3, 4)
11
12 # perform community detection using the Louvain algorithm
13 cd = nk.community.detectCommunities(G, algo=nk.community.PLM(G))
14
15

```

```

Communities detected in 0.00455 [s]
solution properties:
-----
# communities      2
min community size  2
max community size  3
avg. community size 2.5
imbalance          1
edge cut           2
edge cut (portion) 0.4
modularity         0.08
-----

```

▼ Parallelism

The degree of parallelism can be controlled and monitored in the following way:

```

1 print(nk.setNumberOfThreads(4)) # set the maximum number of available threads

None

1 print(nk.getMaxNumberOfThreads()) # see maximum number of available threads

4

1 print(nk.getCurrentNumberOfThreads()) # the number of threads currently executing

```

1

▼ Clustering Coefficient

```
1 # create a small undirected graph with 5 nodes and 5 edges
2 G = nk.Graph(5)
3 G.addEdge(0, 1)
4 G.addEdge(0, 2)
5 G.addEdge(0, 3)
6 G.addEdge(1, 2)
7 G.addEdge(3, 4)
8
9 # calculate the clustering coefficient of each node
10 cc = nk centrality.LocalClusteringCoefficient(G)
11 print("Clustering coefficient of each node:", cc.run().scores())
12
```

Clustering coefficient of each node: [0.3333333333333333, 1.0, 1.0, 0.0, 0.0]

1

Colab notebook completed at 7:15 AM



0s completed at 7:15 AM

