

Accelerating Graph Analysis on GPUs

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What is a Graph?

- A data structure to model relationships between data entities in a network
 - \circ $\,$ Nodes (Vertices) and edges $\,$
 - Directed, undirected, weighted, etc.







What is Graph Analytics?

- Graph Analytics helps understand complex relationships between linked entities
 - Centrality analysis
 - Community detection
 - Connectivity analysis
 - Path analysis
- Combining with ML/AI techniques, e.g GML, GNN, etc.
 - Graph classification
 - Node classification
 - Link prediction
 - etc.

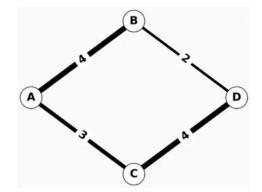
NET





<u>NetworkX</u>

 A Python package for the creation, manipulation, and study of the structure, dynamics, and functions of complex networks



- The most popular python graph analytic
 library available
 G = nx.Graph()
 G.add_edge("A", "B", "
 - More than 40M PyPI downloads per month

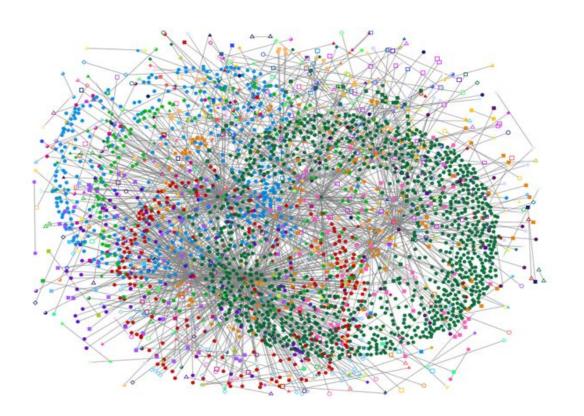
G.add_edge("A", "B", weight=4) >>> G.add_edge("B", "D", weight=2) G.add_edge("A", "C", weight=3) >>> G.add_edge("C", "D", weight=4) >>> nx.shortest_path(G, "A", "D", weight="weight") ['A', 'B', 'D']



When dataset / graph sizes grow ...



R C N E T[™]





cuGraph

- Accelerating Graph analysis on GPUs
- With API similar to NetworkX
 - creating a graph
 - finding influencers
 - finding communities
 - exploring a graph
 - etc.
- Part of the RAPIDS suite
- Scalable
- <u>Supported algorithms</u>

C N F T[™]

```
# call an algorithm
gdf_page = cugraph.pagerank(G)
```



CuGraph: NetworkX Compatibility

- Mimic NetworkX API
- Support NetworkX graph objects
- Suggestions
 - Replacing existing code with cuGraph as much as possible
 - Or simply update the calls to graph algorithms with cuGraph
- <u>Differences in algorithms</u>

import networkx as nx
<pre># create a random graph G = nx.barabasi_albert_graph(N,M) do some NetworkX stuff</pre>
<pre> # call nx algorithms bc = nx.betweenness_centrality(G)</pre>

import networkx as nx
import cugraph as cnx

create a random graph
G = nx.barabasi_albert_graph(N,M)
... do some NetworkX stuff ...
...
call cugrpah algorithms

bc = cnx.betweenness_centrality(G)



nx-cugraph

- Using cuGraph as a backend to NetworkX on GPUs
 - Connects pylibcugraph and CuPy to NetworkX's API
 - pylibcugraph: a python wrapper around cuGraph low-level CUDA-based API
 - <u>CuPy</u>: a GPU-accelerated array library
- Setting an environment variable with zero code changes
 - Running on GPU when cuGraph and an algorithm is supported
 - Otherwise, falling back to CPU-based NetworkX
- Supported Algorithms





Enabling nx-cugraph backend

Via an environment variable with zero code changes

NETWORKX_AUTOMATIC_BACKENDS=cugraph python my_networkx_script.py

Via a keyword argument in function calls

```
import networkx as nx
. . .
nx.betweennees centrality(G, k=1000, backend="cugraph")
```

Via a type-based dispatching

```
import networkx as nx
import nx cugraph
```

NFT

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

• • <u>•</u>

nxcg_G = nx_cugraph.from_networkx(G) # Graph type conversion nx.betweenness centrality(nxcg G, k=1000)

using cugraph backend



Work on the clusters

- Building an Apptainer container from a RAPIDS Docker container
- Both cuGraph and NetworkX are included in a RAPIDS container
 - Docs wiki: <u>https://docs.alliancecan.ca/wiki/RAPIDS</u>
- nx-cugraph needs to be added to a RAPIDS container
 - CUDA 11.2 or up, Python 3.9 or up, and NetworkX v.3.2 or up
 - \circ ~ or RAPIDS v. 23.10 or up





Adding nx-cugraph to a RAPIDS container

- Select a RAPIDS docker container from NVIDIA
- Build an Apptainer sandbox for RAPIDS
- Install nx-cugraph in the sandbox
- Convert the sandbox into an Apptainer image

```
$ apptainer build --sandbox rapids-sandbox docker://<rapids-docker-image-tag>
$ sudo apptainer shell --writable rapids-sandbox
Apptainer> source /opt/conda/etc/profile.d/conda.sh
Apptainer> conda install -c rapidsai-nightly -c conda-forge -c nvidia nx-cugraph
Apptainer> exit
$ apptainer build rapids-nx-cugraph.sif rapids-sandbox
```

Note: Above steps need to be done on your own computer (<u>https://apptainer.org/docs/user/latest/</u>) Submit a ticket (<u>help@sharcnet.ca</u>) for help if needed.

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A RAPIDS-nx-cugraph container

- Based on a RAPIDS docker container from <u>NVIDIA</u>:
 - Docker image tag: nvcr.io/nvidia/rapidsai/notebooks:24.02-cuda11.8-py3.10
 - RAPIDS v.24.02 with a notebook server on Ubuntu 20.04
 - Working with CUDA 11.8 and Python 3.10
- Image file, rapidsas-24.02-nx-cugraph.sif, is available at:
 - <u>https://staff.sharcnet.ca/jhqin/GIS-cuGraph/</u>





Demo example

A large graph

- a citation graph of a <u>U.S. patent dataset</u>
- \circ ~ 4 million nodes, and 16+ million edges
- Compute *betweenness centrality* with approximation
- Comparing the performance
 - NetworkX vs. cuGrapah vs. nx-cugraph





Betweenness-Centrality

A measure of the relative importance of a node or an edge in a graph

 counting the number of shortest paths that pass through a node (or an edge) vs total number of shortest paths for all node pairs

$$BC(v) = \sum_{s,t \in V} \frac{\sigma(s,t \mid v)}{\sigma(s,t)}$$

• Function in NetworkX

nx.betweenness_centrality(G, k, ...)

- k, int, optional (default=None); $k \le$ the total number of nodes;
- higher k gives better approximation



Code examples: NetworkX vs cuGraph

nx-bc-demo.py
import sys
import time
import networkx as nx
import pandas

k = int(sys.argv[1])

Creating Graph from Pandas DataFrame edgelist...
G = nx.from_pandas_edgelist(pandas_edgelist, source="src",
 target="dst", create_using=nx.DiGraph)

Calculating betweenness_centrality

st = time.time()
bc_result = nx.betweenness_centrality(G, k=k)
print(f"BC time with {k=} was: {(time.time() - st):.2f} s")

CNFT[™]

cg-bc-demo.py import sys import time import cugraph as cg import cudf

k = int(sys.argv[1])

Reading dataset into CuDF DataFrame as an edgelist...

cudf_edgelist = cudf.read_csv("cit-Patents.txt", skiprows=4, delimiter="\t", names=["src", "dst"], dtype={"src": "int32", "dst": "int32"})

Creating Graph from cuDF DataFrame edgelist...

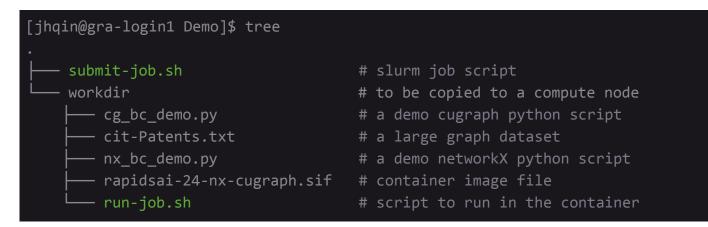
Calculating betweenness_centrality

st = time.time()
bc_result = cg.betweenness_centrality(G, k=k)
print(f"BC time with {k=} was: {(time.time() - st):.2f} s")



Submit a batch job

Demo directory contents:



Submit job:

[jhqin@gra-login1 Demo]\$ sbatch submit-job.sh



Submit a batch job

Slurm script:
submit-job.sh

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#!/bin/bash	
#SBATCHgres=gpu:1	# gpu request
#SBATCHcpus-per-task=N	# number of CPU cores
#SBATCHmem=M	# host memory
#SBATCHtime=DD-HH:MM	# execution time
#SBATCHaccount=def-user	# project account
module load StdEnv/2023 apptainer src_dir=/workdir cp -r \$src_dir \$SLURM_TMPDIR	# copy workdir to local disk on compute node
<pre>work_dir=\$SLURM_TMPDIR/workdir cd \$work_dir container=rapidsai-24-nx-cugraph.si</pre>	if # included in workdir
apptainer execnv \$container \$wo	ork_dir/run-job.sh



Submit a batch job

Script to run in the container

run-job.sh

Note: run-job.sh should be *executable*. #!/bin/bash
source /opt/conda/etc/profile.d/conda.sh

nvidia-smi

echo "======networkX test======="
python nx_bc_demo.py 10
python nx_bc_demo.py 50

echo "=====cuGraph test======"
python cg_bc_demo.py 10
python cg_bc_demo.py 50
python cg_bc_demo.py 500

echo "======nx-cugraph test======="

NETWORKX_AUTOMATIC_BACKENDS=cugraph python nx_bc_demo.py 10 NETWORKX_AUTOMATIC_BACKENDS=cugraph python nx_bc_demo.py 50 NETWORKX_AUTOMATIC_BACKENDS=cugraph python nx_bc_demo.py 500

[jhqin@gra-login1 Demo]\$ chmod +x workdir/run-job.sh



Reference

- CuGraph documentation: <u>https://docs.rapids.ai/api/cugraph/stable/</u>
- CuGraph notebooks: <u>https://github.com/rapidsai/cugraph/tree/main/notebooks</u>
- RAPIDS 23.10 Release: <u>https://medium.com/rapids-ai/rapids-23-10-release-075aa5a50570</u>
- RAPIDS Docker Containers: <u>https://catalog.ngc.nvidia.com/orgs/nvidia/teams/rapidsai/containers/notebooks</u>
- NVIDIA Technical Blog (Data Science): <u>Accelerating NetworkX on NVIDIA GPUs for High Performance Graph</u> <u>Analytics, Nov. 2023 by Rick Ratzel</u>
- NVIDIA Technical Blog (Data Science): <u>Beginner's Guide to GPU Accelerated Graph Analytics in Python, Mar.</u>
 <u>2021 by Tom Drabas</u>
- Medium Blog: Introduction to Graph Analysis using cuGraph, Jul. 2023 by Don Acosta
- Medium Blog: Intro to Graph Analysis using cuGraph: Similarity Algorithms, Oct. 2023 by Don Acosta

