

NetworkX Temporal



- **Tutorial:** <https://github.com/nelsonaloysio/c72h>
- **PyPI:** <https://pypi.org/p/networkx-temporal>
- **Documentation:** <https://networkx-temporal.readthedocs.io>
- **Repository:** <https://github.com/nelsonaloysio/networkx-temporal>
- **Preprint:** <https://nelsonaloysio.github.io>

NetworkX-Temporal

pypi package

1.2.1

docs

passing

downloads

31k

downloads/month

1k

license

BSD



NetworkX-Temporal extends the [NetworkX](#) library to dynamic graphs, i.e., temporal network data.

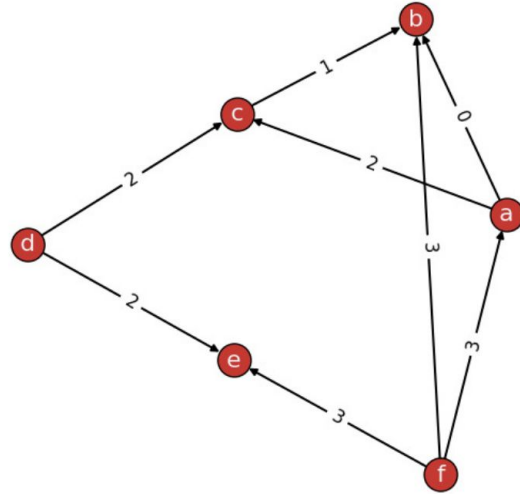
This package provides new `TemporalGraph` classes, which inherit NetworkX's [graph classes](#) and implement additional functions to manipulate temporal data within. Among others, it provides ways to `slice()` a graph into snapshots and `transform` or `convert` it to other libraries and formats.

Install

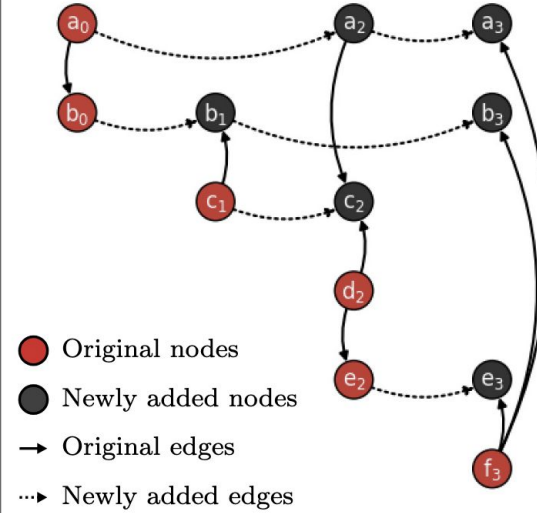
The package supports **Python 3.7+** and is readily available from [PyPI](#):

```
$ pip install networkx-temporal
```

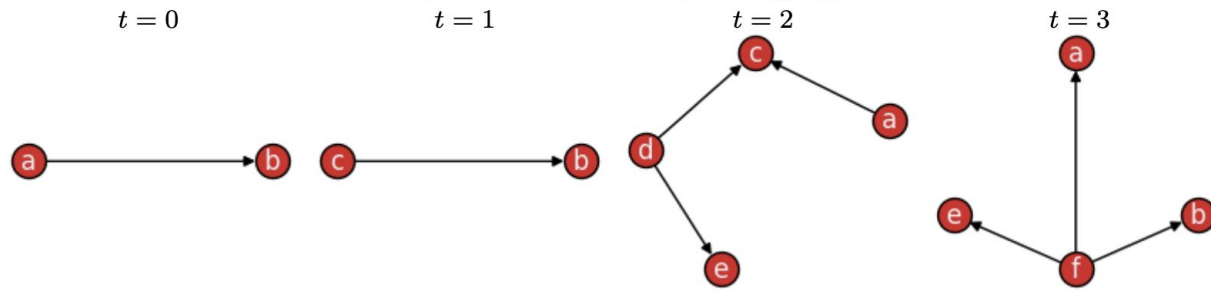
Temporal graph

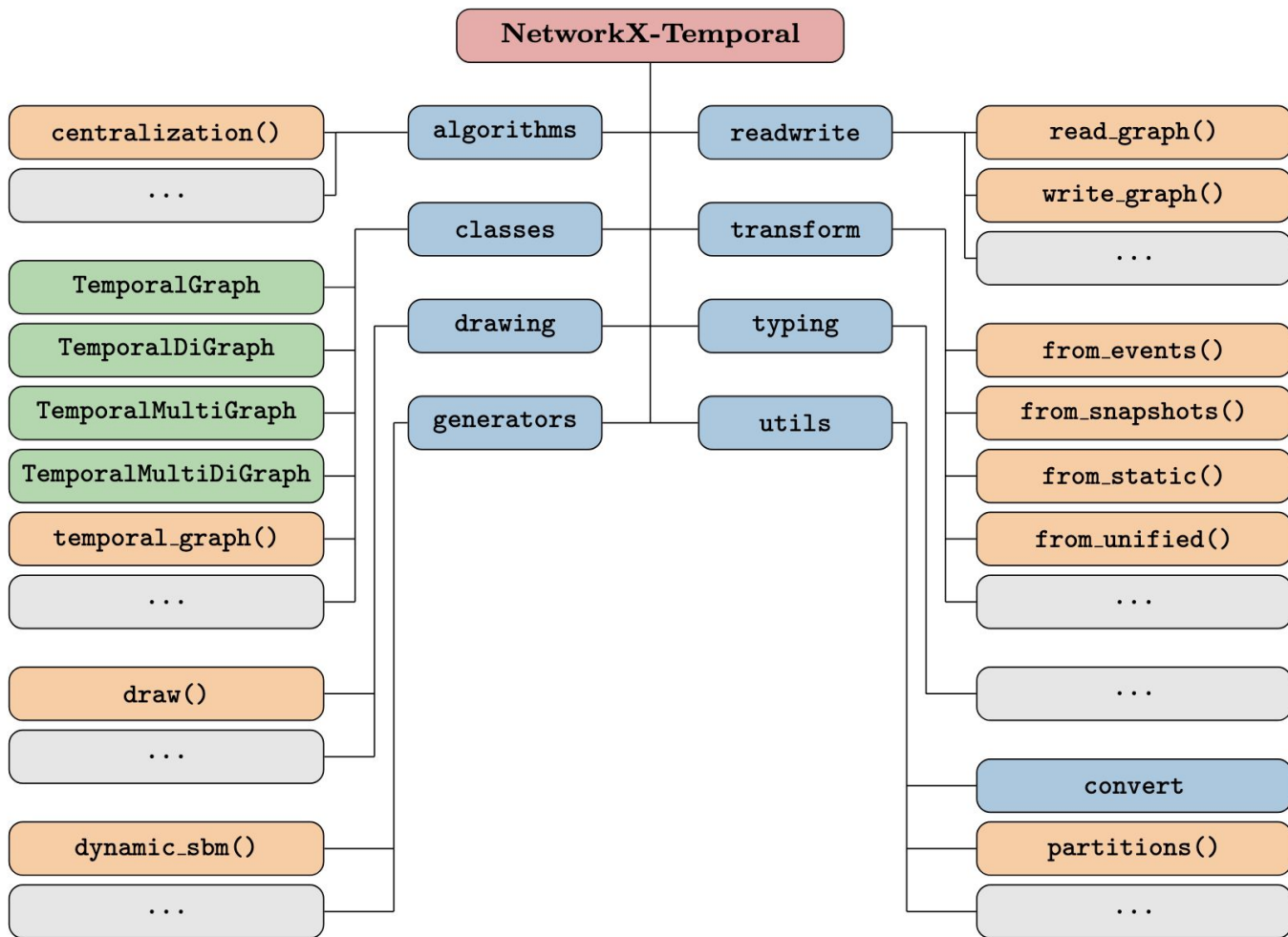


Unrolled temporal graph



Snapshot-based temporal graph





Convert and transform graphs

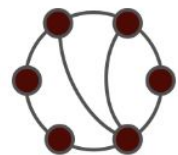


This package allows to transform a `TemporalGraph` between different **graph representations**:

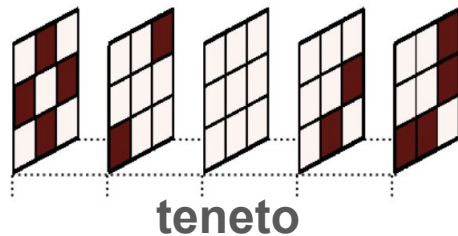
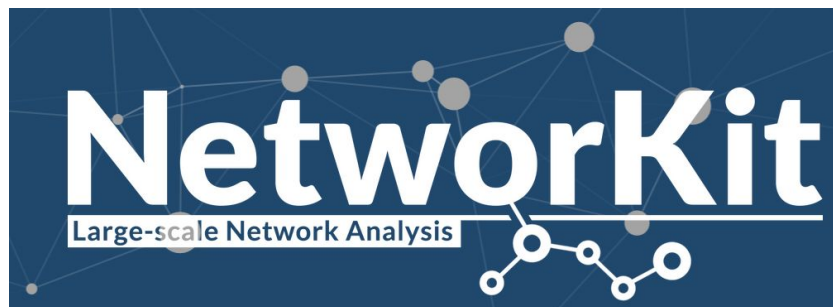
- **Static graphs**: flattened or aggregated version of the temporal graph.
- **Snapshot-based temporal graphs**: a list of node- or edge-level snapshots.
- **Event-based temporal graphs**: a sequence of edge-level events (interactions).
- **Unified temporal graphs**: a single graph with time-stamped nodes and edges.

```
>>> G = TG.to_static()           # TG = tx.from_static(G)
>>> STG = TG.to_snapshots()      # TG = tx.from_snapshots(STG)
>>> ETG = TG.to_events()         # TG = tx.from_events(ETG)
>>> UTG = TG.to_unified()        # TG = tx.from_unified(UTG)
```

In addition, both static and temporal graphs may be converted to the following **graph formats**:



graph-tool





Data Science

English ▾

NetworkX Introduces Zero Code Change Acceleration Using NVIDIA cuGraph

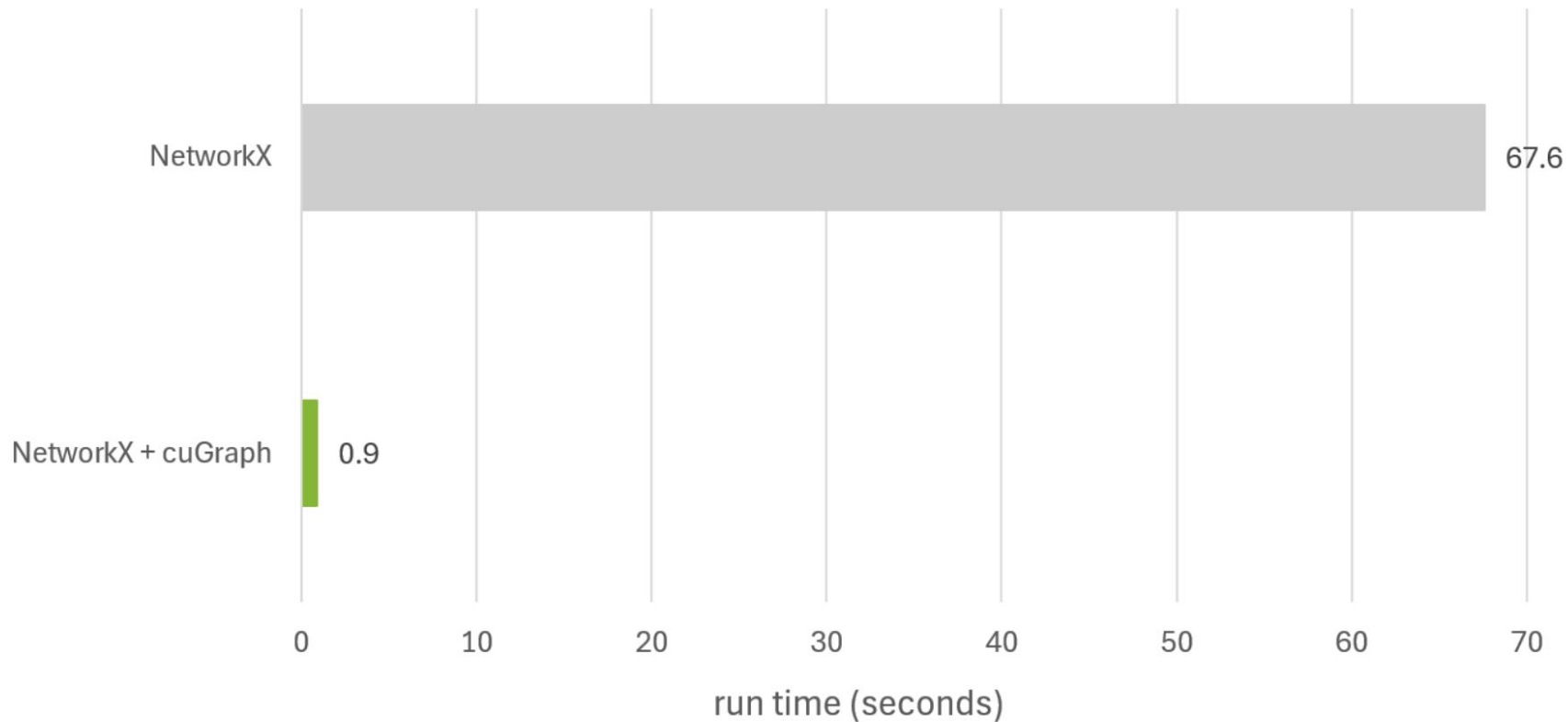
Oct 22, 2024

 +18 Like

 Discuss (0)

By [Rick Ratzel](#) and [Jenn Yonemitsu](#)

PageRank algorithm used to compute values for a citation graph of U.S. patents (4M nodes, 16M edges) is 70x faster than NetworkX on CPU



File: demo.ipynb

```
import pandas as pd
import networkx as nx

url = "https://data.rapids.ai/cugraph/datasets/cit-Patents.csv"
df = pd.read_csv(url, sep=" ", names=["src", "dst"], dtype="int32")
G = nx.from_pandas_edgelist(df, source="src", target="dst")

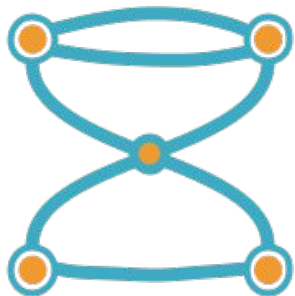
%time result = nx.betweenness centrality(G, k=10)
```

Output:

```
user@machine:~/# ipython demo.ipynb
CPU times: user 7min 36s, sys: 5.22 s, total: 7min 41s
Wall time: 7min 41s

user@machine:~/# NX_CUGRAPH_AUTOCONFIG=True ipython demo.ipynb
CPU times: user 4.14 s, sys: 1.13 s, total: 5.27 s
Wall time: 5.32 s
```

- **Software:** NetworkX 3.4.1, cuGraph/nx-cugraph 24.10
- **CPU:** Intel Xeon Gold 6128 CPU @ 3.40GHz 45GB RAM
- **GPU:** NVIDIA Quadro RTX 8000 50GB RAM



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