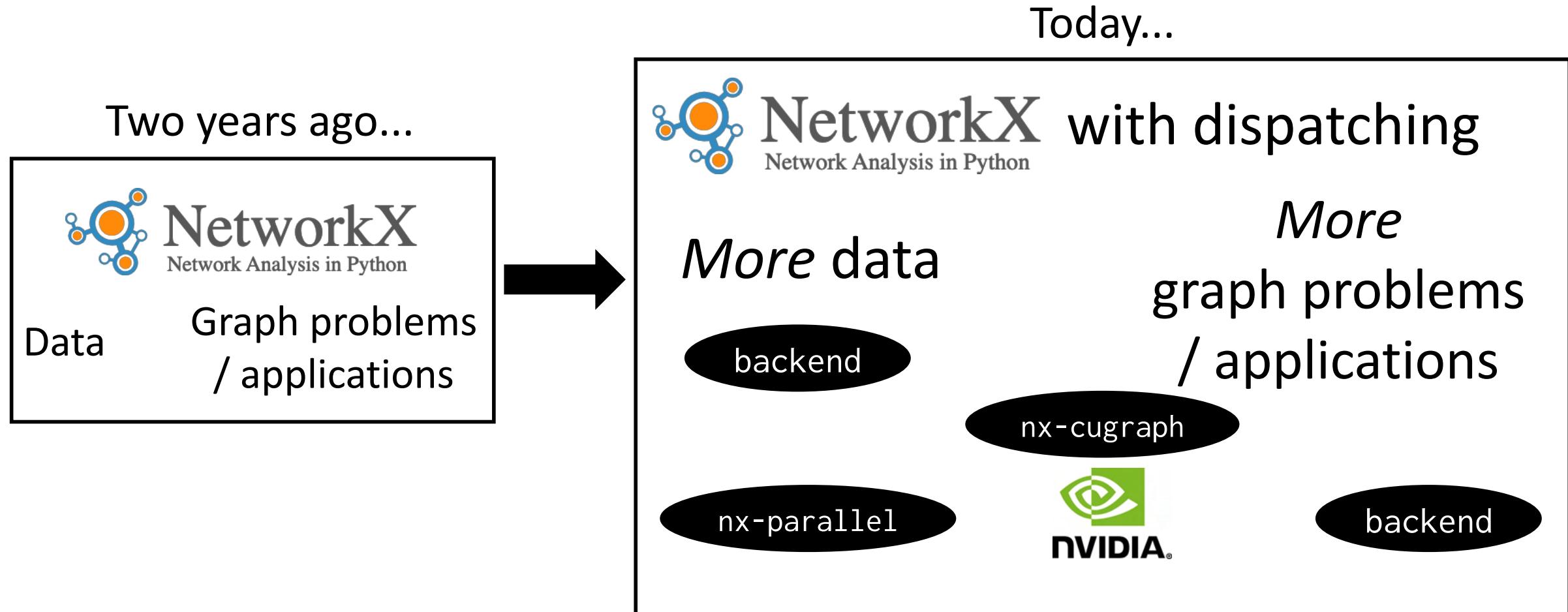


# Understanding NetworkX's API Dispatching with a parallel backend

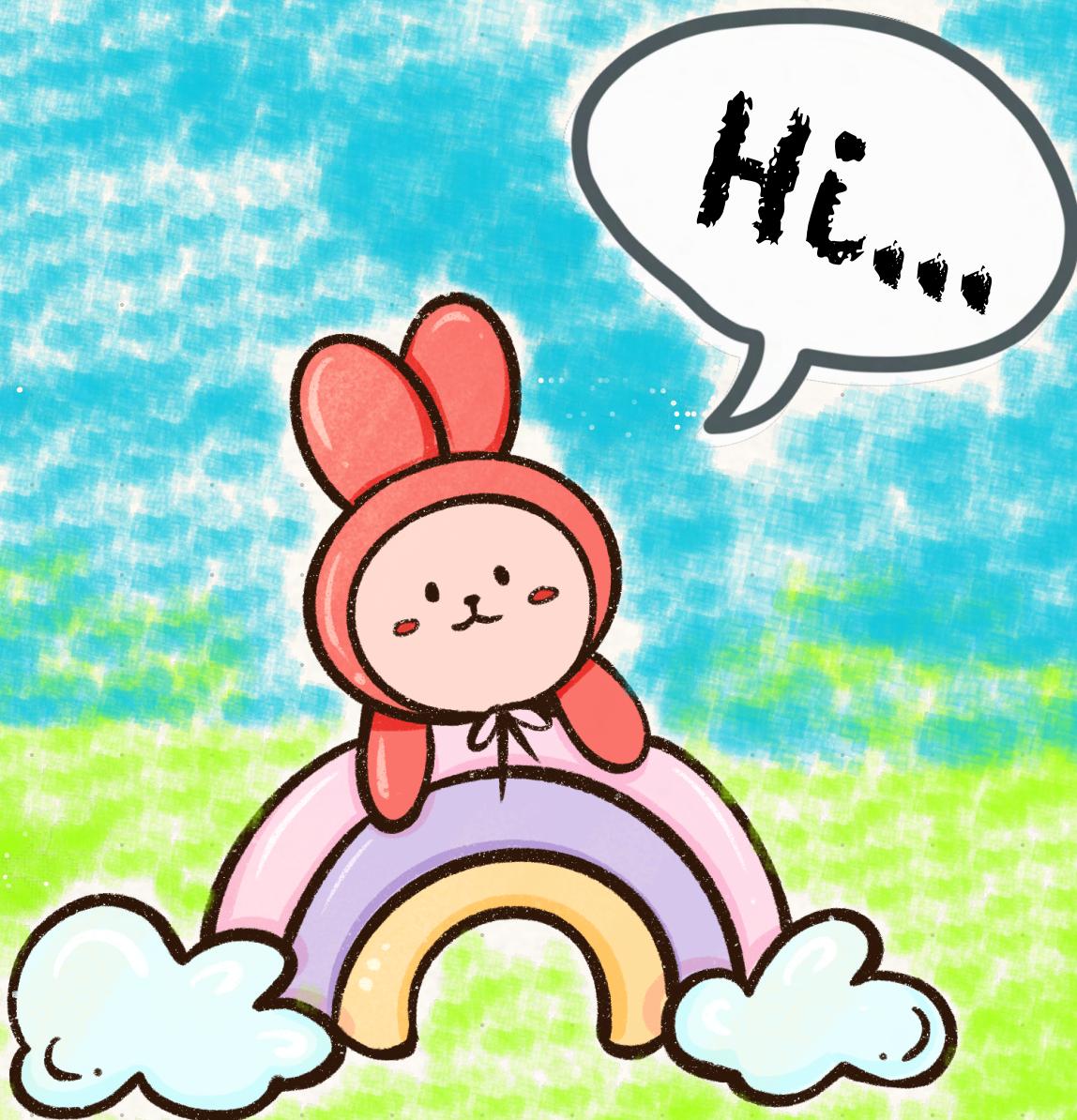
-By Erik Welch and  
Aditi Juneja

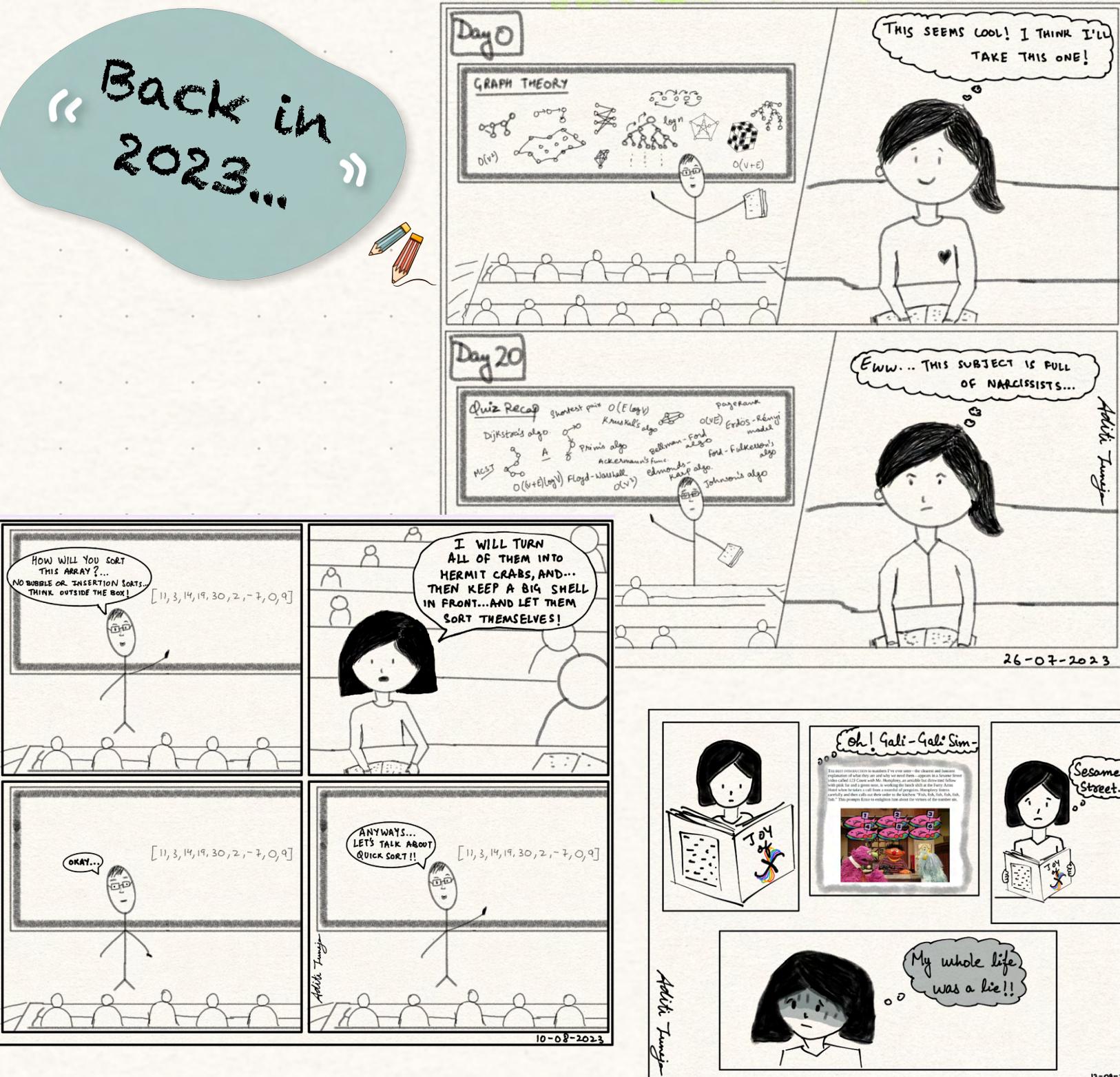
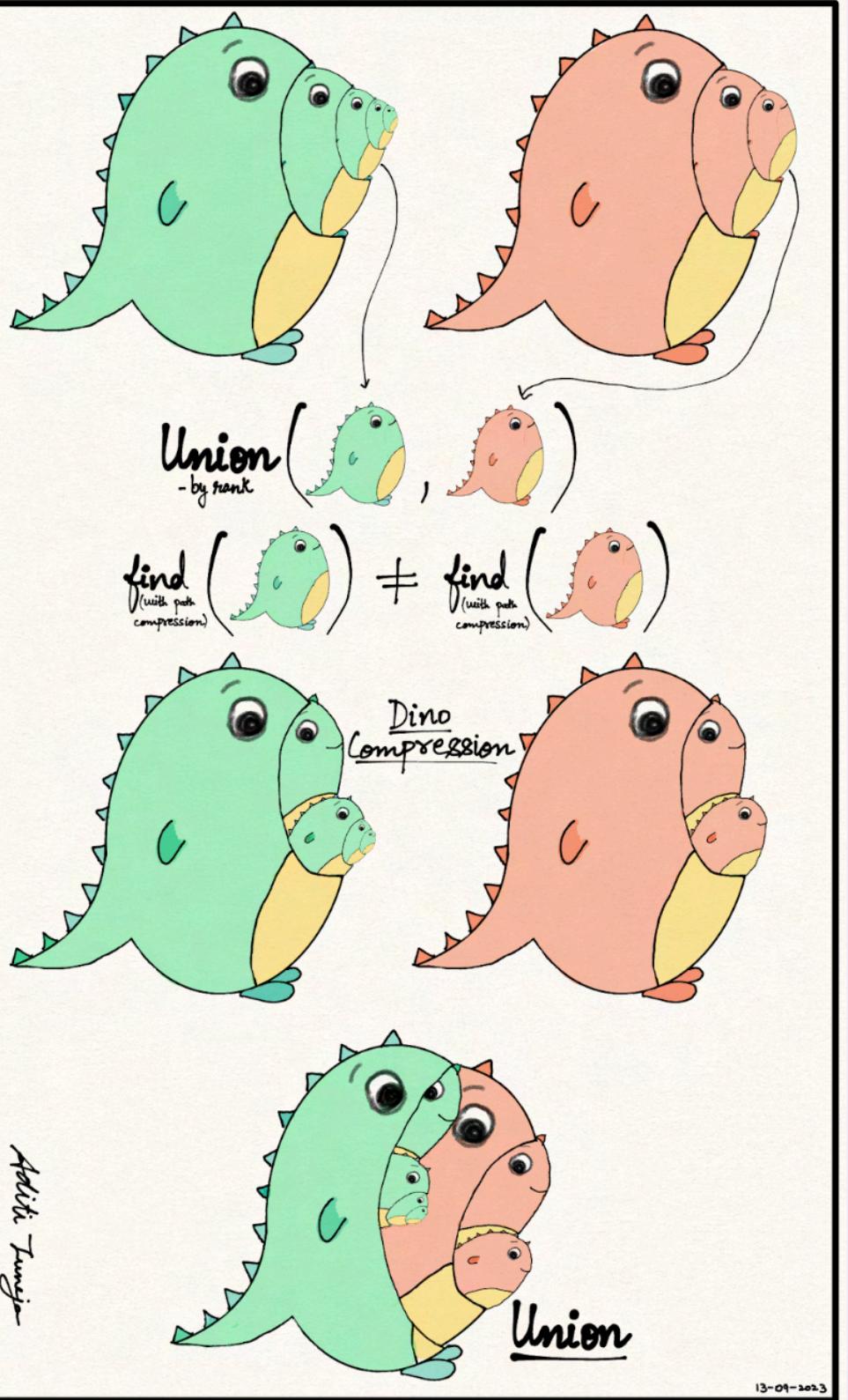
Erik Welch, core NetworkX dev, OSS engineer on NVIDIA **RAPIDS** team



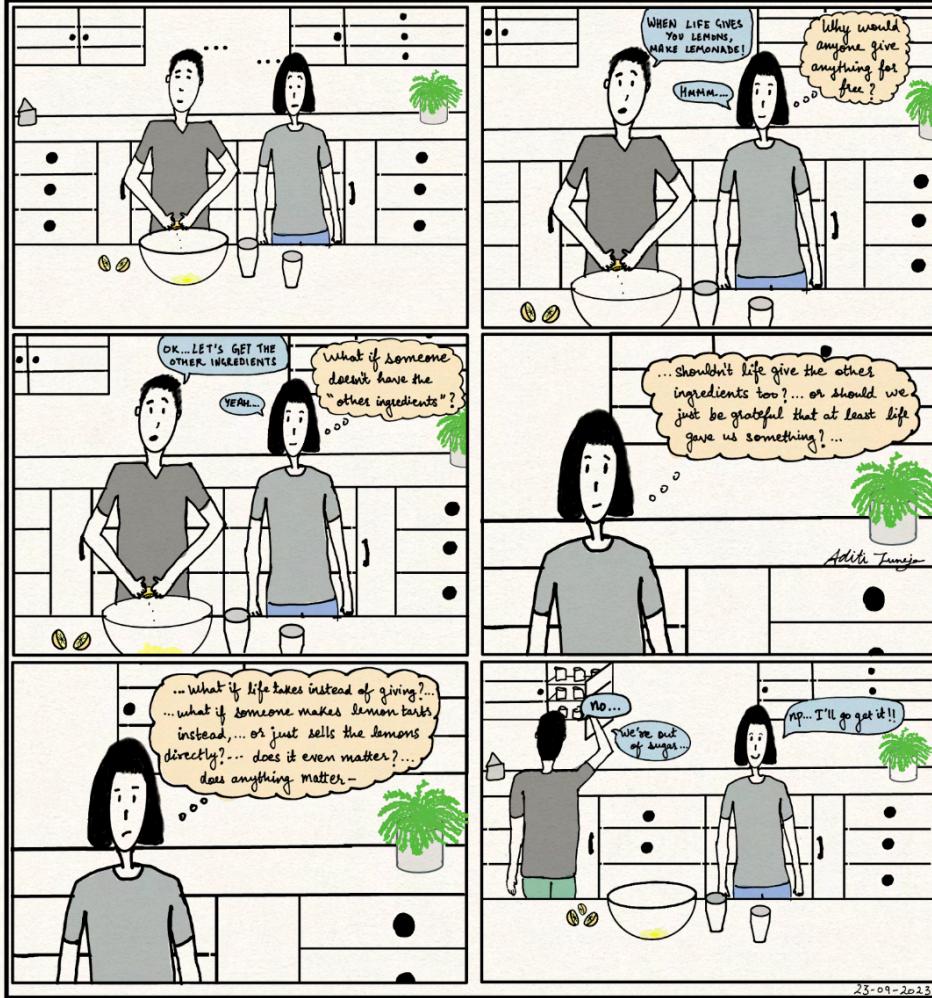
See also: "Dispatching, Backend Selection, and Compatibility APIs"

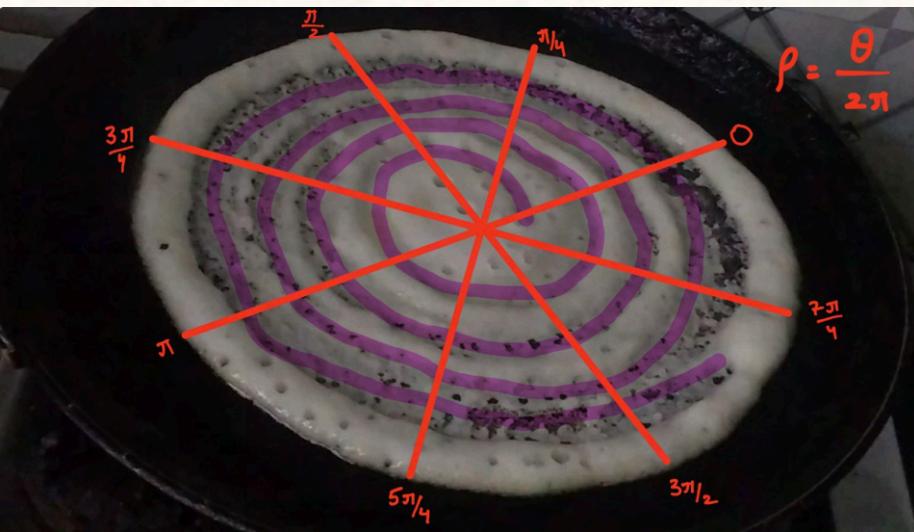
Tomorrow 13:20, Room 5





# "Some more..."

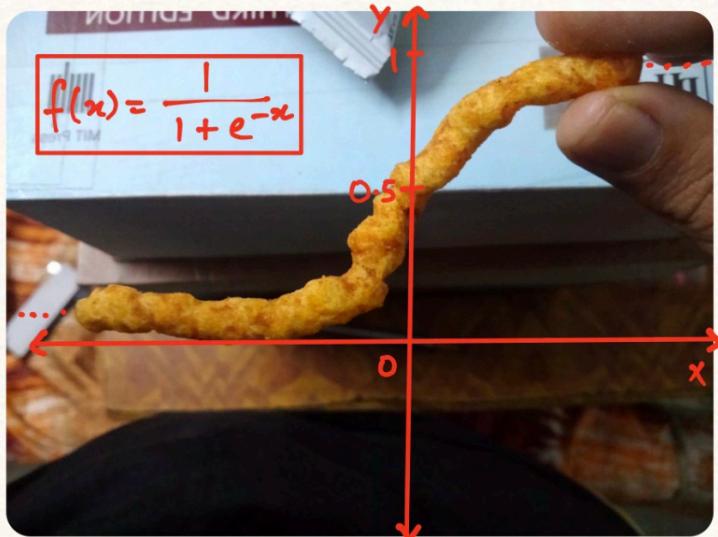




the perfect evening...



# Math-food play



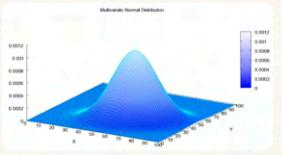
...that's a good-sigmoid-looking Kurkura.



These should be eaten the way they were built - fry by fry, layer by layer, like a seashell on the shore, formed over the years, layer by layer building up to that beautiful, intricate structure.



...looks like a gaussian distribution in 3d (Bivariate normal joint density)



*My last contribution  
to the "graph world"...*

schefflera-arboricola.github.io/Schefflera-Arboricola/Carving-out-a-path-in-a-garden

Blogs About Projects Comics

## Carving out a path in a garden

Published on Sep 11, 2023

There is a beautiful garden composed of several smaller flower gardens. It's soon going to open to the general public. Before that, the gardener wants to carve out a path and pave it with concrete, allowing visitors to stroll around the garden. He wants to put in the least effort and time, while also minimizing the damage to the ground and flowers. He also wants all the flower gardens' centers to be accessible from one another.



*My last contribution  
to the "graph world"...*

schefflera-arboricola.github.io/Schefflera-Arboricola/Carving-out-a-path-in-a-garden

Blogs About Projects Comics

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*But....*

Outreachy

NetworkX

nx-parallel



# SciPy 2024 Poster

## NetworkX Backend Dispatching

Plugin Arch.

entry\_points

Automatic backend testing

For testing:  
NETWORKX\_TEST\_BACKEND=parallel  
NETWORKX\_FALLBACK\_TO\_NX=True  
pytest --pyargs networkx "\$@"

9th July, 2024

### NetworkX

A pure Python library for network analysis

### nx-parallel

A parallel backend for NetworkX, utilizing Joblib to implement graph algorithms on multiple CPU cores.

Backend Interface

- can\_run
- should\_run
- on\_start\_tests
- caching
- logging
- backend priority

config

speed-ups

- local\_efficiency-**8.8x**
- betweenness\_centrality-**6.4x**
- square\_clustering-**4.1x**
- bipartite.node\_redundancy-**4.1x**

\*for 300-node random graphs (edge probability = 0.6)

feel free to contribute and nurture!

Aditi Taneja (@Schefflera-Arboicola)

## Usage:

```
>>> import nx_parallel as nxp
>>> nx.betweenness_centrality(G,
>>> backend="parallel")
>>> H = nxp.ParallelGraph(G)
>>> nx.betweenness_centrality(H)

$ export NETWORKX_AUTOMATIC_BACKEND=parallel
&& python nx_code.py
```

## To-Do

- Designing the pipeline nicely
- Benchmarking
- Adding distributed graph algorithms

### Joblib

### Loky

### Threading

### Multiprocessing

### Dask, Ray...

# 2 days ago...

## Google Summer of Code 2024 Final Report

(May 2024 - August 2024)

Project Title - [Revisiting and expanding nx-parallel](#)

Organisation - NumFOCUS(NetworkX : nx-parallel)

Mentors - [Dan Schult](#), [Mridul Seth](#)

Contributor - [Aditi Juneja](#)

### 1. Abstract

This report details the work accomplished during Google Summer of Code 2024, focusing on the nx-parallel project, a parallel backend for NetworkX that leverages [joblib](#) for running graph algorithms in parallel. Expanding the nx-parallel project involved integrating the [joblib](#) and the networkx's configuration systems in nx-parallel and adding a pre-commit hook to automate the [get\\_info](#) updation. Additionally, the project involved switching to setuptools for the build tool, revisiting the previously added algorithms and enhancing their performance, enabling custom chunking, adding tests, updating the docs syntax, renaming the 'Dispatcher' class and enhancing the functionalities of the 'ParallelGraph' class. Beyond my GSoC project, I also created nx-parallel's conda-forge feedstock and added the [conda](#) installation guide to the nx-parallel README. Apart from these, there were several notable achievements and contributions made outside the scope of the GSoC project and towards the NetworkX dispatching and the bigger NetworkX project.

The contributions made during this period are crucial for future development, and this report, along with my [blogs](#), will serve as a valuable resource for anyone continuing this work.

### 2. Background

The [nx-parallel](#) package is a backend for NetworkX designed to optimise the execution of graph algorithms through parallel computing using [joblib](#). As an active contributor to the project, I had previously set up the [ASV benchmarking](#)



GSoC 2024 aimed to revisit and expand the nx-parallel project, specifically focusing on integrating the joblib library. These efforts are now complete, and the project is better integrated with the rest of NetworkX.

Check out my [GitHub](#) and [GSoC blogs](#) for more details on my contributions.

ParallelGraph with n parallel chunks: When a pdftohtml command is run on a large section of the ParallelGraph, it generates multiple parallel chunks. This ensures that all chunks are processed simultaneously.

Ensuring that all setup tools are standing by: When a pdftohtml command is run on a large section of the ParallelGraph, it generates multiple parallel chunks. This ensures that all chunks are processed simultaneously.

# DONE!



# **Understanding NetworkX's API Dispatching with a parallel backend**

# Understanding NetworkX's API Dispatching with a parallel backend



*SPEC2 - API  
Dispatching*

# Understanding NetworkX's API Dispatching with a parallel backend

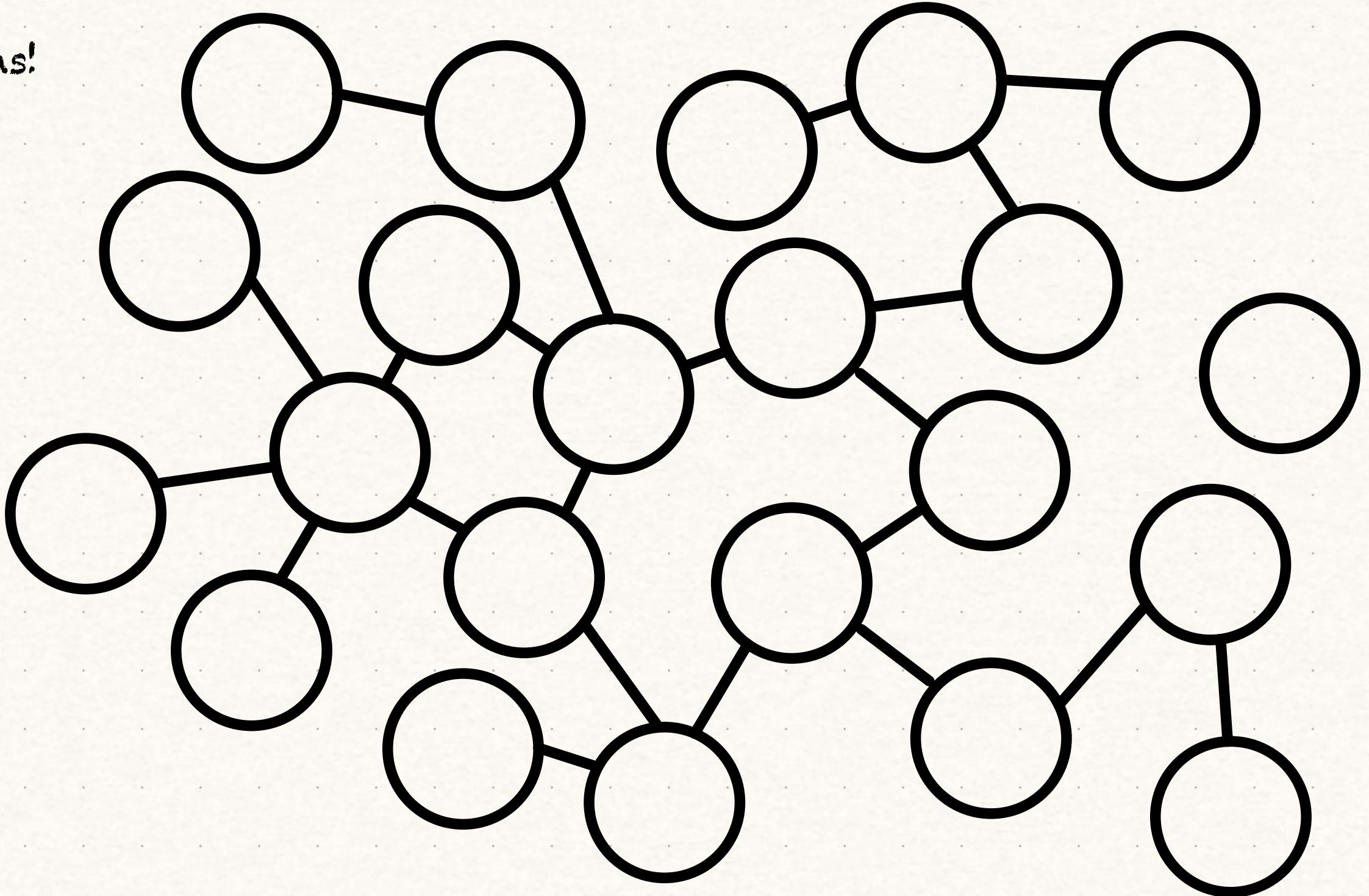
# Understanding NetworkX's API Dispatching with a parallel backend

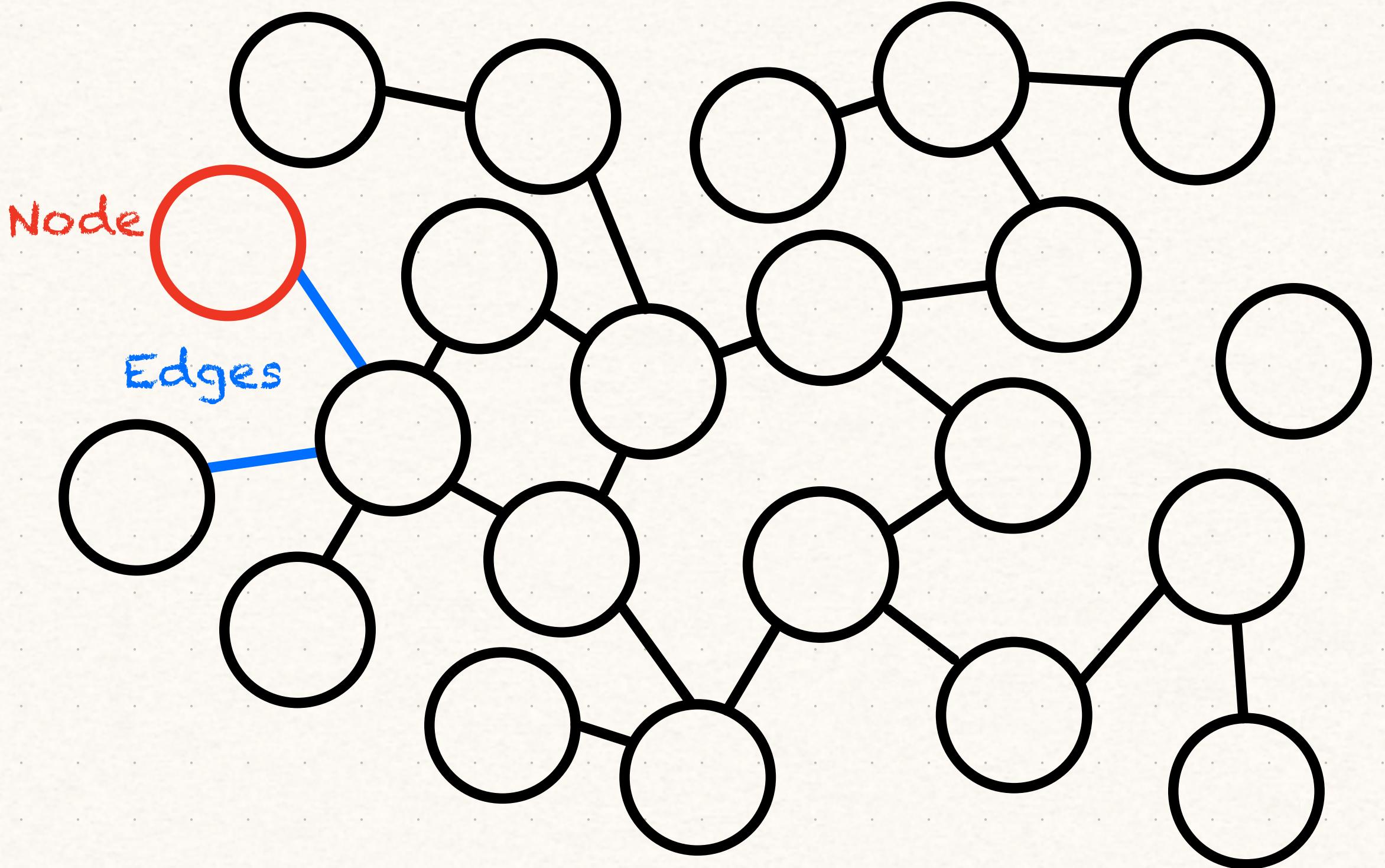
A graph (aka  
network) analysis  
Python library



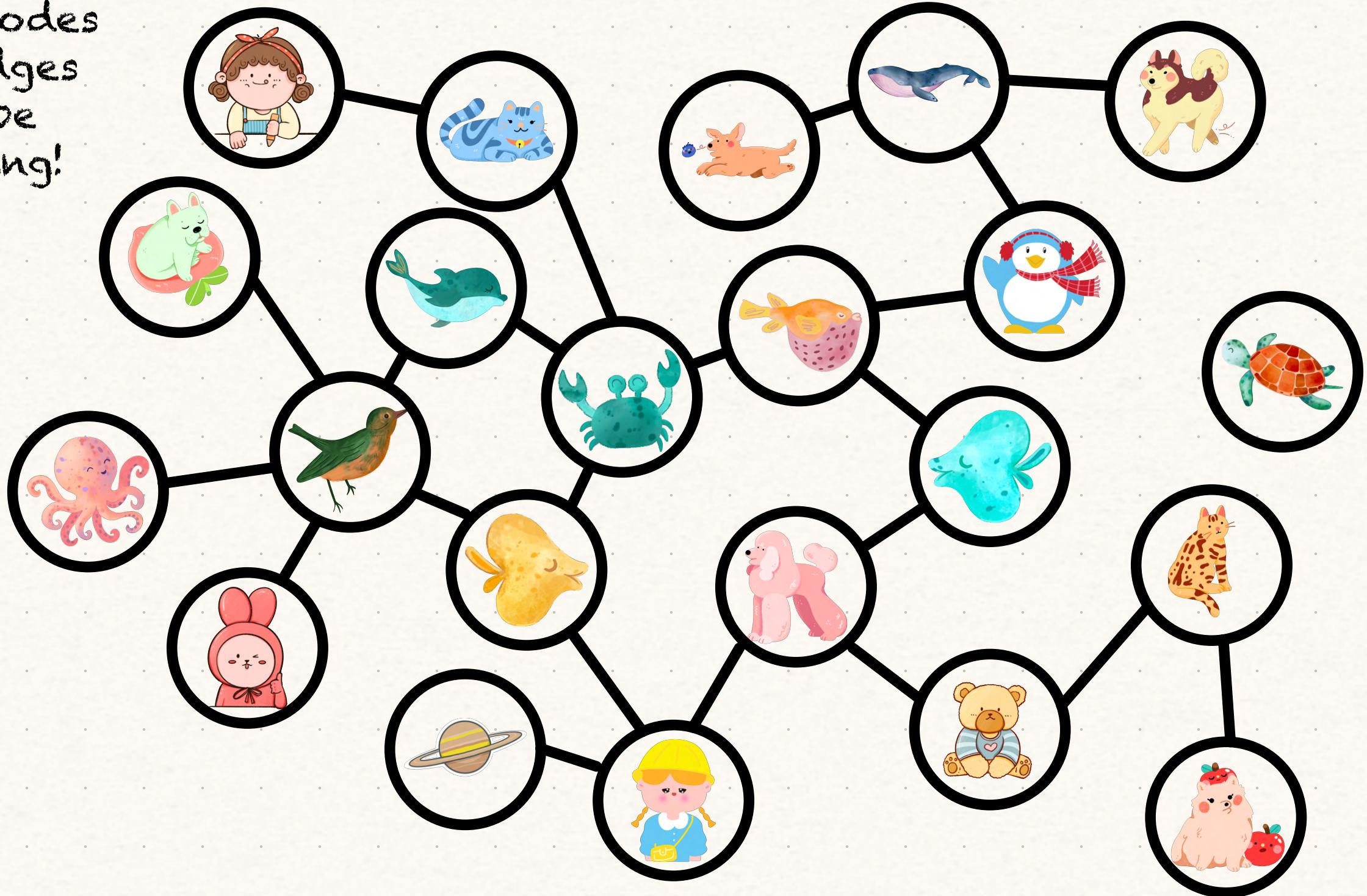
NOT these kind  
of graphs

Graphs!



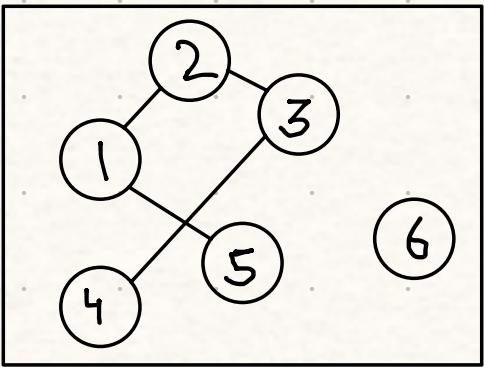


These nodes  
and edges  
can be  
anything!



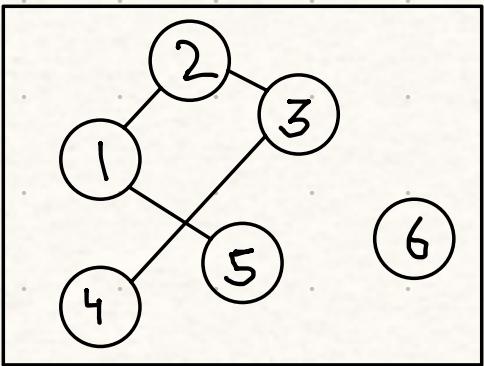
# NetworkX API

```
>>> import networkx as nx  
  
>>> G = nx.Graph()  
  
>>> G.add_edges_from([(1,2), (3,4), (2,3),  
 (5,1)])  
  
>>> G.add_node(6)  
  
>>> nx.betweenness_centrality(G)
```



# NetworkX API

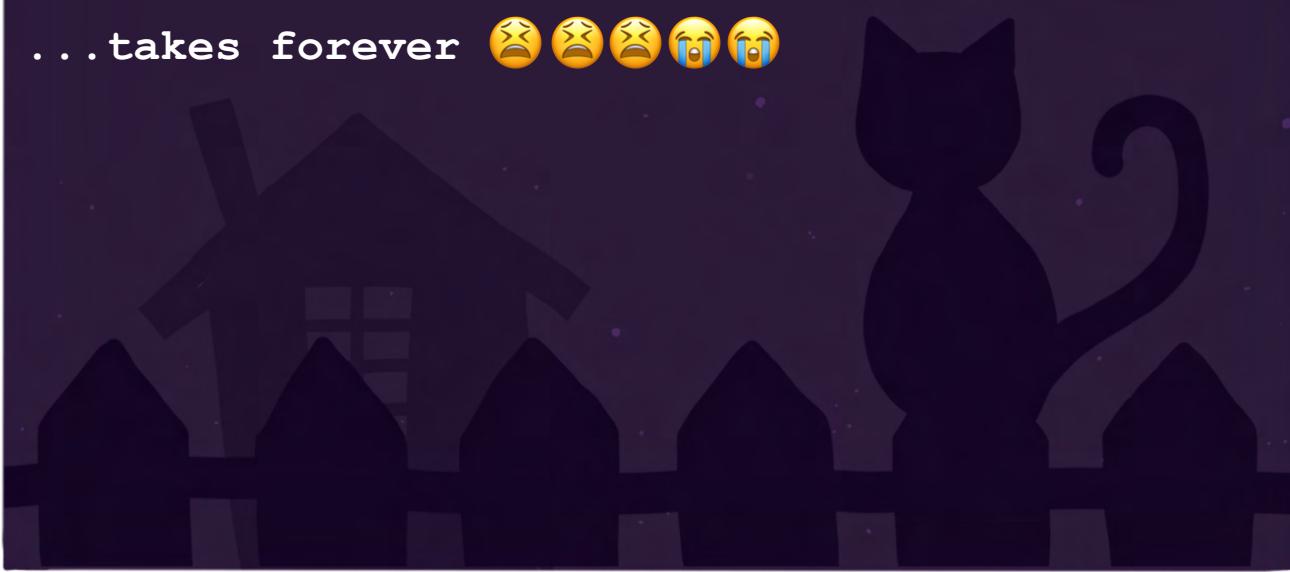
```
>>> import networkx as nx  
  
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 (5,1)])  
  
>>> G.add_node(6)  
  
>>> nx.betweenness_centrality(G)  
  
{1: 0.3000000000000004, 2: 0.4, 3:  
 0.3000000000000004, 4: 0.0, 5: 0.0, 6: 0.0}
```



## NetworkX API

```
>>> import networkx as nx  
  
>>> G = nx.Graph()  
  
>>> G.add_edges_from([(1,2), (3,4), (2,3),  
 (5,1)])  
  
>>> G.add_node(6)  
  
>>> nx.betweenness_centrality(G)  
  
>>> G = nx.fast_gnp_random_graph(1000000, 0.5)  
  
>>> nx.betweenness_centrality(G)  
  
... takes forever 😣😭😭😭😭
```

Now, lets try this...



## NetworkX API

```
>>> import networkx as nx  
  
>>> G = nx.Graph()  
  
>>> G.add_edges_from([(1,2), (3,4), (2,3),  
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```

Now, lets try this...



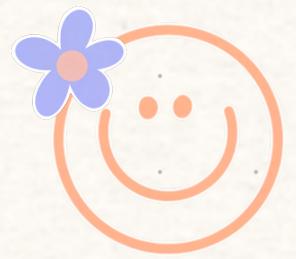
What to do now?

# **Understanding NetworkX's API Dispatching with a parallel backend**

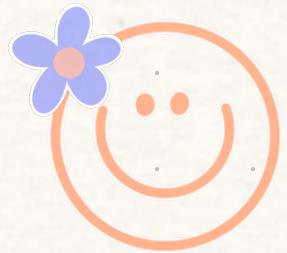
# Understanding NetworkX's API Dispatching with a parallel backend

# Understanding NetworkX's API Dispatching with a parallel backend

“What is  
Dispatching?”

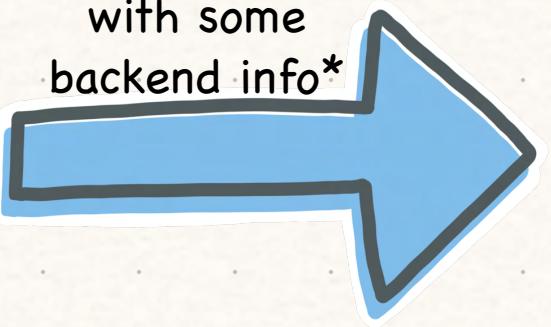


User

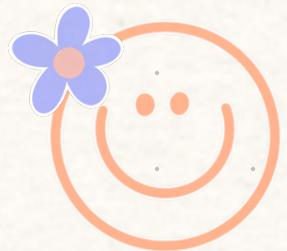


User

function call  
with some  
backend info\*

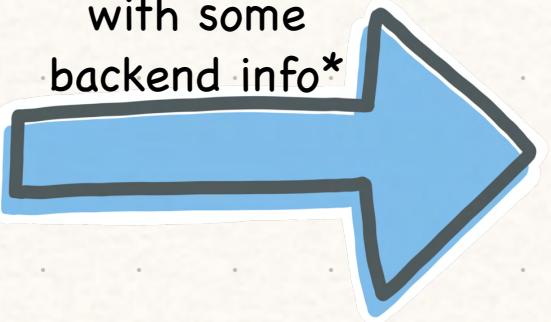


The main  
Library



User

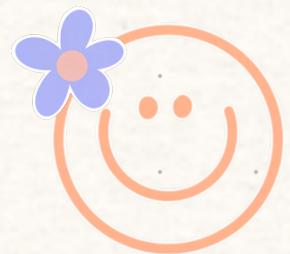
function call  
with some  
backend info\*



## The main Library

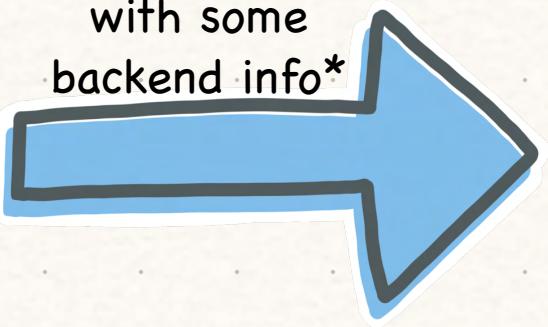


\*This backend information  
might not be entered in  
the function call itself but  
can also be globally  
stored, like as an  
environment variable.

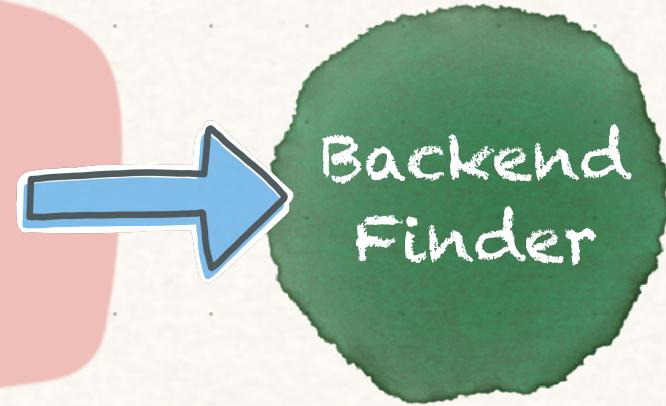


User

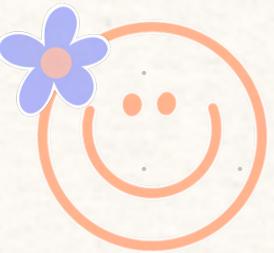
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backend info\*



The main  
Library

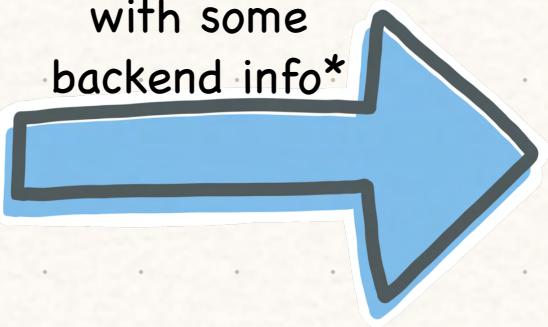


Backend  
Finder



User

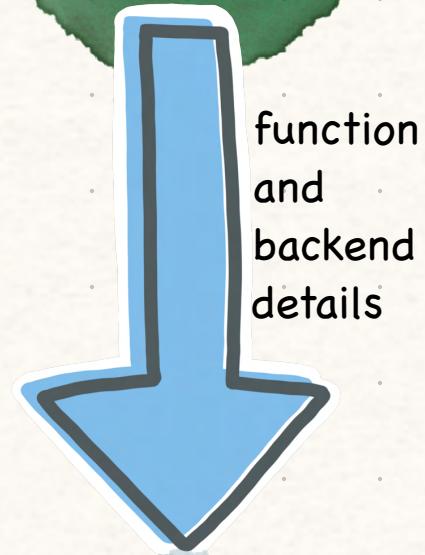
function call  
with some  
backend info\*



The main  
Library

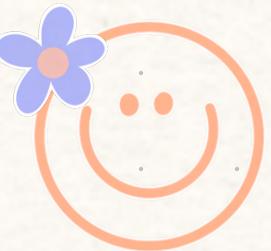


Backend  
Finder



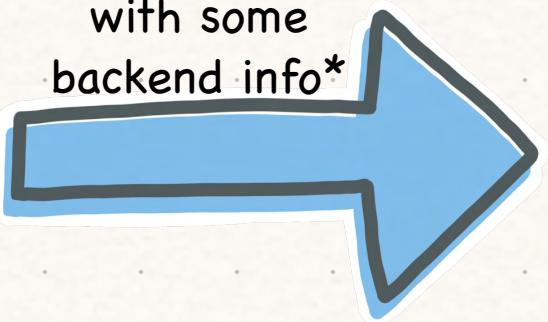
Backend  
translator

function  
and  
backend  
details



User

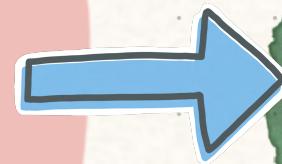
function call  
with some  
backend info\*



The main  
Library

Backend  
Finder

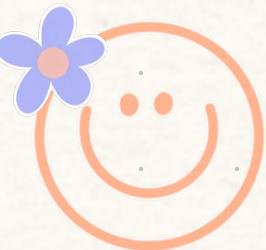
function  
and  
backend  
details



Backend  
translator

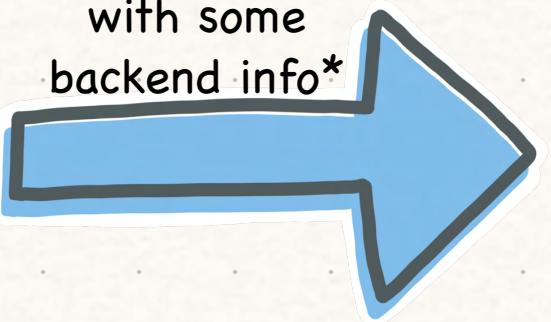


“The backend  
Library ”

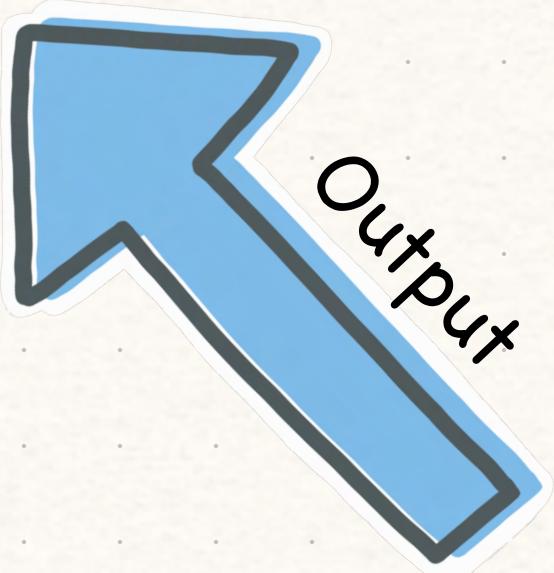


User

function call  
with some  
backend info\*

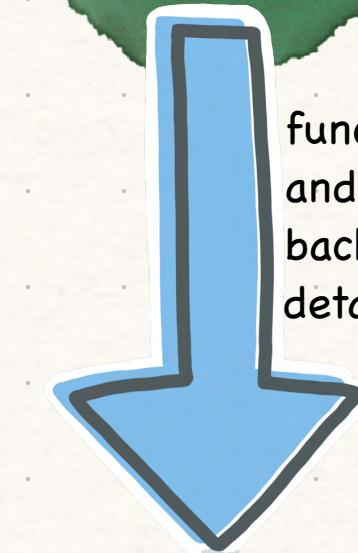
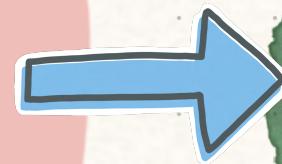


The main  
Library



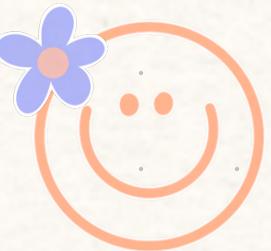
Backend  
Finder

function  
and  
backend  
details



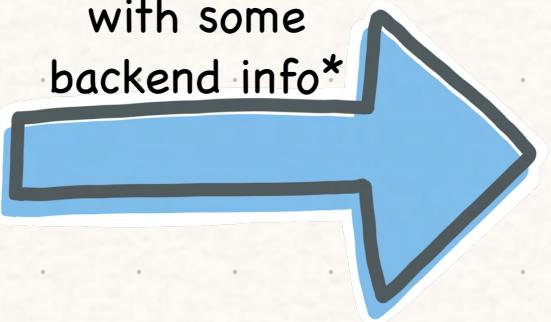
“The backend  
Library ”

Backend  
translator

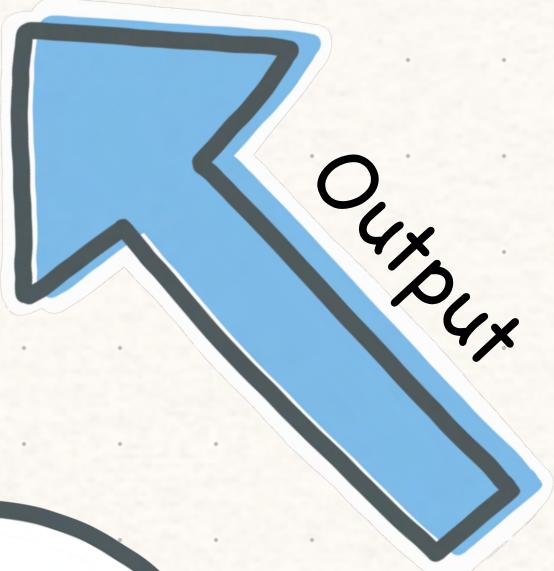


User

function call  
with some  
backend info\*

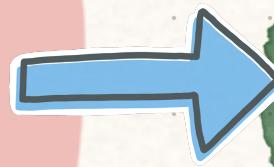


The main  
Library



Backend  
Finder

function  
and  
backend  
details



“The backend  
Library ”

Backend  
translator

But why?

# But why dispatching?

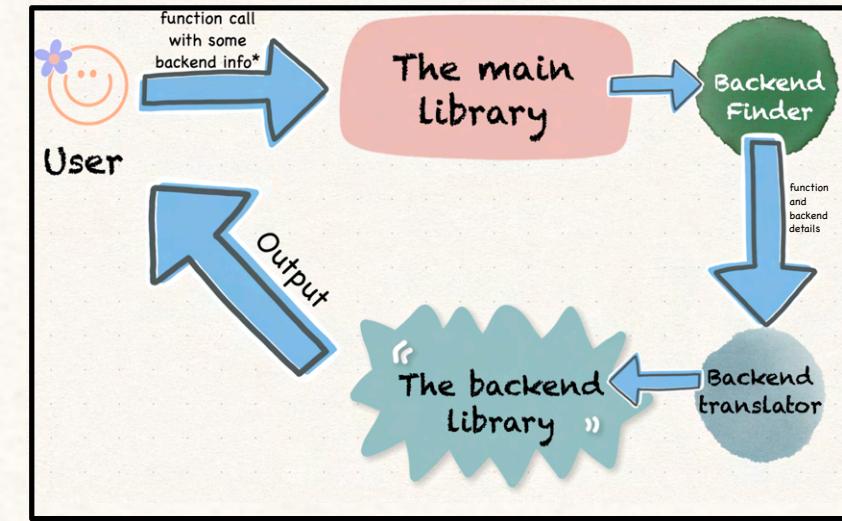
- The API remains similar for the end user, except they just have to pass some additional backend information.
- Is API dispatching the best thing for your package? - It depends...
- NetworkX is a big and old project.

Lets see API  
dispatching in  
NetworkX

# Function call with "some backend information" in NetworkX

1. `backend` kwarg in the function call:

```
nx.betweenness_centrality(G, backend="parallel")
```



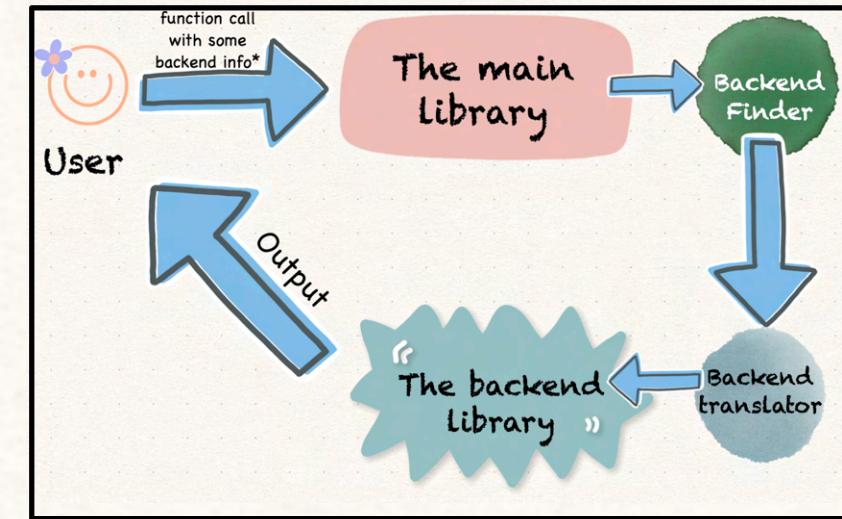
# Function call with "some backend information" in NetworkX

1. `backend` kwarg in the function call:

```
nx.betweenness_centrality(G, backend="parallel")
```

2. Type-based dispatching

```
import nx_parallel as nxp  
H = nxp.ParallelGraph(nx.complete_graph(3))  
nx.betweenness_centrality(H)
```



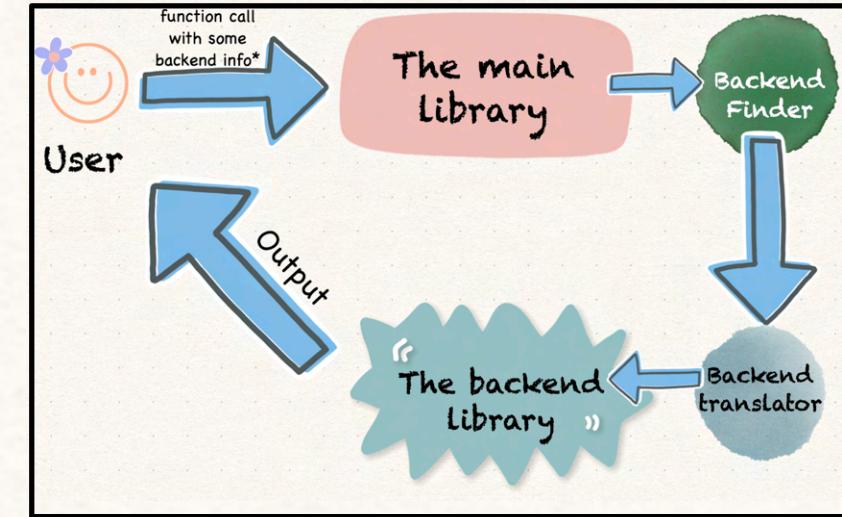
# Function call with "some backend information" in NetworkX

1. `backend` kwarg in the function call:

```
nx.betweenness_centrality(G, backend="parallel")
```

2. Type-based dispatching

```
import nx_parallel as nxp  
H = nxp.ParallelGraph(nx.complete_graph(3))  
nx.betweenness_centrality(H)
```



```
class ParallelGraph:  
    __networkx_backend__ = "parallel"  
  
    def __init__(self, ...):  
        ....
```

# Function call with "some backend information" in NetworkX

1. `backend` kwarg in the function call:

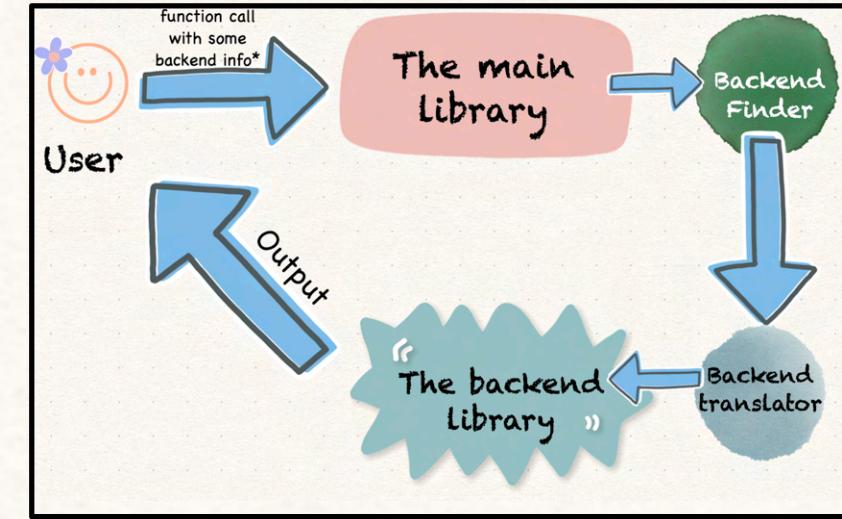
```
nx.betweenness_centrality(G, backend="parallel")
```

2. Type-based dispatching

```
nx.betweenness_centrality(H)
```

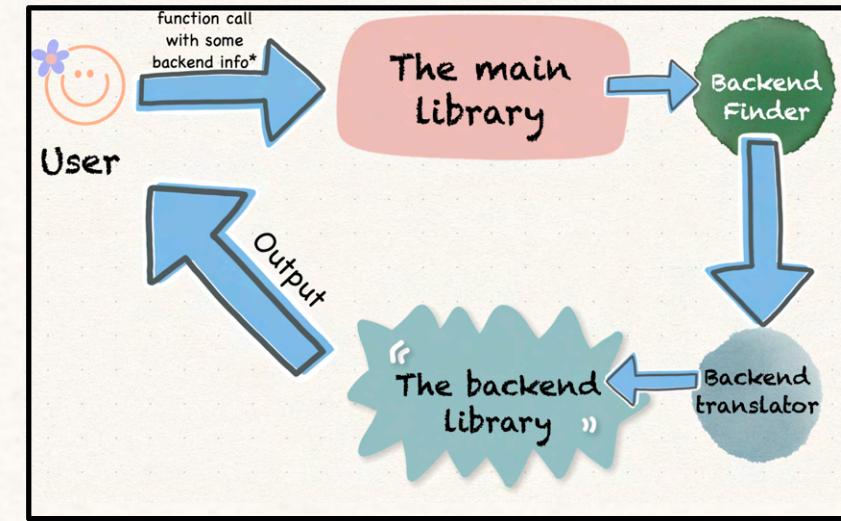
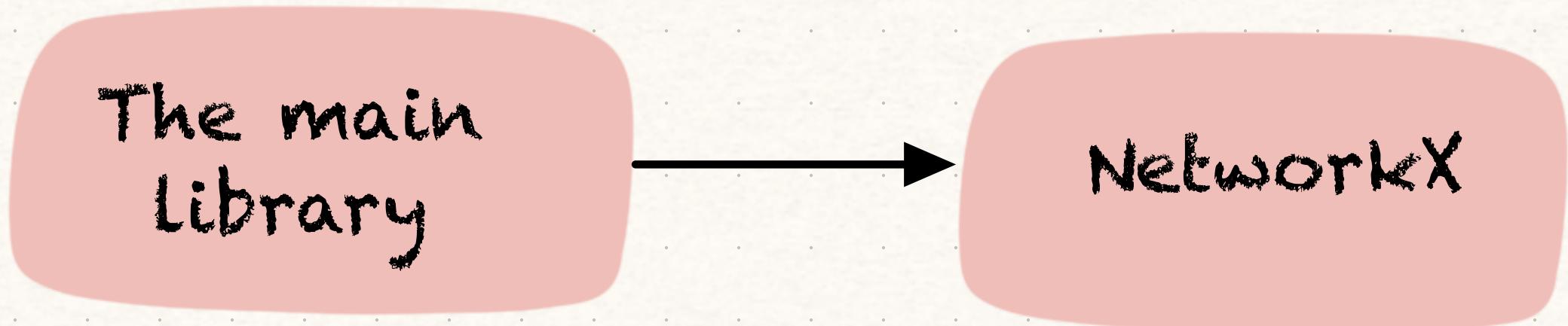
3. Using networkx's configurations

```
with nx.config(backend_priority=['parallel',]):  
    nx.square_clustering(G)
```



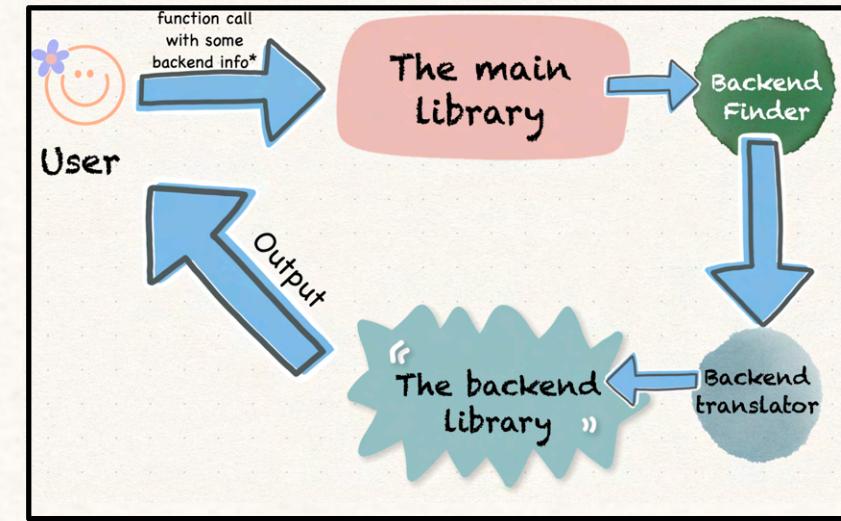
`@nx.\_dispatchable` decorator's job:

To identify "some backend info" and accordingly redirect the call to the specified backend's implementation.



# Backend Finder

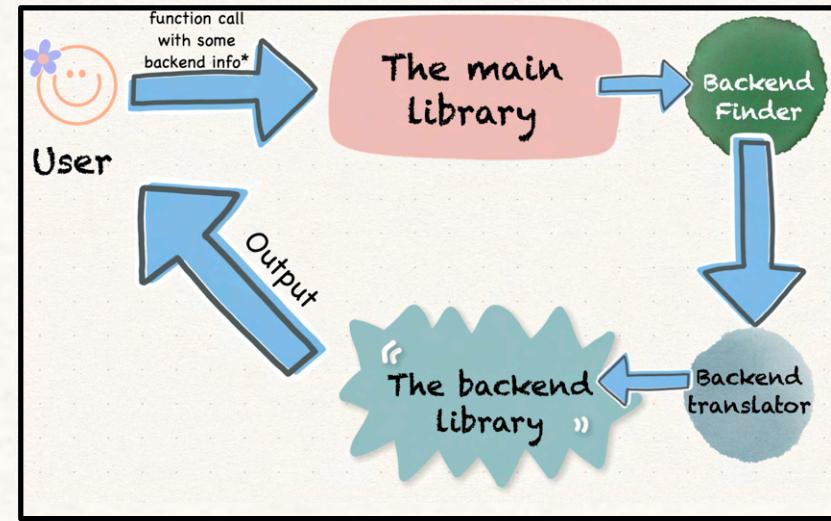
in NetworkX



# Backend Finder

## in NetworkX

*NetworkX discovers backend packages by loading the `networkx.backends` Python entry\_point*

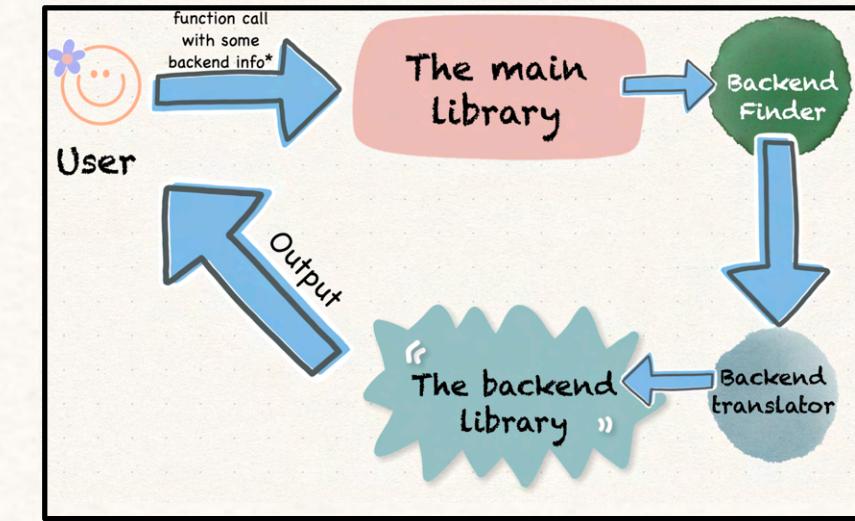


# Backend Finder

## in NetworkX

*NetworkX discovers backend packages by loading the `networkx.backends` Python entry\_point*

*What is `networkx.backends`?*



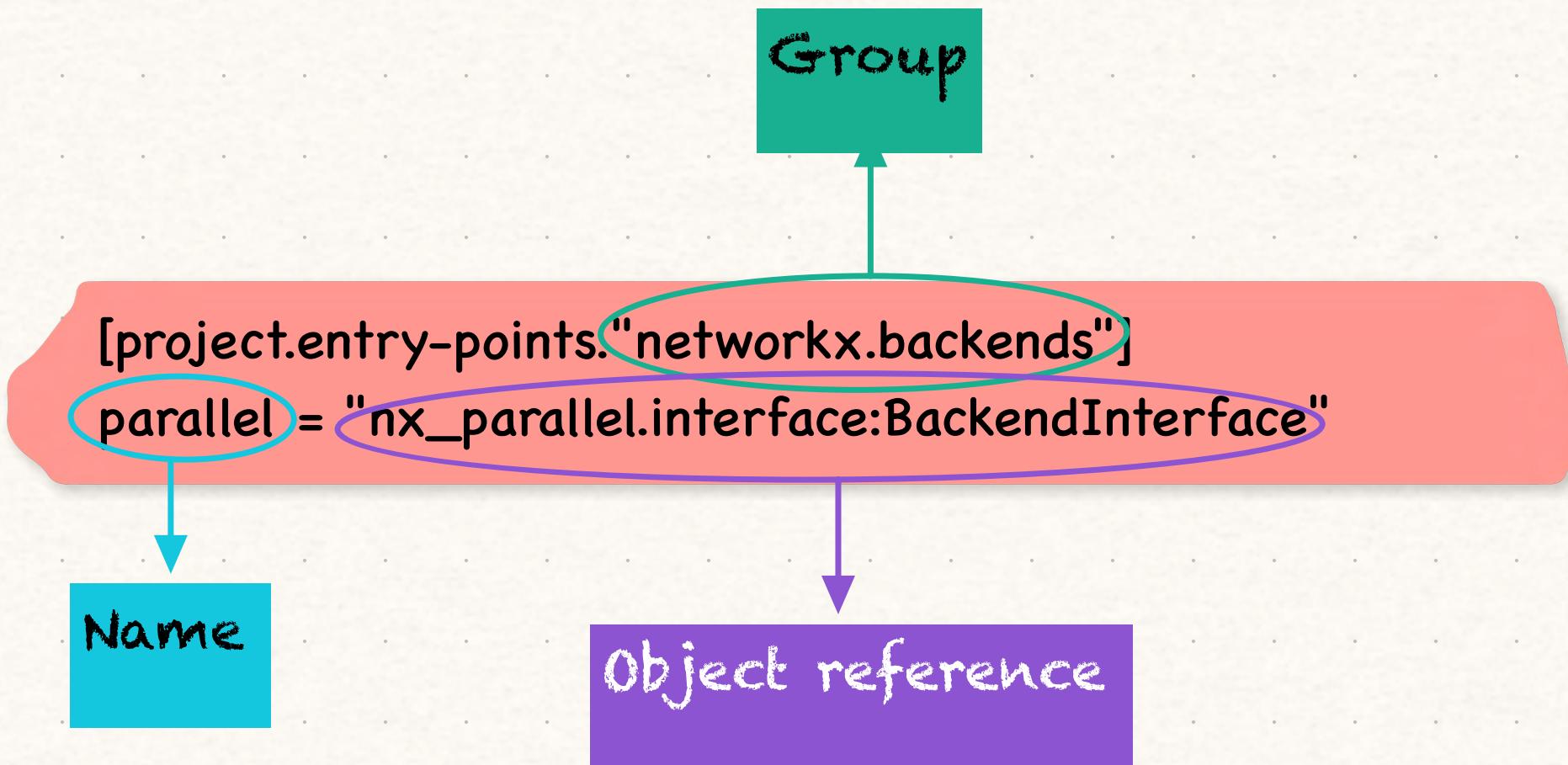
*what is an entry\_point?*

# NetworkX's entry\_points

```
[project.entry-points."networkx.backends"]  
parallel = "nx_parallel.interface:BackendInterface"
```

```
[project.entry-points."networkx.backend_info"]  
parallel = "_nx_parallel:get_info"
```

Three required properties to define a Python entry\_point:

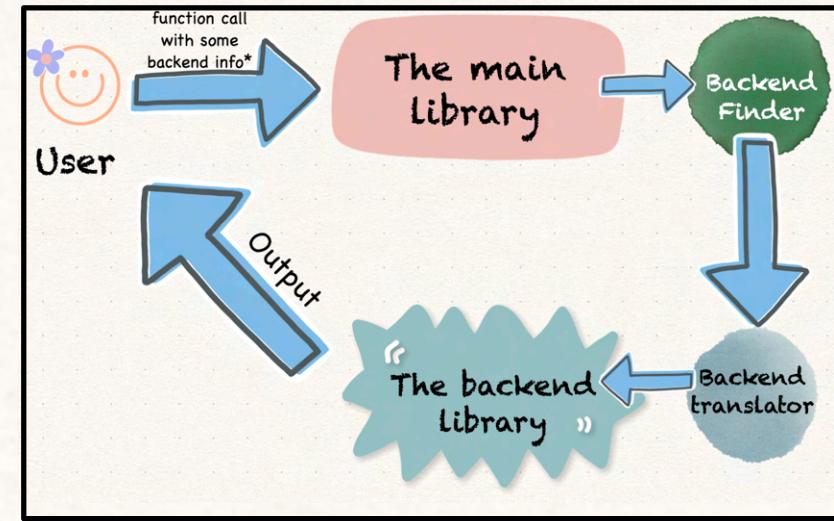


ref. <https://packaging.python.org/en/latest/specifications/entry-points/>

# Backend Finder

## in NetworkX

NetworkX discovers backend packages by loading the `networkx.backends` Python entry\_point



Loading? -

```
>>> from importlib.metadata import entry_points
>>> entry_points(group="networkx.backends")
EntryPoint(name='parallel',
value='nx_parallel.interface:BackendInterface',
group='networkx.backends'),
EntryPoint(name='nx_loopback',
value='networkx.classes.tests.dispatch_interface:backend_interface', group='networkx.backends'))
```

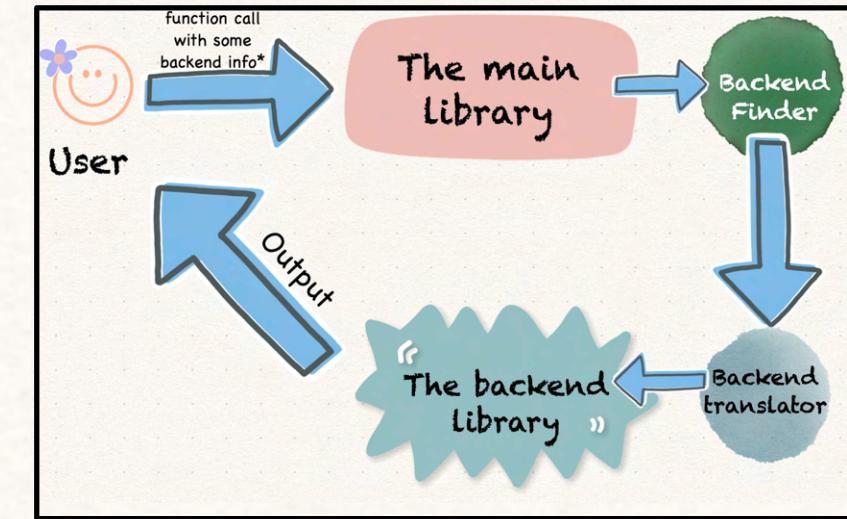
# Backend translator

## in NetworkX

To convert args in the function call into something a backend can understand we use

`'convert_from_nx'` and `'convert_to_nx'`

(which are the attributes of the object referenced in the "networkx.backends" entry\_point by the backend)



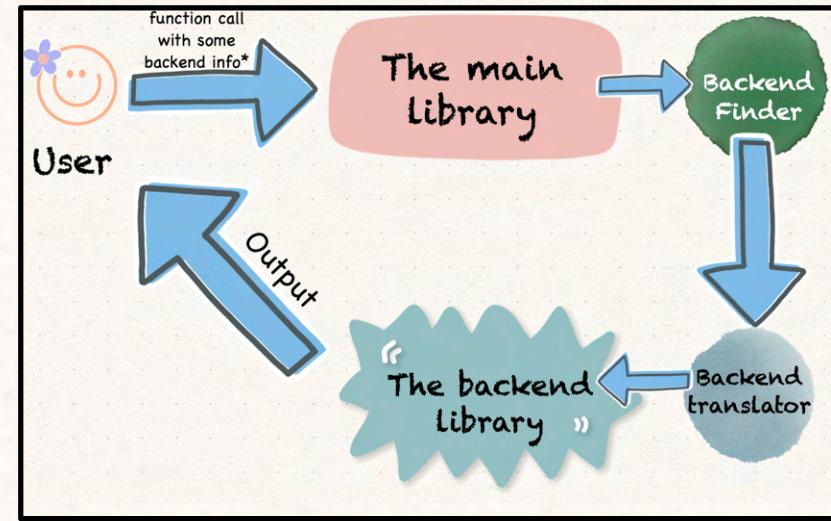
`'convert_from_nx'` and `'convert_to_nx'` are just for args that are Graph objects.

“The backend library”

# in NetworkX

What makes a networkx backend?

- `networkx.backends` entry\_point referring to a
- `BackendInterface` object with attributes -
- `convert\_from\_nx`, `convert\_to\_nx`, and all the algorithms implemented in the backend. And a backend graph object with
- `\_\_networkx\_backend\_\_` attribute.



# So, to summarise...

`nx.betweenness_centrality(G, backend="parallel")`

`nx.betweenness_centrality(cug)`

`with nx.config(backend_priority=[graphblas]):`  
`nx.betweenness_centrality(G)`

## NetworkX

`@_dispatchable`  
`def betweenness_centrality(G, ...):`  
...

`nx-parallel`

```
def betweenness_centrality(G, ...):  
    ...
```

`nx-cugraph`

```
def betweenness_centrality(G, ...):  
    ...
```

`graphblas`

```
def betweenness_centrality(G, ...):  
    ...
```

# Going back to the poster...

**NetworkX Backend Dispatching**

**Plugin Arch.**

**entry\_points**

**Automatic backend testing**

**For testing:**  
NETWORKX\_TEST\_BACKEND=parallel  
NETWORKX\_FALLBACK\_TO\_NX=True  
pytest --pyargs networkx "\$@"

**9th July, 2024**

**NetworkX**  
A pure Python library for network analysis

**Backend Interface**

**config**

**can\_run**  
**should\_run**  
**on\_start\_tests**  
**caching**  
**logging**  
**backend priority**

**nx\_parallel**

A parallel backend for NetworkX, utilizing Joblib to implement graph algorithms on multiple CPU cores.

**Speed-ups**

local\_efficiency-**8.8x**  
betweenness\_centrality-**6.4x**  
square\_clustering-**4.1x**  
bipartite.node\_redundancy-**4.1x**

\*for 300-node random graphs (edge probability = 0.6)

**Usage:**

```
>>> import nx_parallel as nxp
>>> nx.betweenness_centrality(G,
>>> backend="parallel")
>>> H = nxp.ParallelGraph(G)
>>> nx.betweenness_centrality(H)
```

```
$ export NETWORKX_AUTOMATIC_BACKEND=parallel
&& python nx_code.py
```

**TO-DO**

- Designing the pipeline nicely
- Benchmarking
- Adding distributed graph algorithms

**joblib**

**loky**

**threading**

**multiprocessing**

**dask, ray...**

**feel free to contribute and nurture!**

Aditi Taneja (@Schefflera-Arboicola)

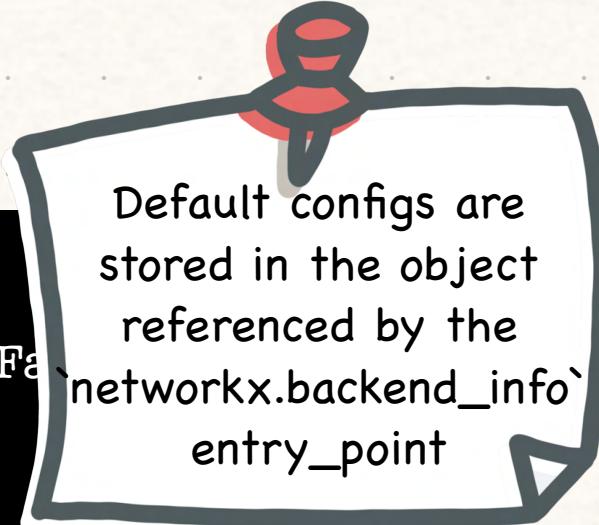
# Understanding NetworkX's API Dispatching with a parallel backend

# Configs in NetworkX

```
>>> import networkx as nx
>>> nx.config
NetworkXConfig(backend_priority=[], backends=Config(parallel=ParallelConfig(active=False,
backend='loky', n_jobs=None, verbose=0, temp_folder=None, max_nbytes='1M',
mmap_mode='r', prefer=None, require=None, inner_max_num_threads=None,
backend_params={})), cacheConvertedGraphs=True)
```

# Configs in NetworkX

```
>>> import networkx as nx  
>>> nx.config  
NetworkXConfig(backend_priority=[], backends=Config(parallel=ParallelConfig(active=False,  
backend='loky', n_jobs=None, verbose=0, temp_folder=None, max_nbytes='1M',  
mmap_mode='r', prefer=None, require=None, inner_max_num_threads=None,  
backend_params={})), cacheConvertedGraphs=True)
```



# A standard nx-parallel algorithm

(a few days ago)

```
def xyz(G,..., get_chunks="chunks"):  
    chunks = chunk(G)  
    results = joblib.Parallel(n_jobs=-1)(  
        joblib.delayed(chunks)  
    )  
    result = combine(results)  
    return result
```

# A standard nx-parallel algorithm

(a few days ago)

```
def xyz(G,..., get_chunks="chunks",
        chunks = chunk(G)
        results = joblib.Parallel(n_jobs=-1)|
            joblib.delayed(chunks)
        )
result = combine(results)
return result
```

allowing users  
to modify  
these  
parameters

## NetworkX Backend Dispatching



Plugin Arch.

entry\_points

Automatic backend testing

```
For testing:  
NETWORKX_TEST_BACKEND=parallel  
NETWORKX_FALLBACK_TO_NX=True  
pytest --pyargs networkx "$@"
```

9th July, 2024

can\_run  
should\_run  
on\_start\_tests  
backend priority  
caching  
logging  
config

## NetworkX

A pure Python library for  
network analysis

“Speed-ups”

local\_efficiency-**8.8x**  
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## nx-parallel

A parallel backend for NetworkX, utilizing Joblib to implement graph  
algorithms on multiple CPU cores.

TO-DO

- Designing the pipeline nicely
- Benchmarking
- Adding distributed graph algorithms

feel free to contribute and nurture!

Aditi Taneja (@Schefflera-Arboicola)

## Usage:

```
>>> import nx_parallel as nxp  
>>> nx.betweenness_centrality(G,  
backend="parallel")  
>>> H = nxp.ParallelGraph(G)  
>>> nx.betweenness_centrality(H)  
  
$ export NETWORKX_AUTOMATIC_BACKEND=parallel  
&& python nx_code.py
```



Loky

threading

multiprocessing

dask, ray...

## joblib

Dashed line --> Solid line

# Configuring nx-parallel

Joblib

```
joblib.parallel_config(n_jobs = 6, verbose=50)  
  
G = nx.complete_graph(5)  
  
# n_jobs = 6, verbose = 50  
nx.square_clustering(G, backend="parallel")  
  
with joblib.parallel_config(n_jobs=8):  
    # n_jobs = 8, verbose = 50  
    nx.square_clustering(G, backend="parallel")
```

NetworkX

```
nx.config.backends.parallel.active = True  
nx.config.backends.parallel.n_jobs = 6  
nx.config.backends.parallel.verbose=50  
  
G = nx.complete_graph(5)  
  
# n_jobs = 6, verbose = 50  
nx.square_clustering(G, backend="parallel")
```

```
with nx.config.backends.parallel(n_jobs=8):  
    # n_jobs = 8, verbose = 50  
    nx.square_clustering(G, backend="parallel")
```

# Syncing the two?

```
nx.config.backends.parallel.n_jobs = 6
joblib.parallel_config(verbose=50)

G = nx.complete_graph(5)

# n_jobs = 6, verbose = 50
nx.square_clustering(G, backend="parallel")

with nx.config.backends.parallel(n_jobs=8):
    # n_jobs = 8, verbose = 50
    nx.square_clustering(G, backend="parallel")
    with joblib.parallel_config(verbose=10):
        # n_jobs = 8, verbose = 10
        nx.square_clustering(G, backend="parallel")
```

ref. [https://github.com/networkx\(nx-parallel/issues/76](https://github.com/networkx(nx-parallel/issues/76) for more.

# Dispatching evolves an API... to be exactly the same!

- "Double down" on the strength of a library
  - **API:** becomes *the* API, avoids fragmentation
  - **Community:** grow a community of backends
- Enable new capabilities
  - Scale performance and data
  - Allow specialization and customization
  - New possibilities, new use cases
- All without fundamentally changing a library
  - Keep what works and what makes it good!



nx-parallel

nx-pandas

nx-cugraph

nx-arangodb

graphblas  
-algorithms

<private>

nx-???  
(what next?)

## Coming soon:

spatch to generalize and document dispatching  
<https://github.com/scientific-python/spatch>

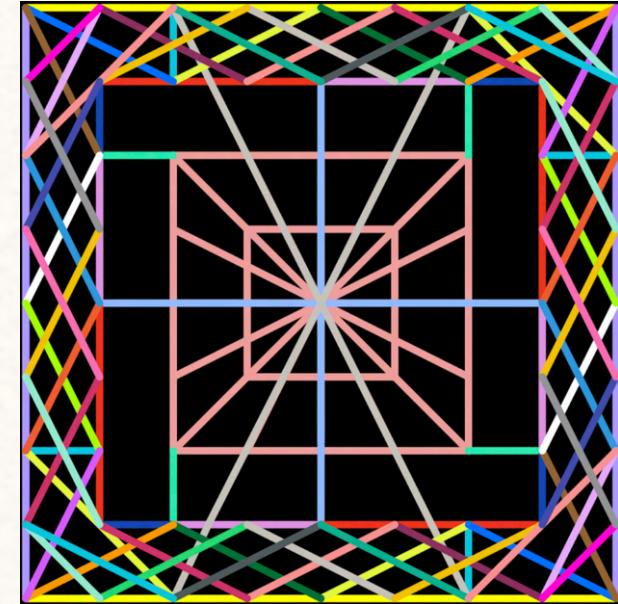
## Join us:

NetworkX dispatching call  
<https://scientific-python.org/calendars>



Erik  
Welch

“  
Thank you :)  
”



Aditi Juneja  
(Mostly  
Schefflera  
Arboricola)