



# Python coding with graphs

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*Distributed Autonomous Systems M*  
*A.A. 2024-2025*

# Graph theory recap

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A **digraph** (directed graph) is a pair  $G = (I, E)$  where

- $I = \{1, \dots, N\}$  is a collection of elements, called **nodes**
- $E \subset I \times I$  is a set of ordered node pairs, called **edges**

An edge from node  $i$  to  $j$  is a pair denoted as  $(i, j)$

In Python we can use the package **networkx** <https://networkx.org/>

**import networkx as nx**



# Networkx: graph creation

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An *undirected* graph in Python can be instantiated via

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

A node, e.g., with index 0, can be added to the graph **G** using

```
G.add_node(0)
```

```
G.add_nodes_from([1, 2, 3])
```

An edge, e.g., between node 1 and 2, can be added to the graph via

```
G.add_edge(1, 2)
```

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

We can draw a graph with

```
nx.draw(G) or nx.draw_circular(G, with_labels=True)
```

# Networkx: create a directed graph

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A *directed* graph can be created via

```
directedG = nx.DiGraph()
```

We can obtain the list of neighbors of a node, e.g., of node **0**, using

```
N_i = nx.neighbors(directedG, 0)
```

(Its explicit casting as a **list** is useful to use it)

Extract the adjacency matrix as a *numpy array* using

```
Adj = nx.adjacency_matrix(directedG).toarray()
```

# Networkx: compute the Laplacian of a graph

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We can compute the Laplacian matrix via

**`L = nx.laplacian_matrix(G).toarray()`**

Alternatively, we can resort to its definition and manually compute it

First, let **`I_N`** be the identity matrix of order **`N`**, then compute the degree matrix via

**`degree = np.sum(Adj + I_N, axis = 0)`**

**`D_IN = np.diag(degree)`**

The in-Laplancian matrix is finally given by

**`L_IN = D_IN - Adj.T`**



# Networkx: predefined graphs

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We can create a *path graph* via

```
pathG = nx.path_graph(N)
```

We can create a *cycle graph* via

```
cycleG = nx.cycle_graph(N)
```

We can create a *star graph* via

```
starG = nx.star_graph(N)
```

We can also create more general random graph as, e.g., an Erdős–Rényi graph as

```
graph_ER = nx.binomial_graph(n=N, p=p_ER)
```

A non-negative square matrix  $A$  is

- row stochastic if its rows sum up to 1
- column stochastic if its columns sum up to 1
- doubly stochastic if it is both row and column stochastic

What happens to the linear averaging algorithm when

- the matrix is column stochastic but not row stochastic?
- the matrix is row stochastic but not column stochastic?