

INFO 1368

Introduction to Network Science

Filippo Menczer



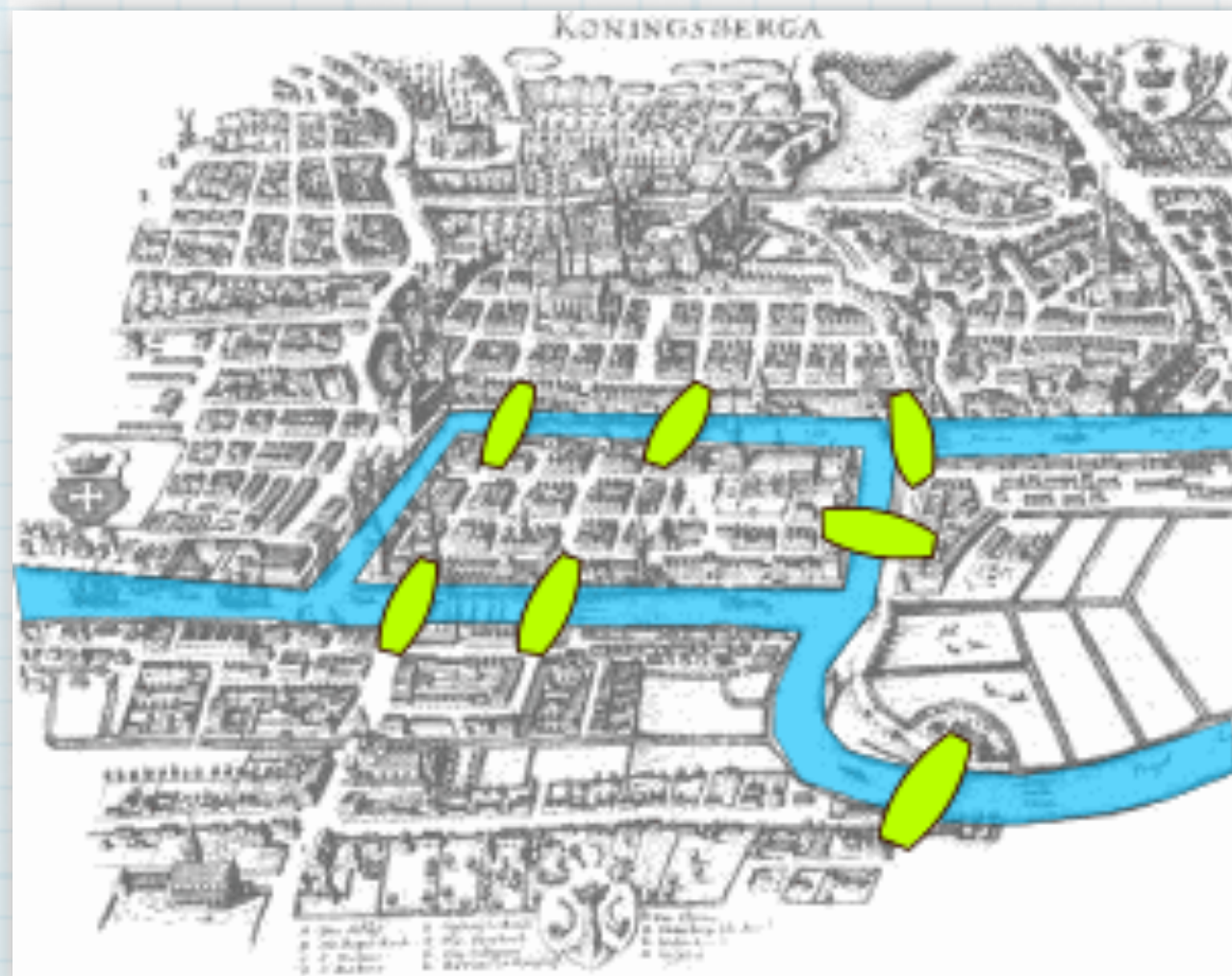
Center for Complex Networks
and Systems Research



SCHOOL OF Informatics and Computing

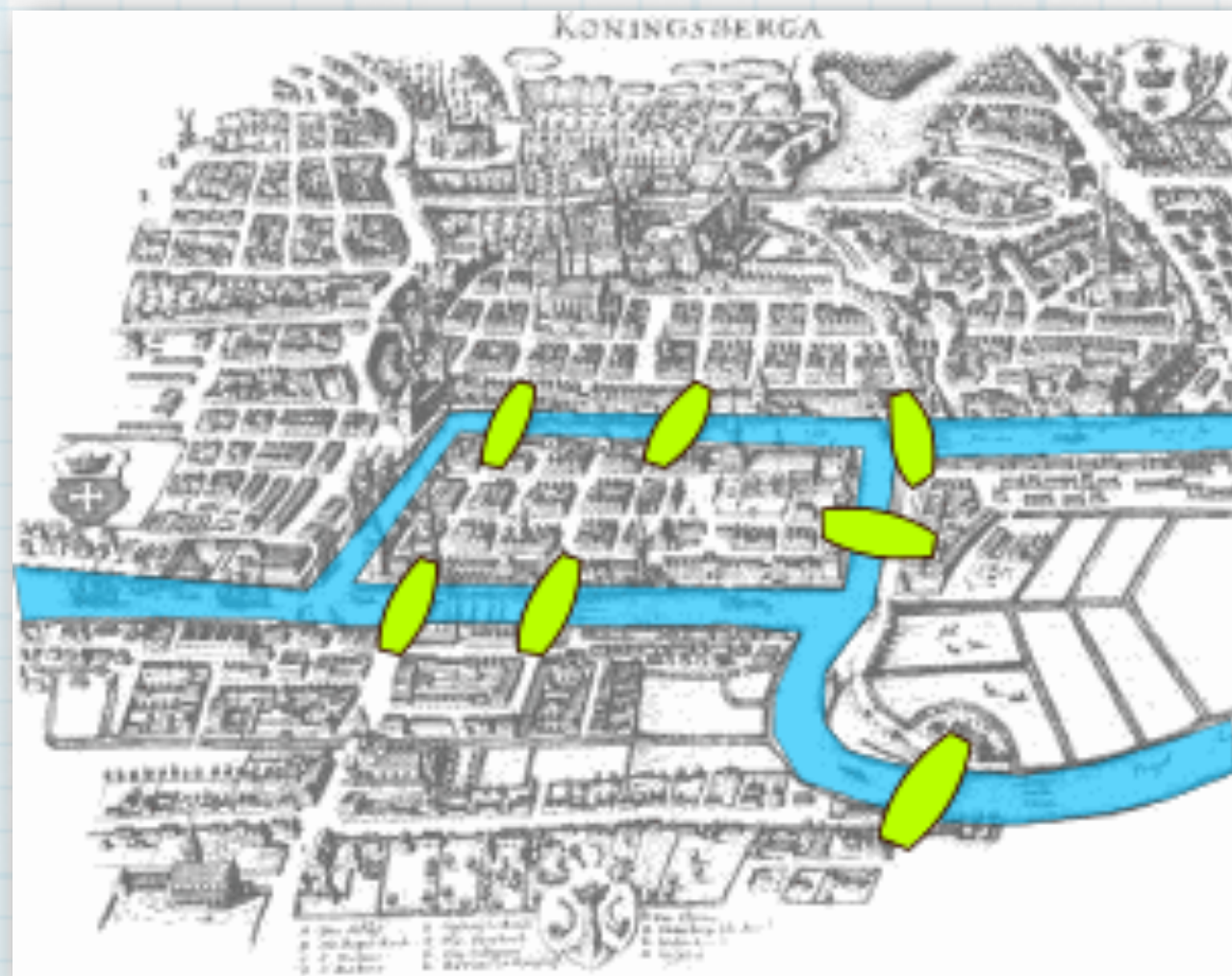
- Attendance
- Schedule update
- Review

Euler circa 1736: Koningsberg bridges

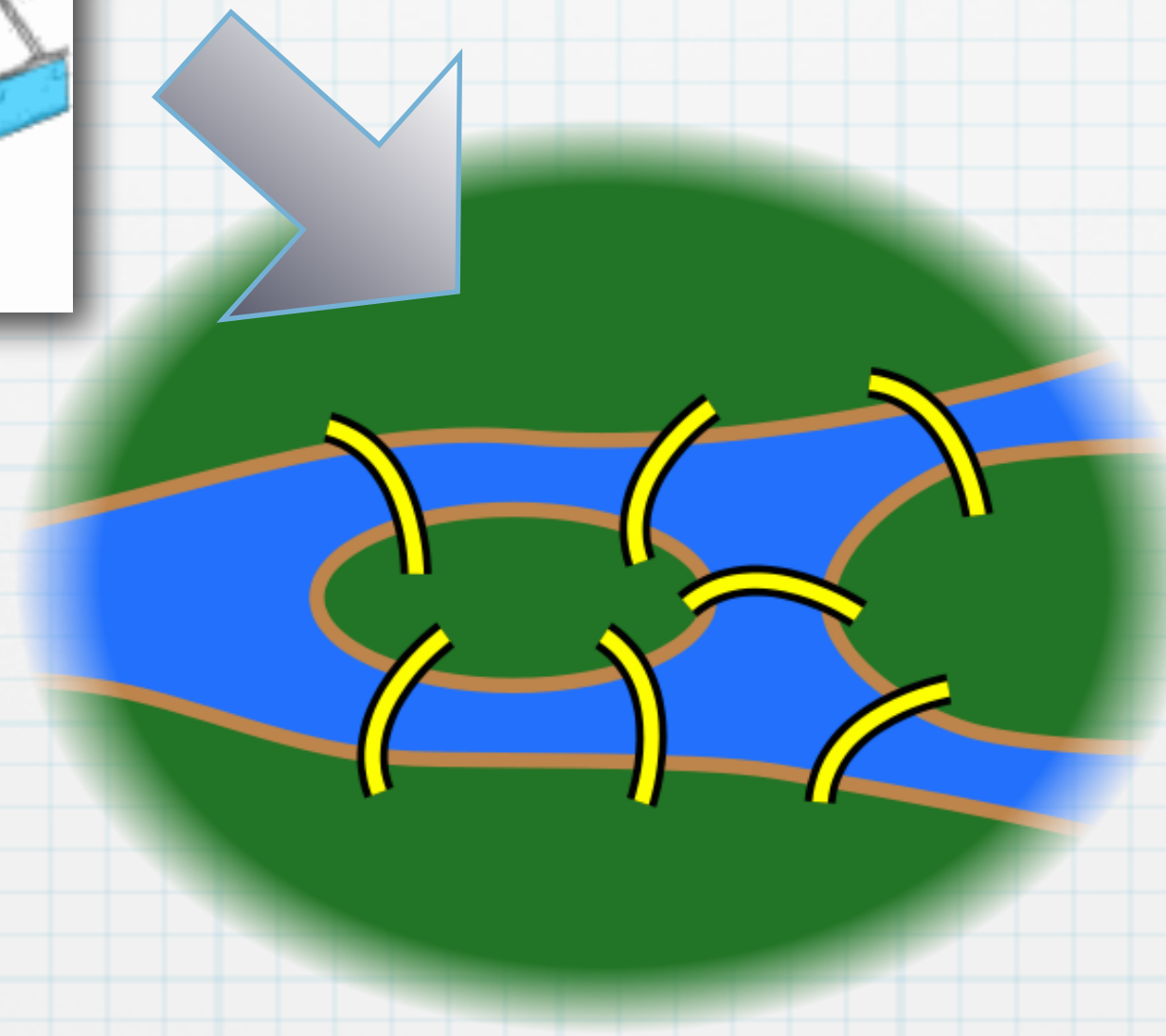


Can you cross
all 7 bridges
just once each?

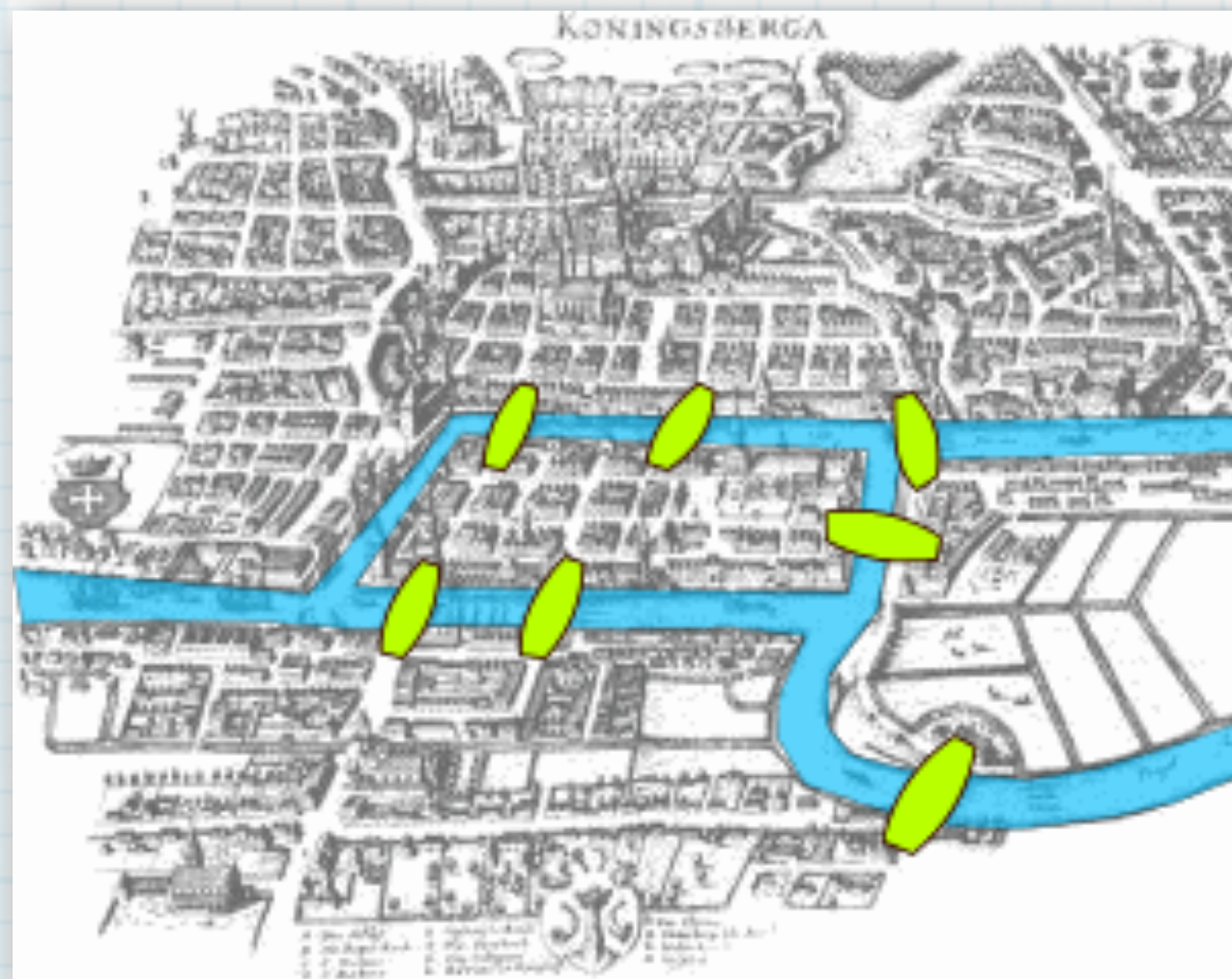
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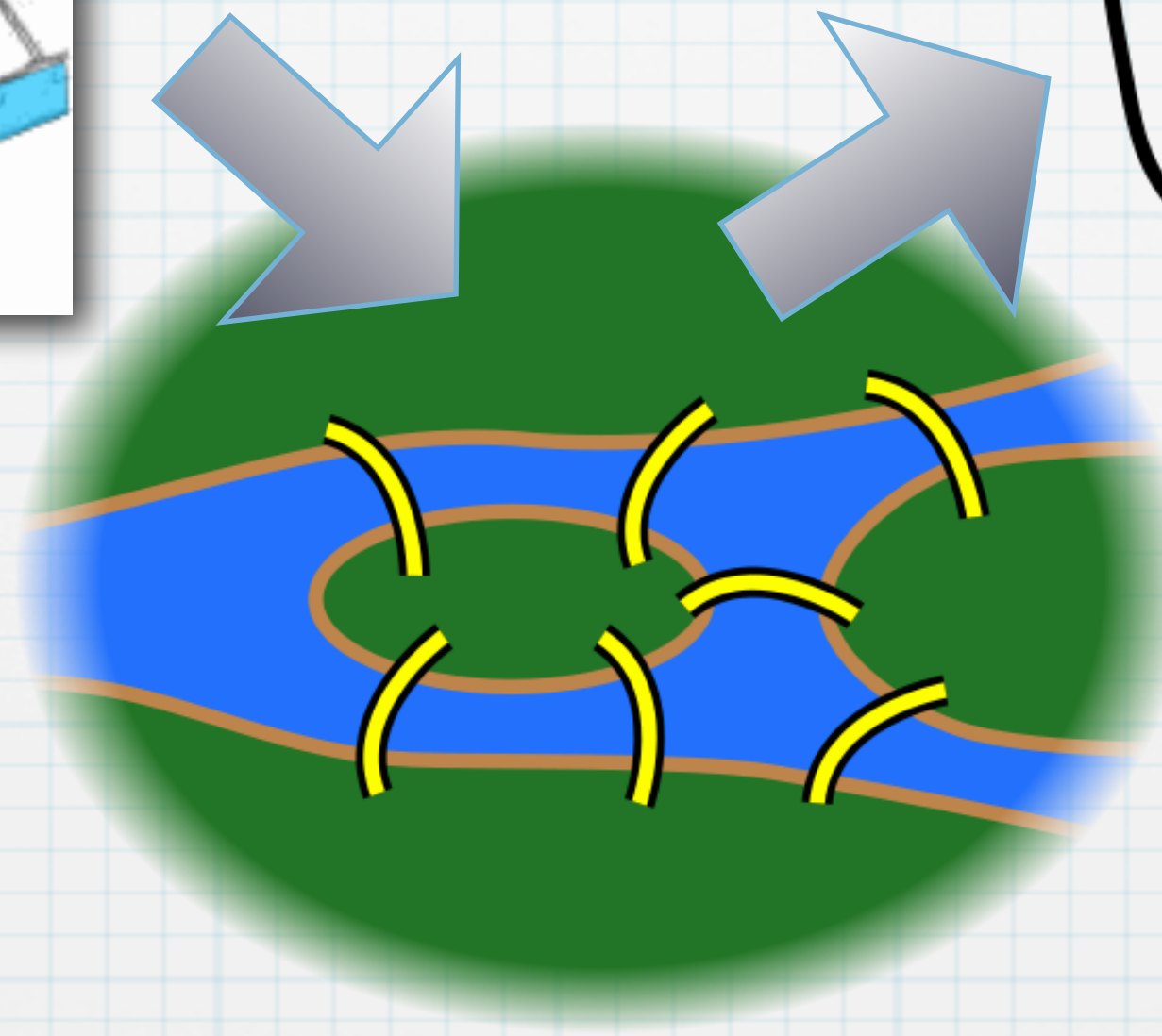
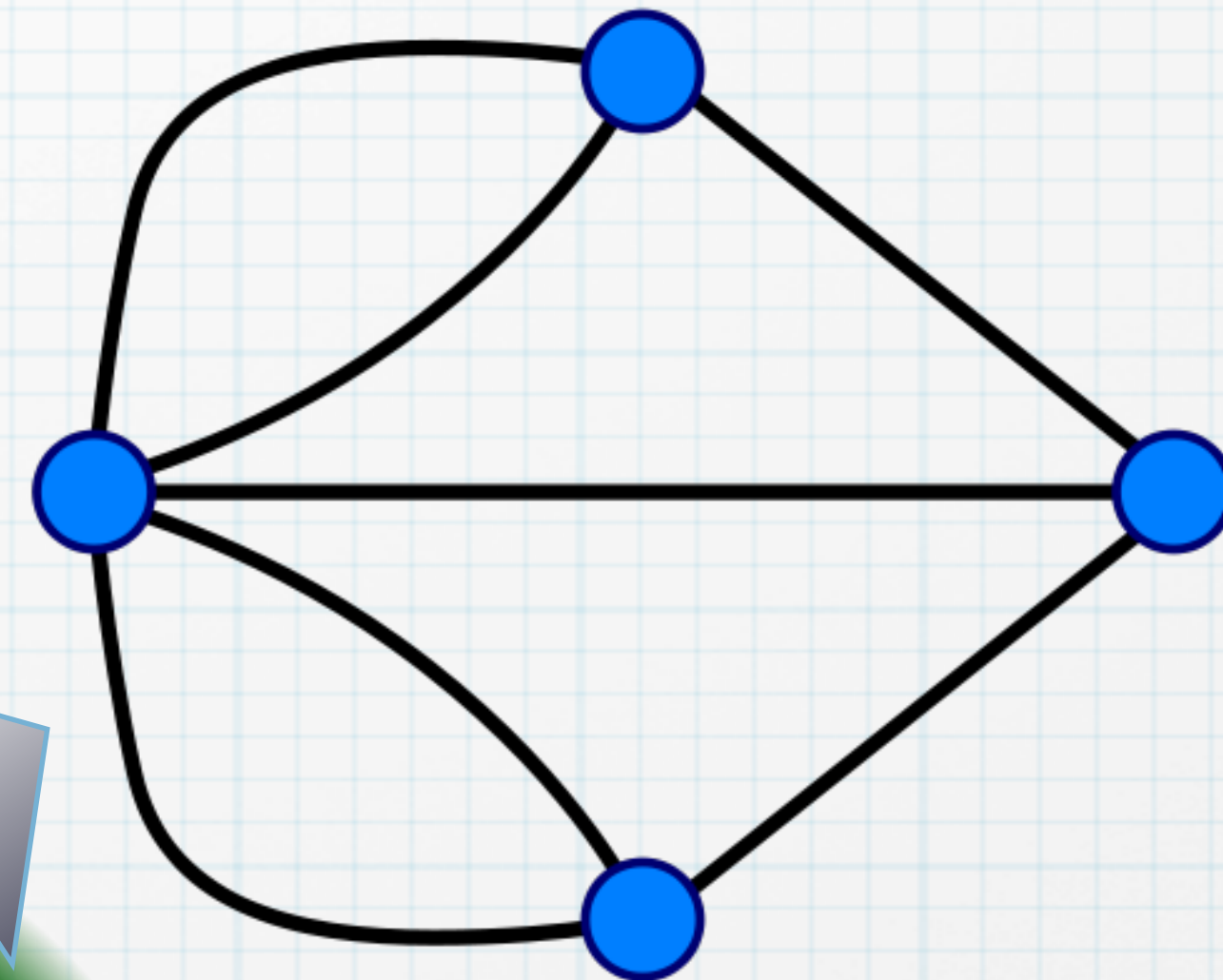
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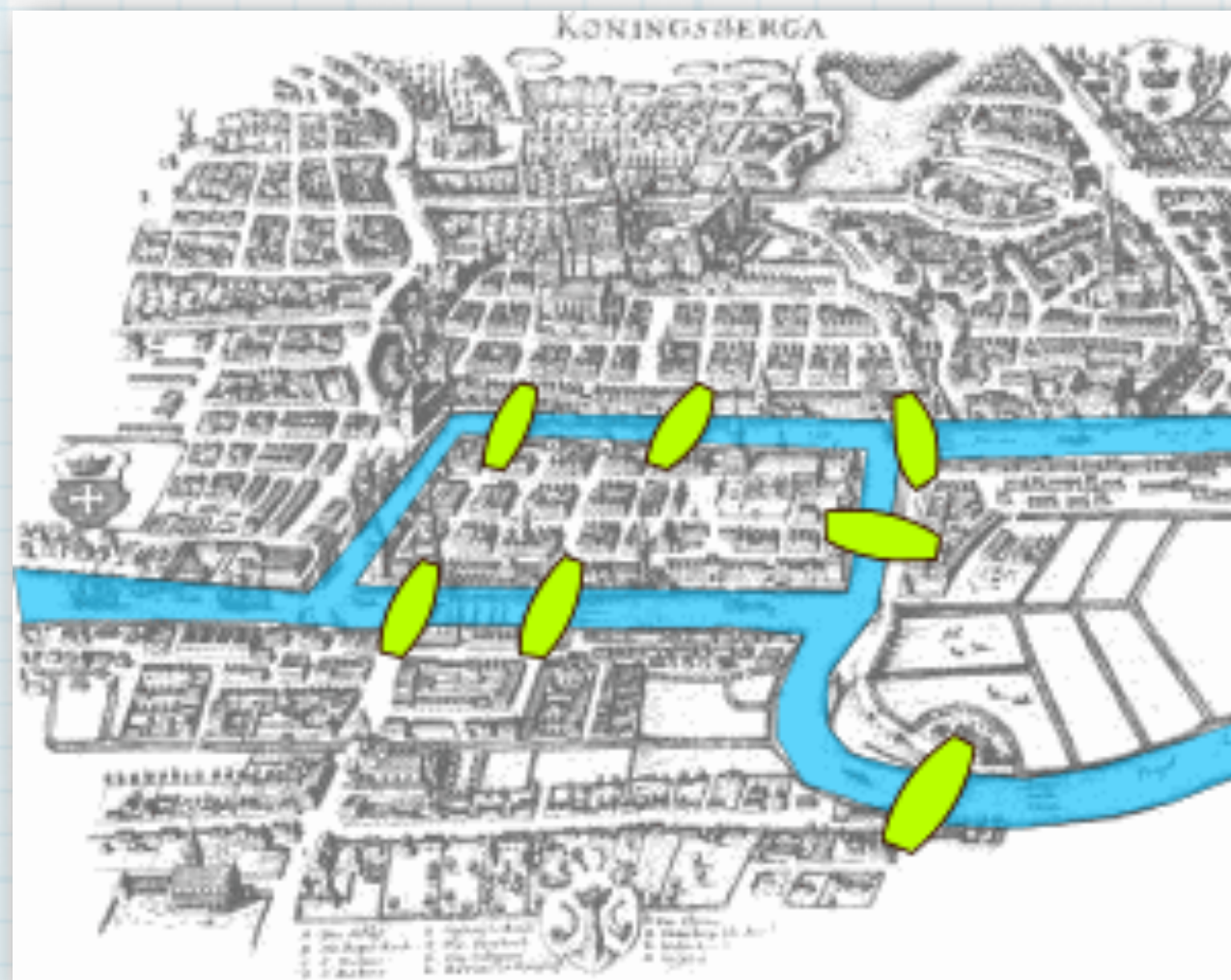
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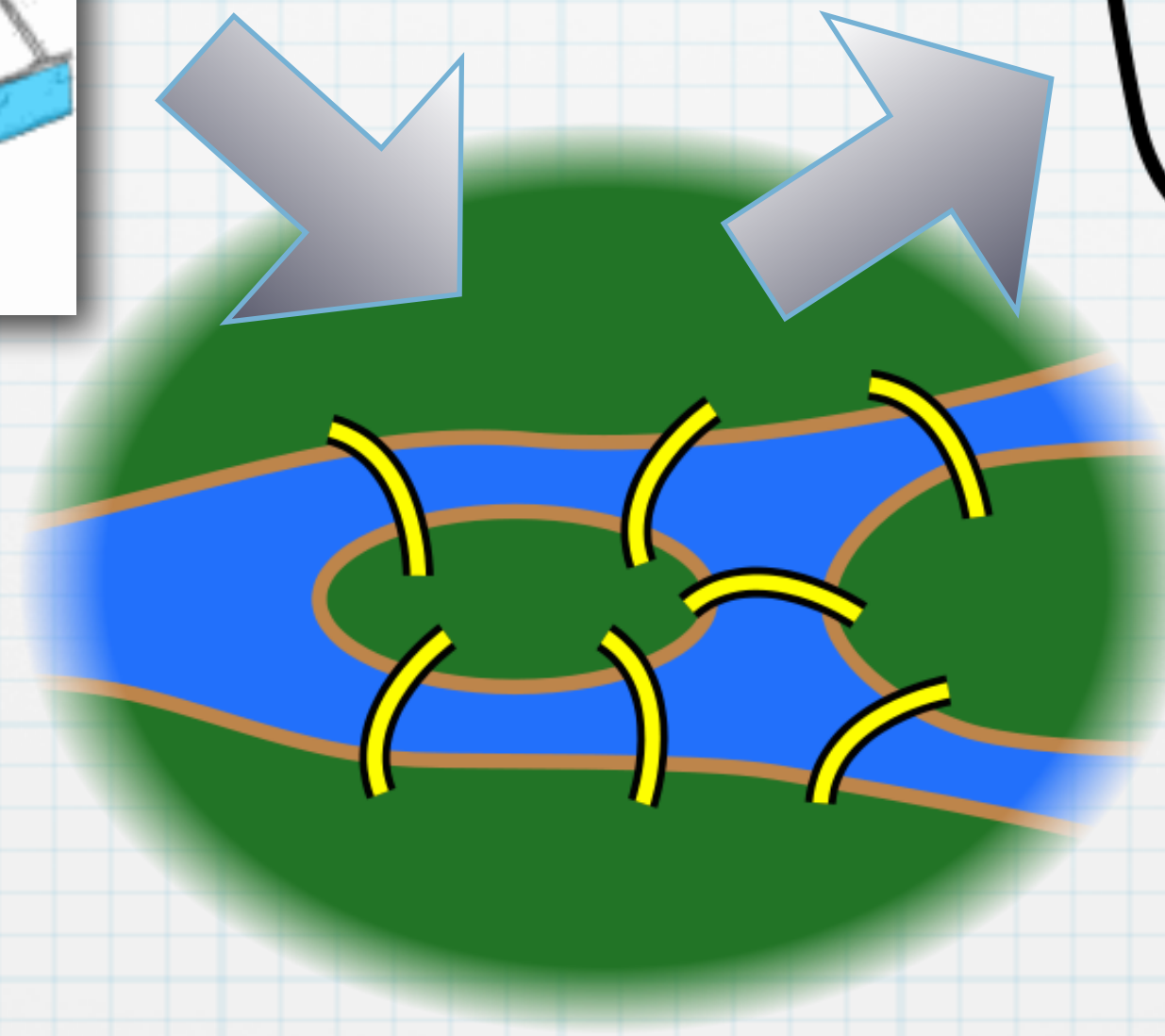
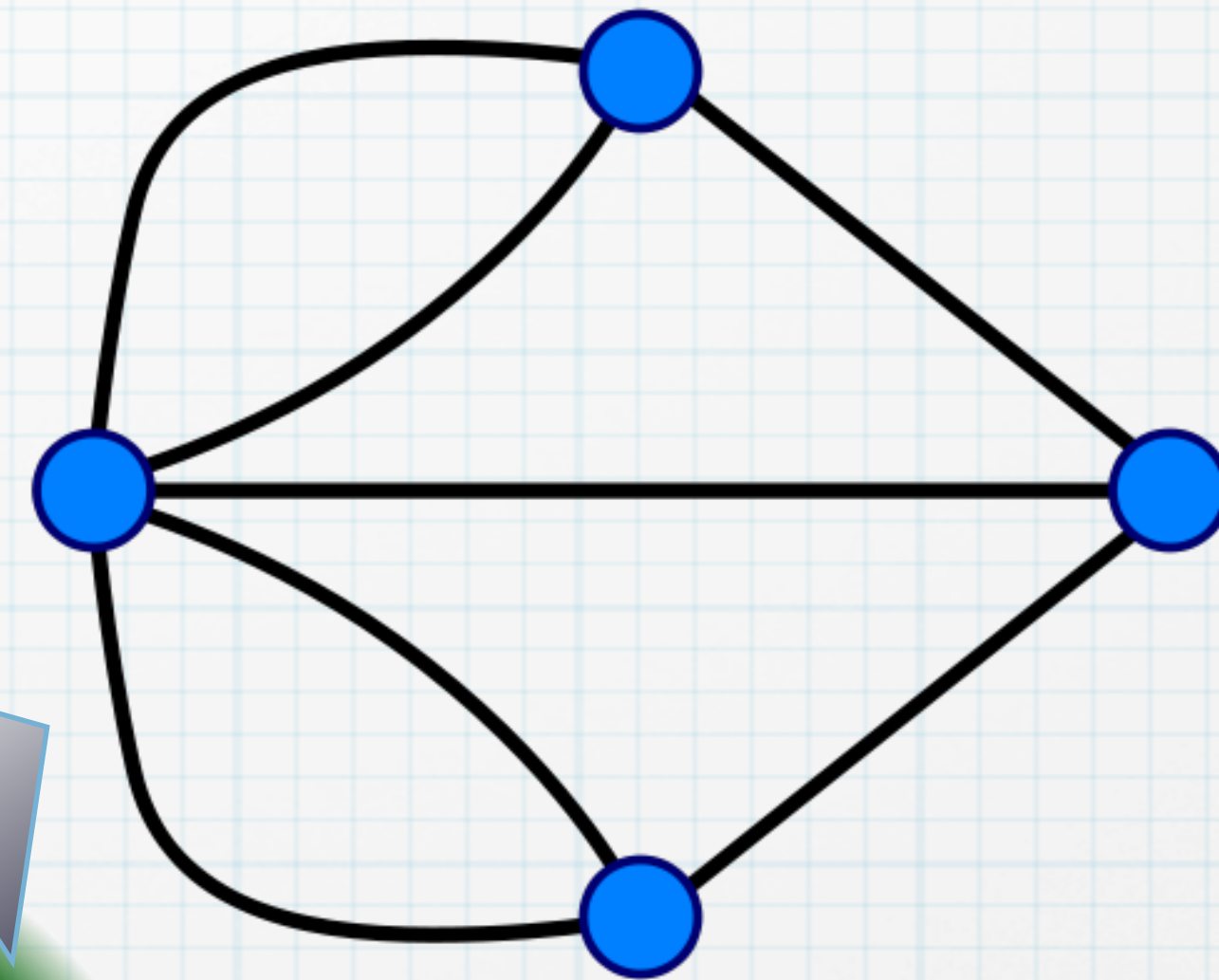
Can you cross
all 7 bridges
just once each?



Euler circa 1736: Koningsberg bridges



Can you cross
all 7 bridges
just once each?



Answer: no. At most
two nodes (start, end)
may have odd degree

What's new?

- * 1930-40's Moreno (sociogram, sociometry) – psychology and social sciences
- * 1960-70's Erdos (random graph theory) – math
- * 1970-80's Granovetter (weak ties) – economics and sociology
- * 1980-90's Waxman (topology generator) – computer science
- * 2000's Non-equilibrium (statistical mechanics) – physics

The age of networks

- * Data gathering and integration
- * Biology
- * Internet, WWW, P2P, ad-hoc.....
- * Social sciences
- * Social media
- * Human mobility/activity

New challenges & new understanding

- * Data size shifts ($10^2 \rightarrow 10^8$ elements) **(complexity)**
- * Different domains (biology, info-structures, infrastructures, social, scientometrics) **(universality)**
- * Large scale longitudinal studies (time series) **(dynamical modeling)**

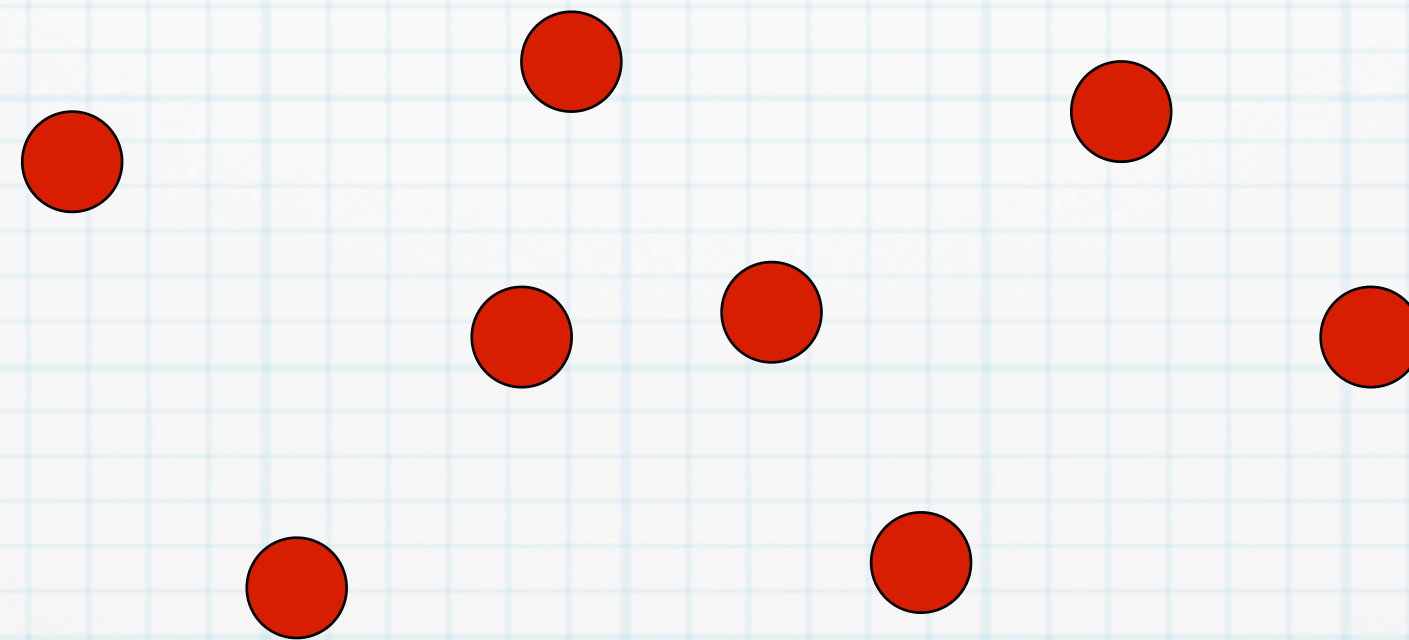
The plan

- * Last week we saw a bunch of interesting networks arising in many different domains
- * Now let us learn the language of networks
 - * The components: nodes, links
 - * Types of networks and representations
 - * Features of nodes and edges

We want to be able to talk about...

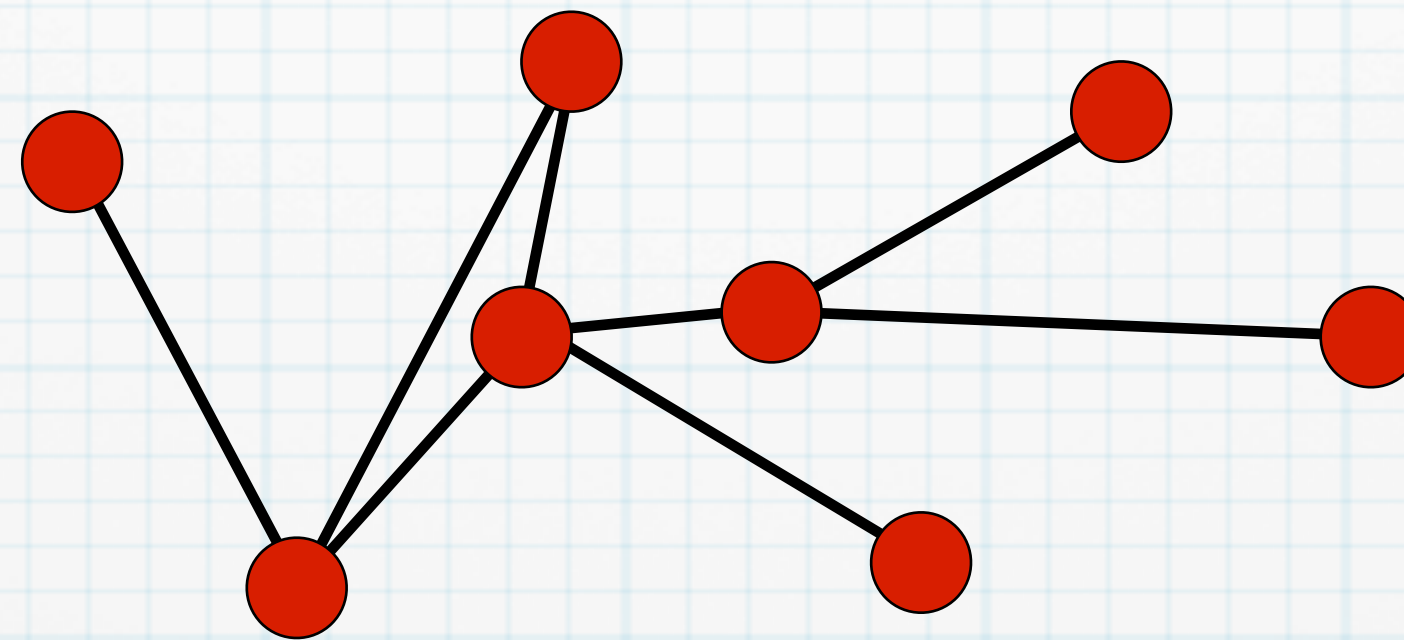
- * properties to characterize structure & behavior of networks
- * roles of networks in affecting processes occurring on network structures

Basic components



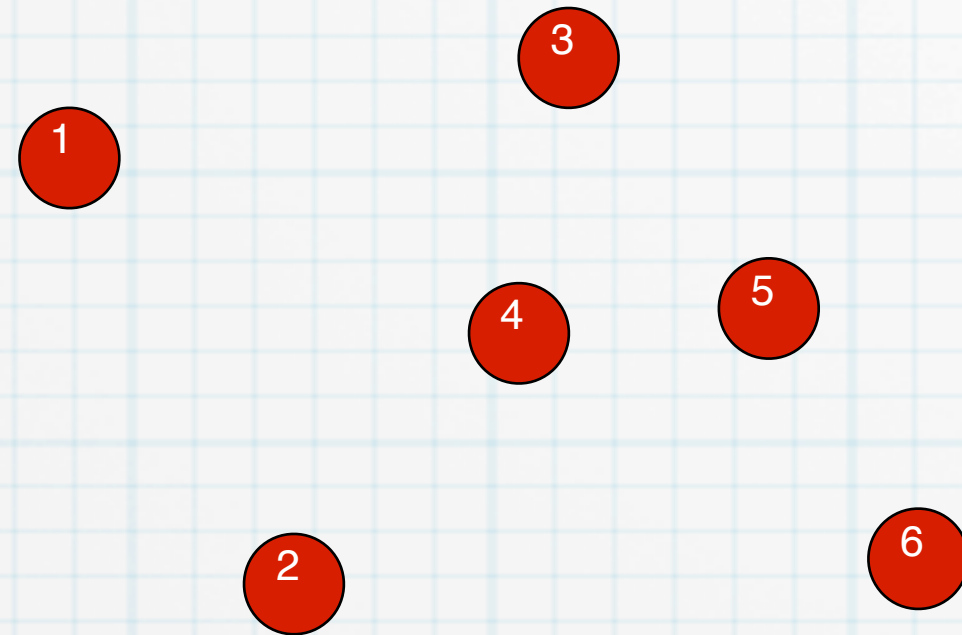
- **elements:** nodes, vertices, ...

Basic components



- **elements:** nodes, vertices, ...
- **interactions:** links, edges, neighbors...

Python and NetworkX

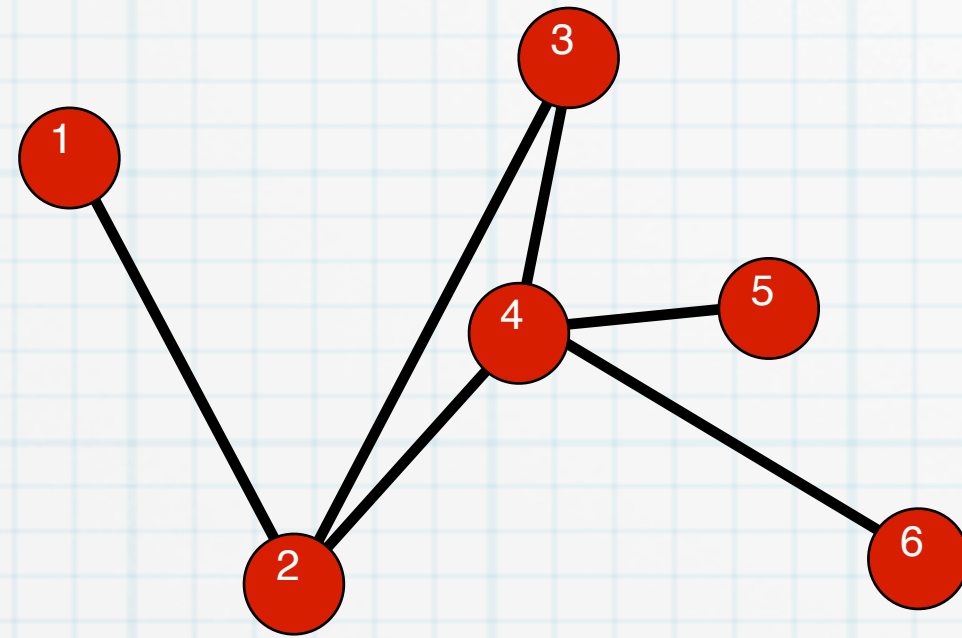


```
import networkx as nx # always!

G = nx.Graph()
G.add_node(1)
G.add_nodes_from([2,3,...])
...
G.add_edge(1,2)
G.add_edges_from([(2,3),(2,4),...])
...
G.nodes()
G.edges()
G.neighbors(4)

for n in G.nodes:
    print(n, G.neighbors(n))
for u,v in G.edges:
    print(u, v)
```


Python and NetworkX



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Some definitions

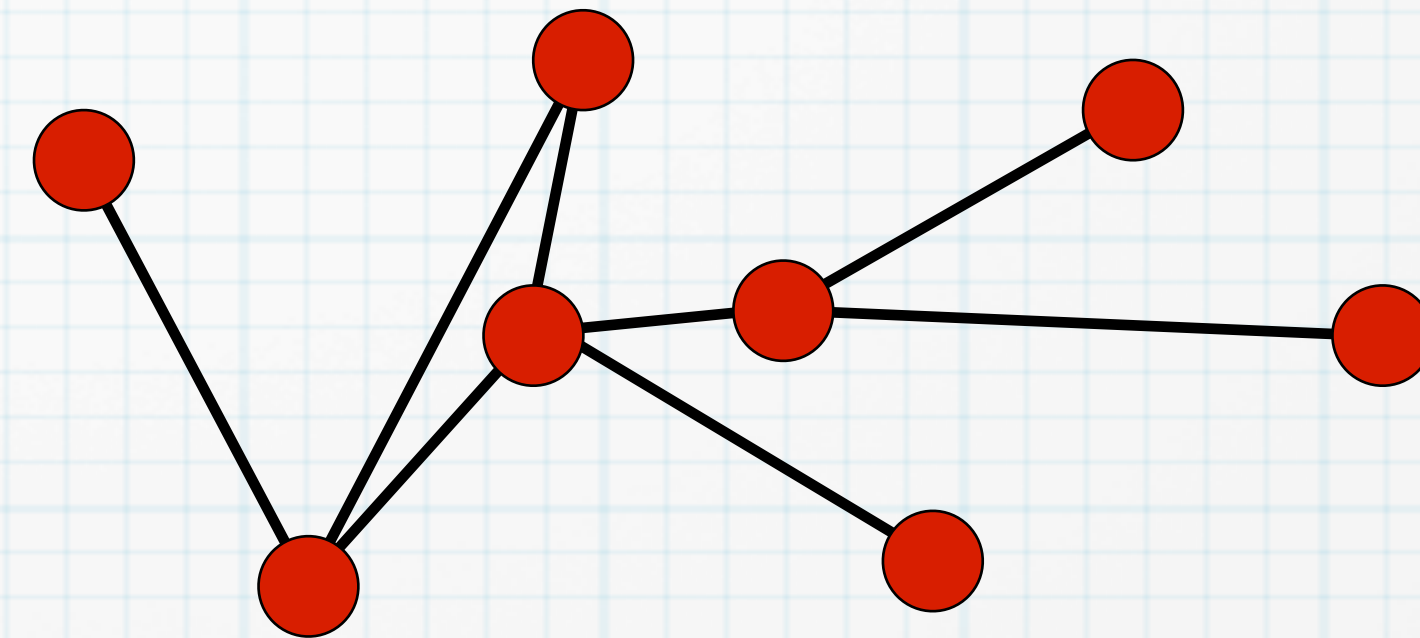
- * Vertices + edges

- * size: # Vertices

- * ~~self loops~~

- * ~~multiple edges~~
~~(multigraphs)~~

- * directed networks: directed edges or links



Some definitions

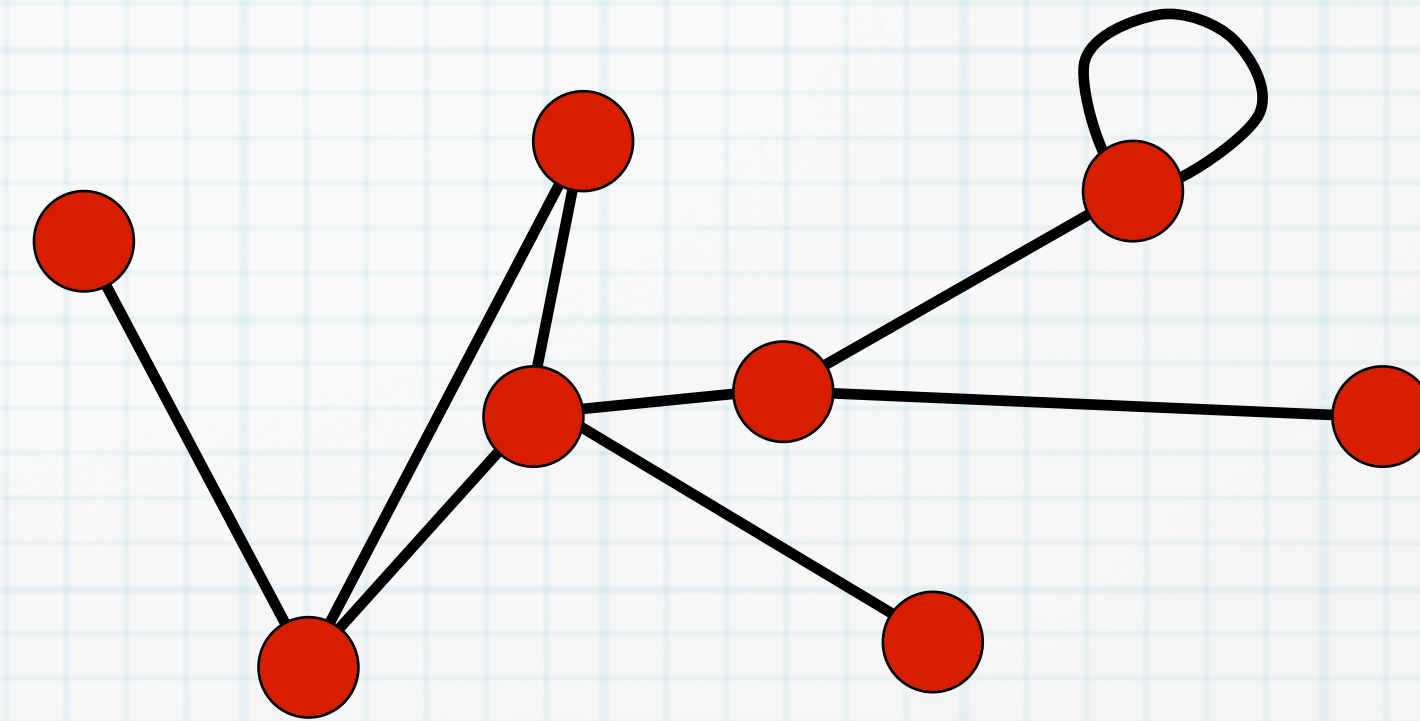
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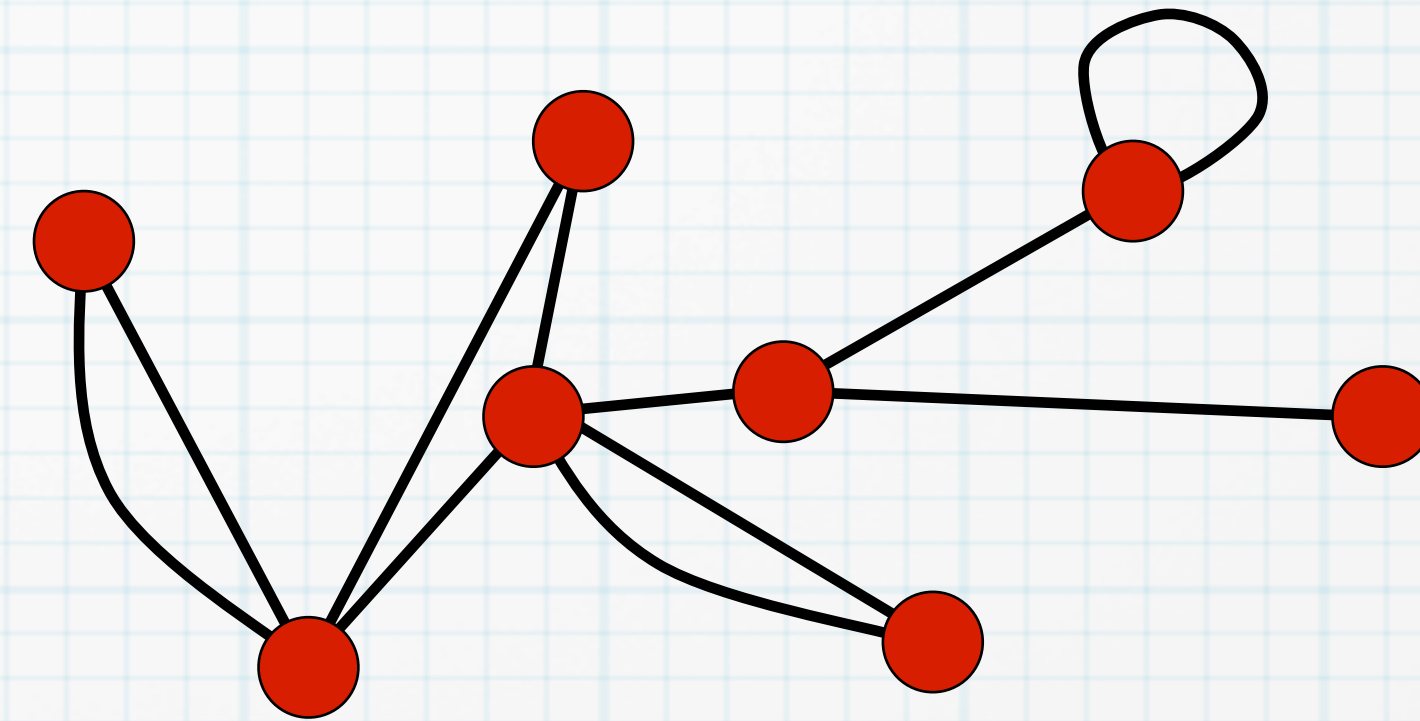
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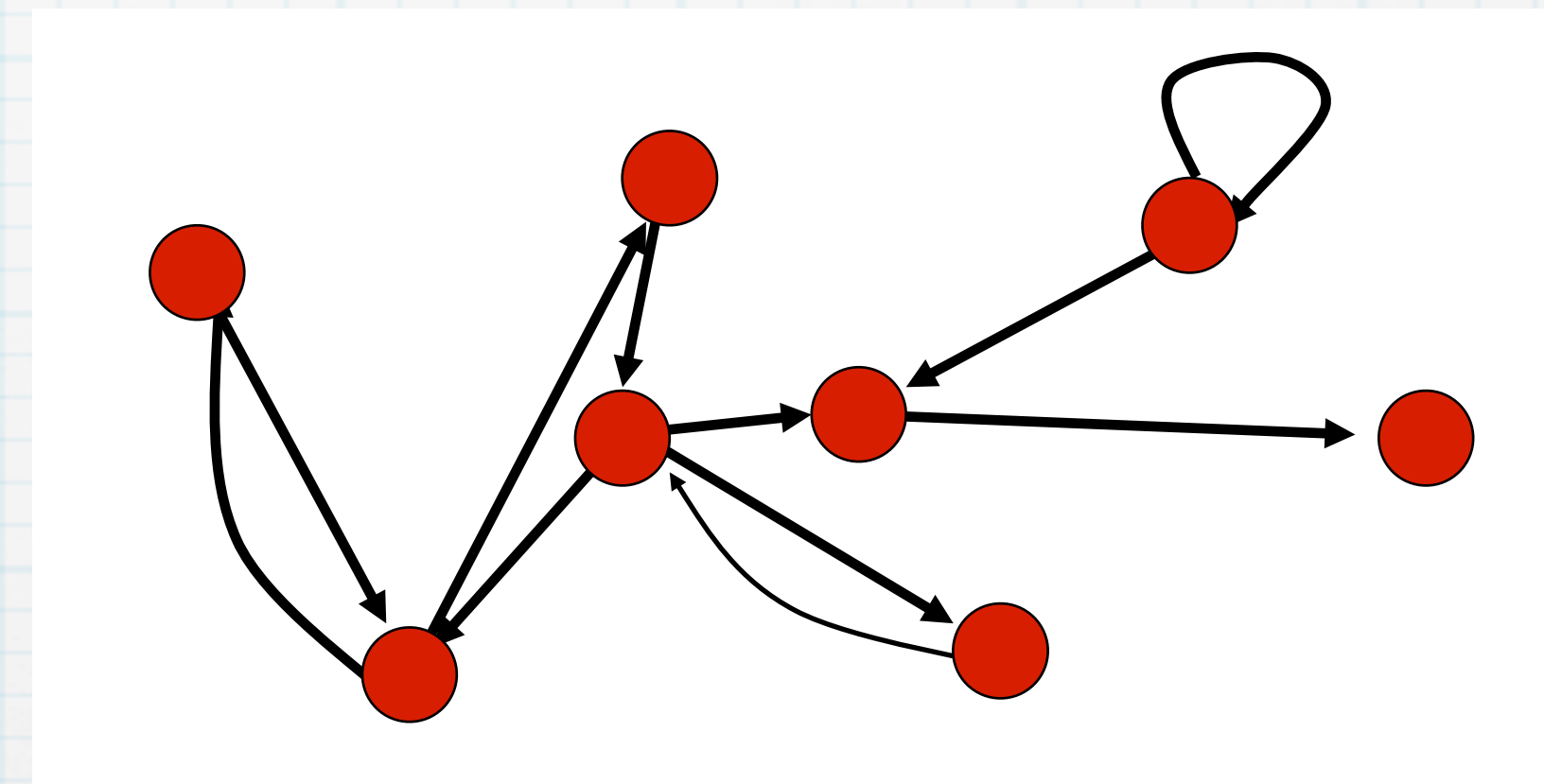
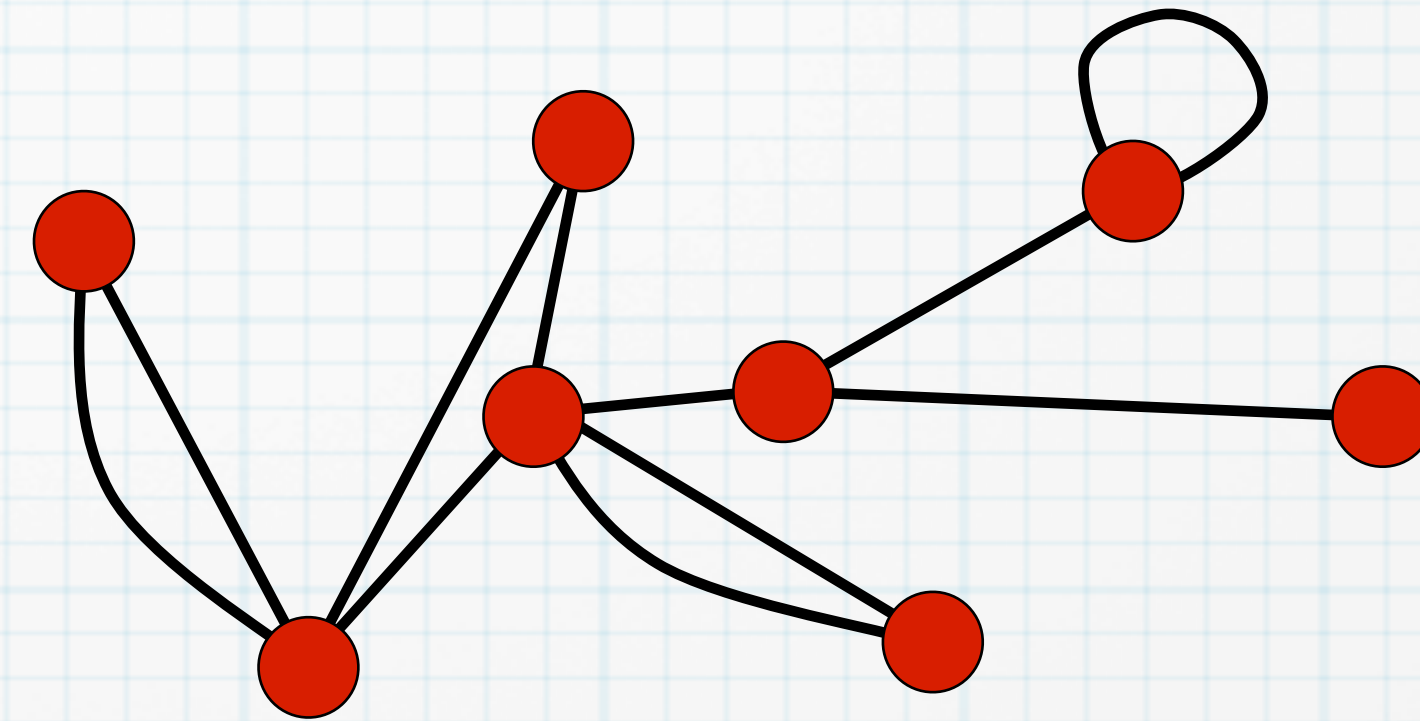
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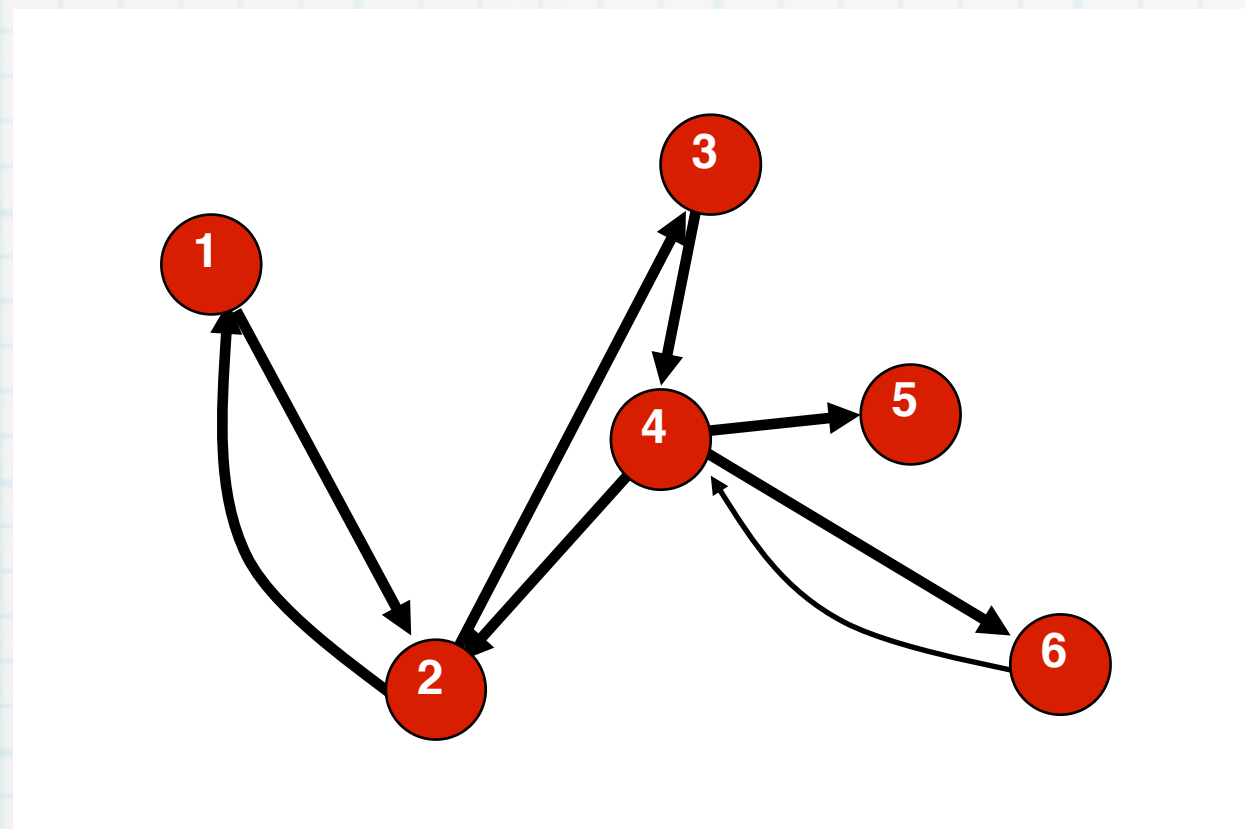
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Python and NetworkX

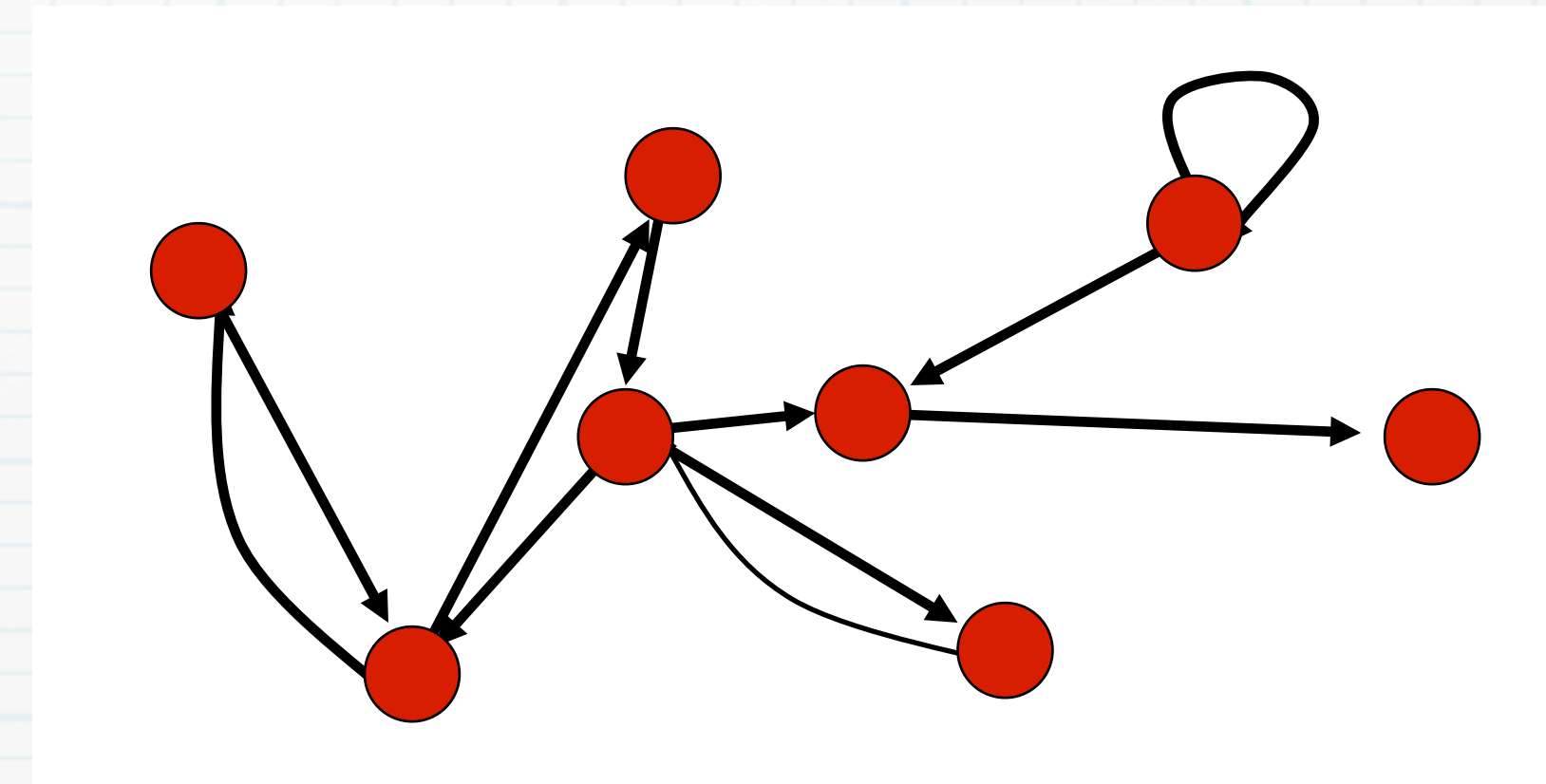


```
import networkx as nx # don't forget!
```

```
D = nx.DiGraph()  
D.add_edge(1,2)  
D.add_edge(2,1)  
D.add_edges_from([(2,3),(3,4),...])  
...  
D.number_of_nodes()  
D.number_of_edges()  
D.edges()  
D.successors(2)  
D.predecessors(2)  
D.neighbors(2)
```

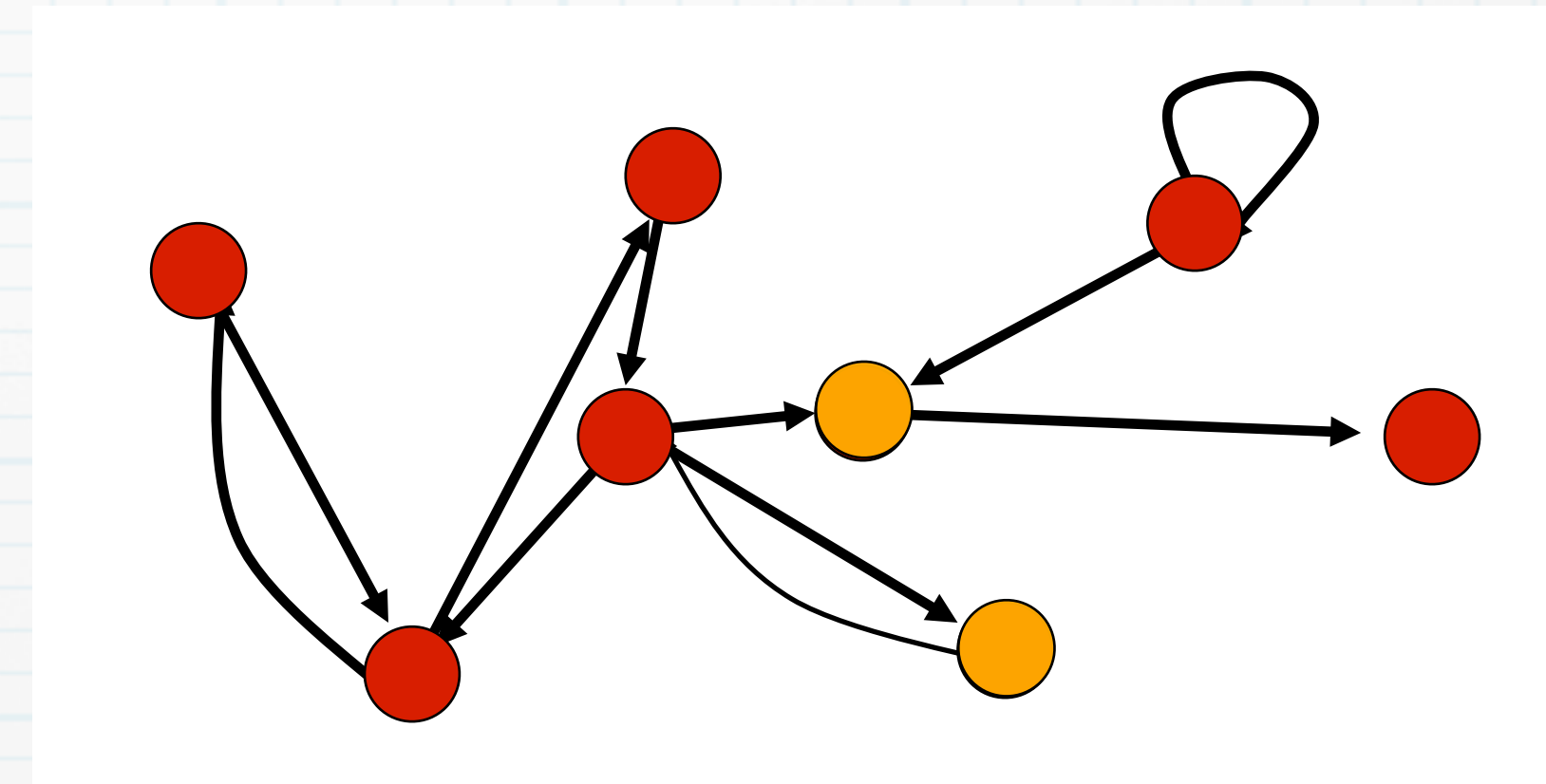

More definitions

- * different types of vertices
- * different types of edges
- * weights
- * bipartite network:
 - * 2 types of nodes
 - * links between different types



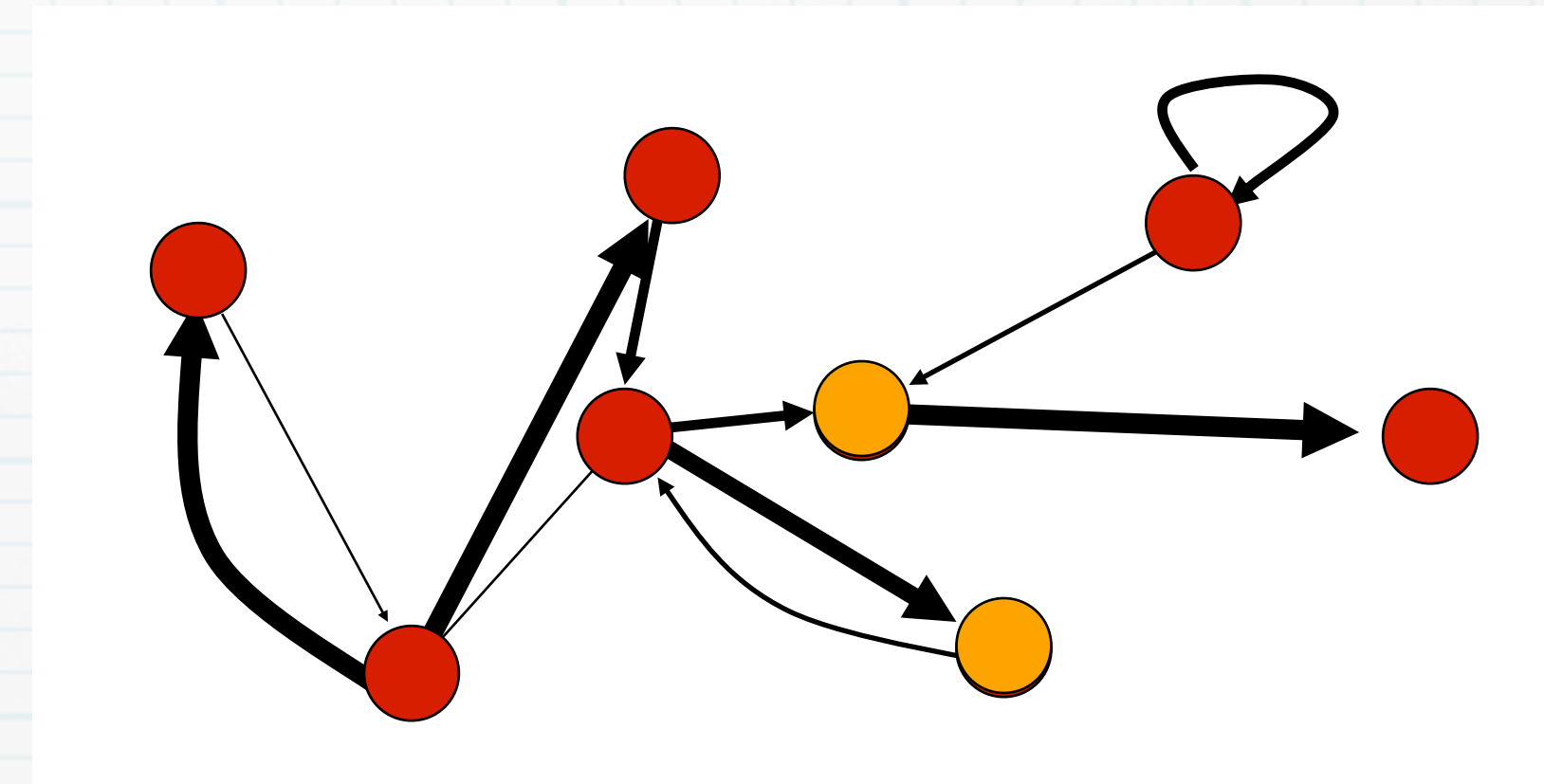
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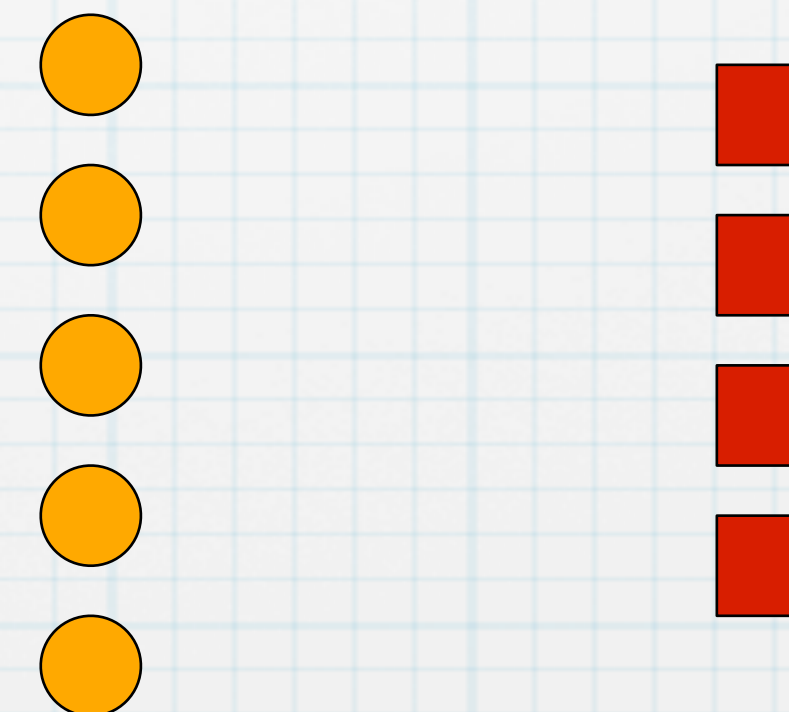
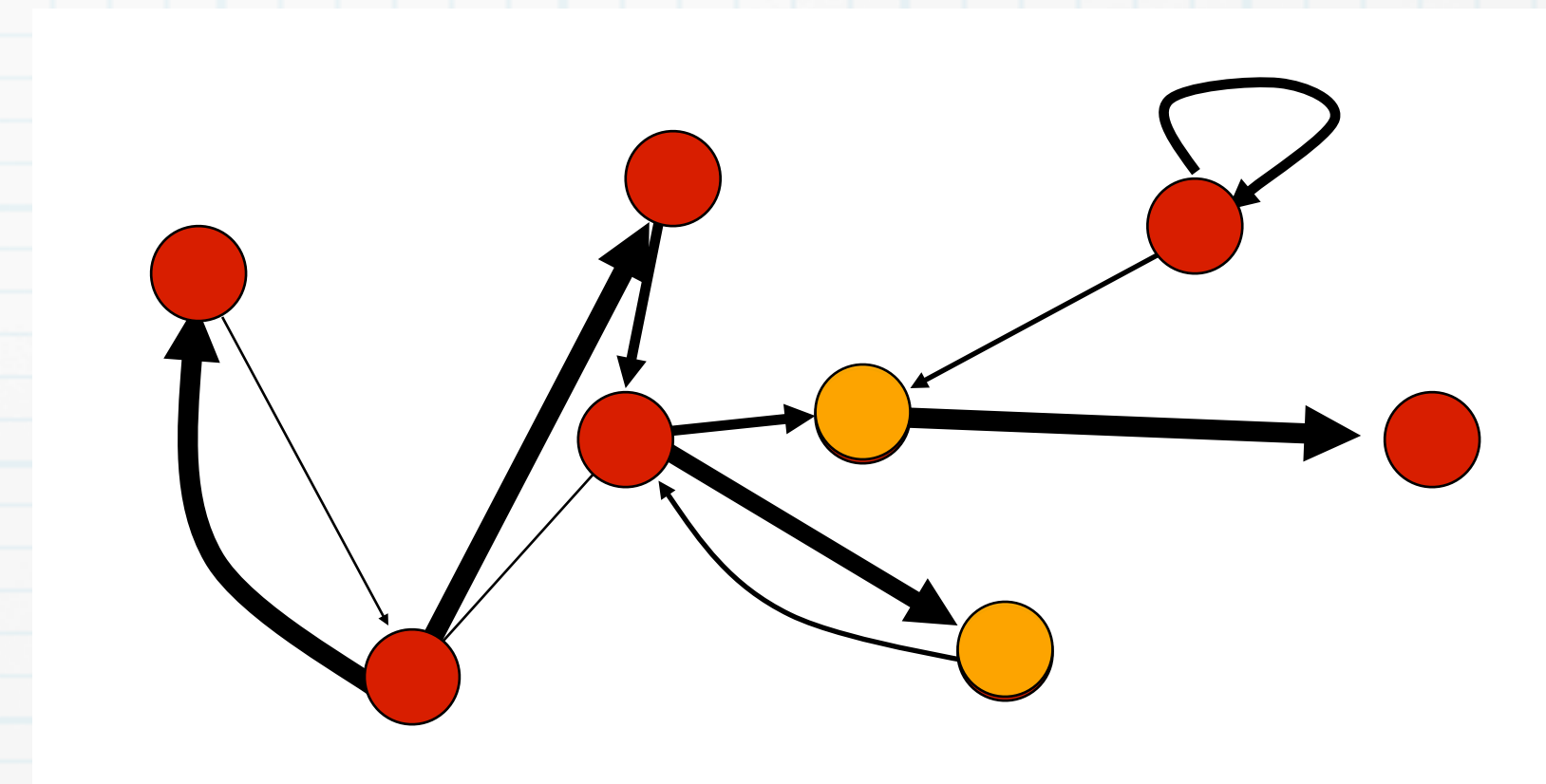
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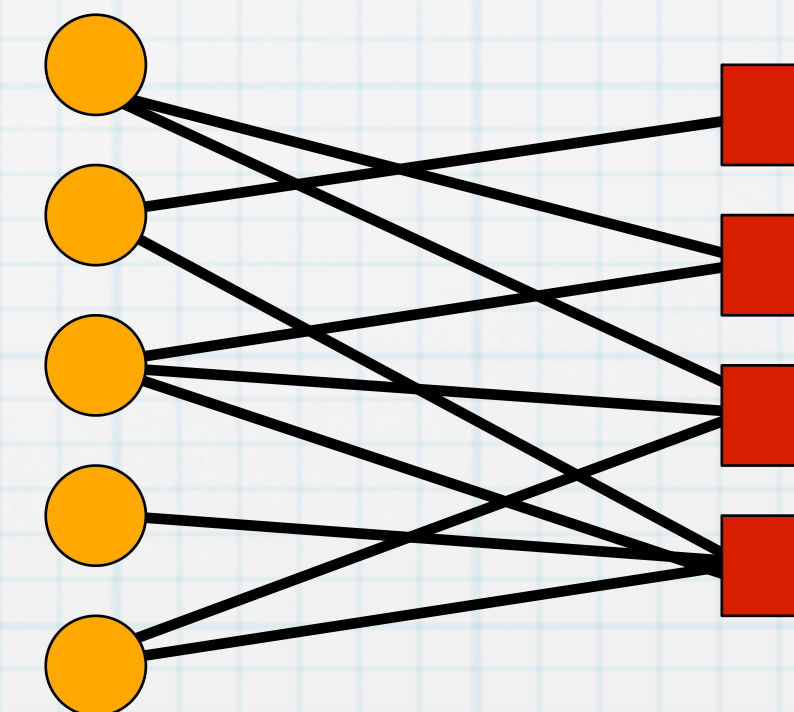
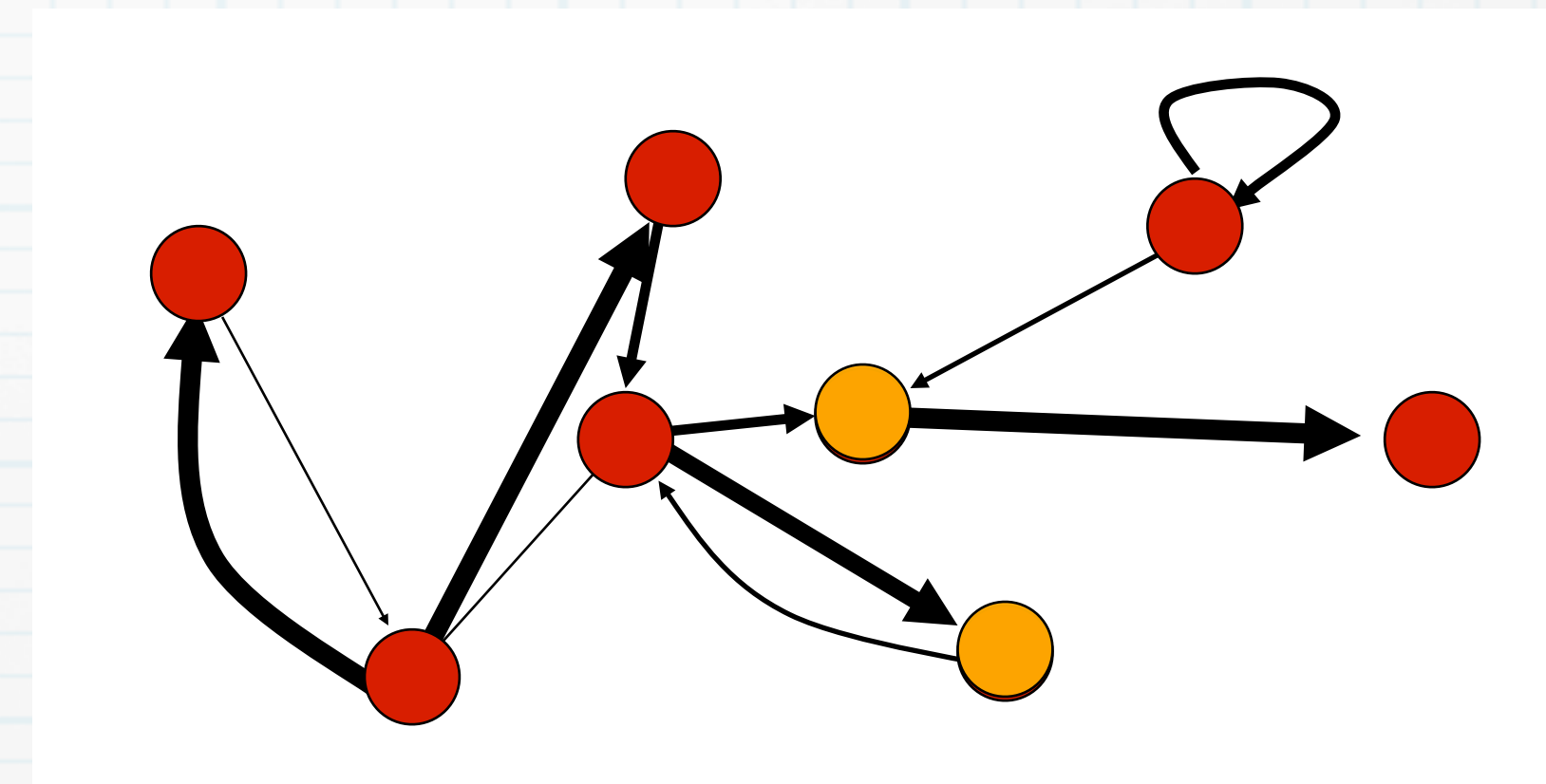
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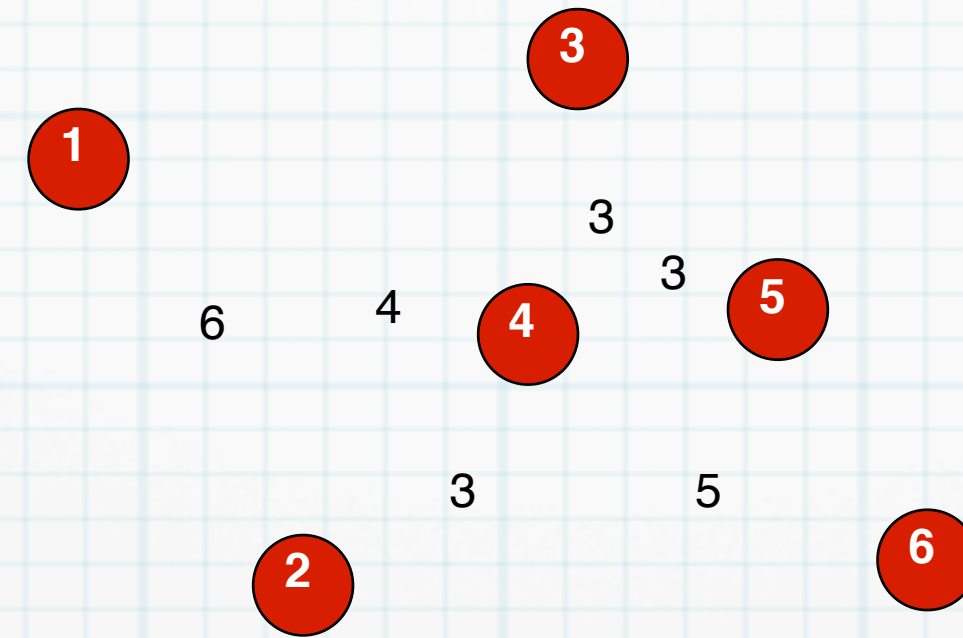


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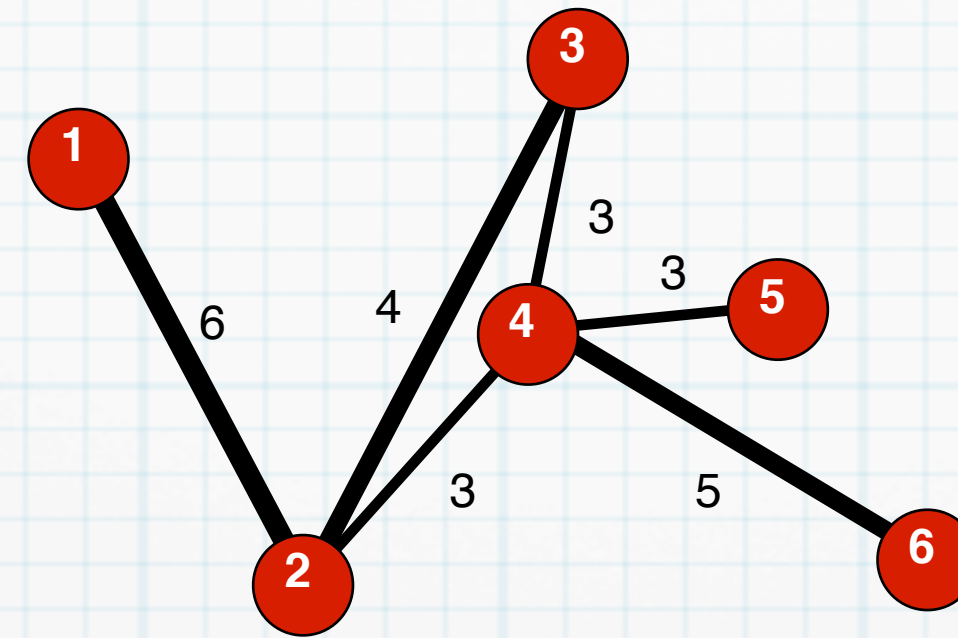


Python and NetworkX



```
W = nx.Graph()
W.add_edge(1,2,weight=6)
...
W.add_weighted_edges_from([(4,5,3),(4,6,5),...])
...
W.edges()
W.edges(data='weight')
for (u,v,d) in W.edges(data='weight'):
    if d>3:
        print('(%d, %d, %d)'%(u,v,d))
B = nx.complete_bipartite_graph(4,5)
```

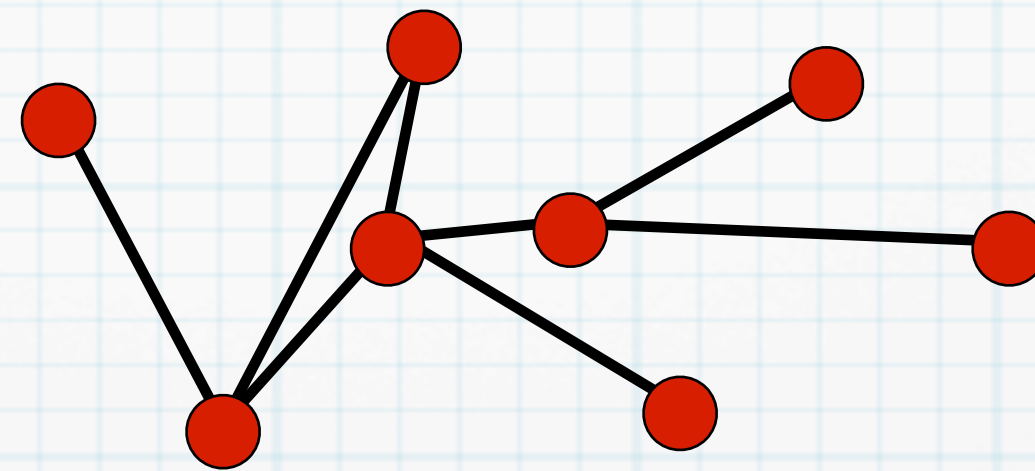

Python and NetworkX



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More definitions



- * subgraph

- * complete graph (clique)

- * size $N \rightarrow E = N * (N-1) / 2$

- * tree: connected, no cycles

- * size $N \rightarrow E = N-1$

More definitions

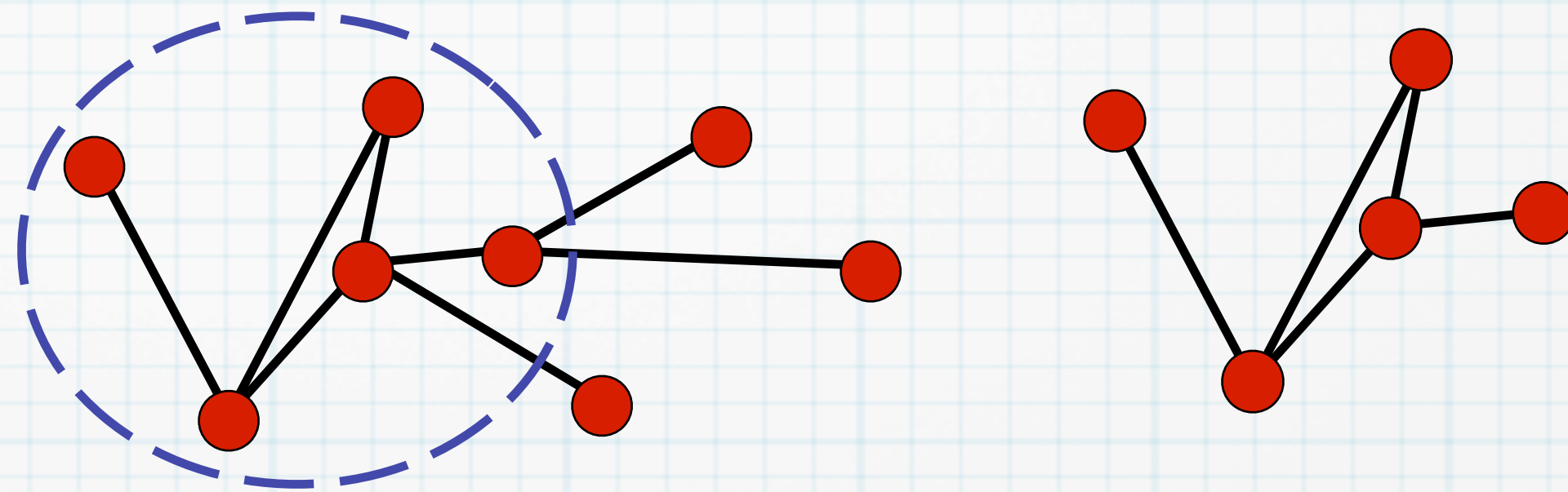
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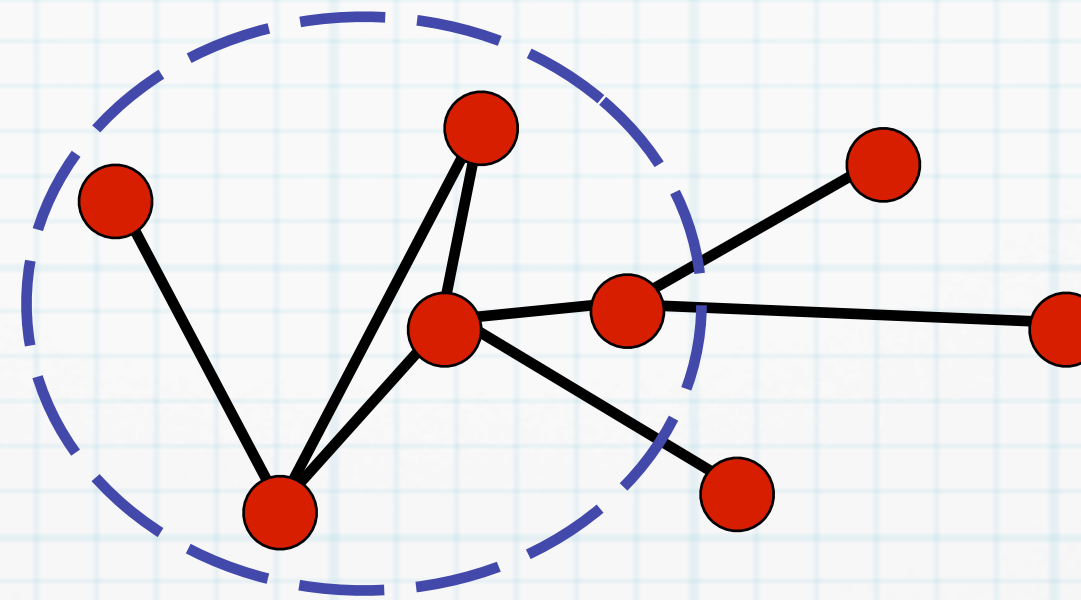
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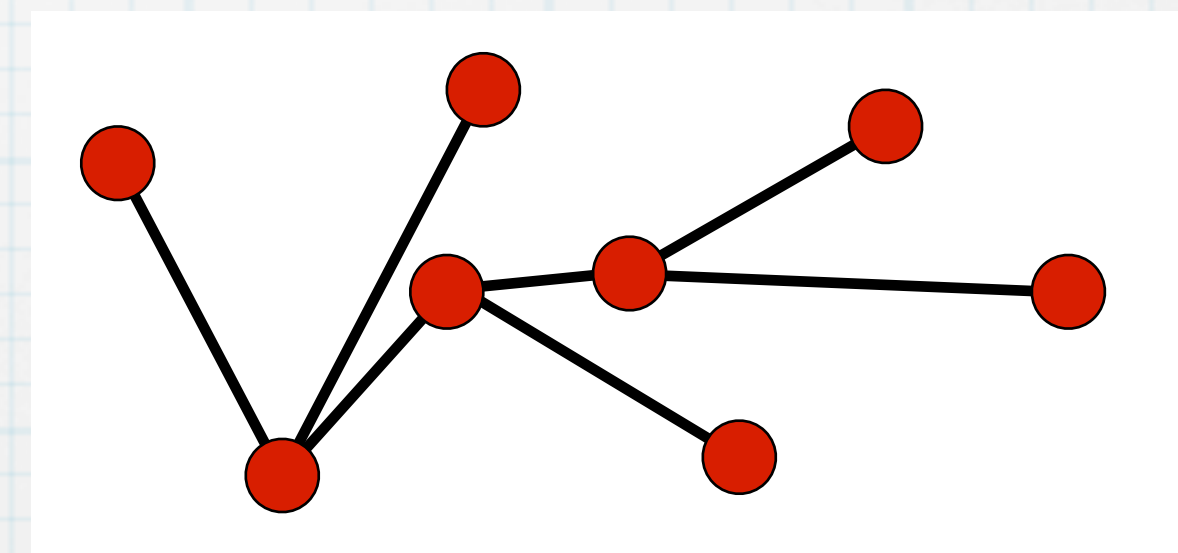


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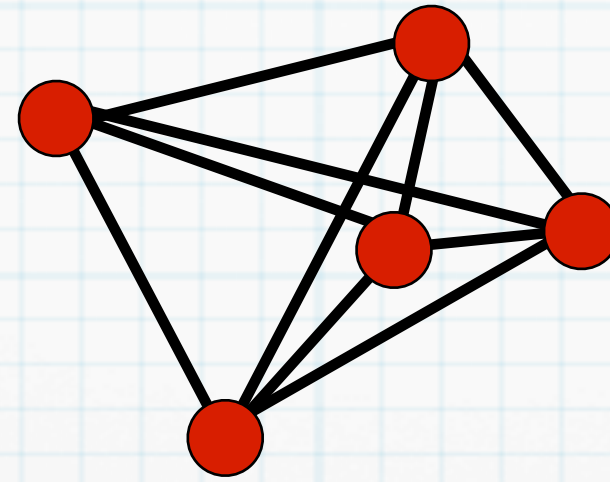
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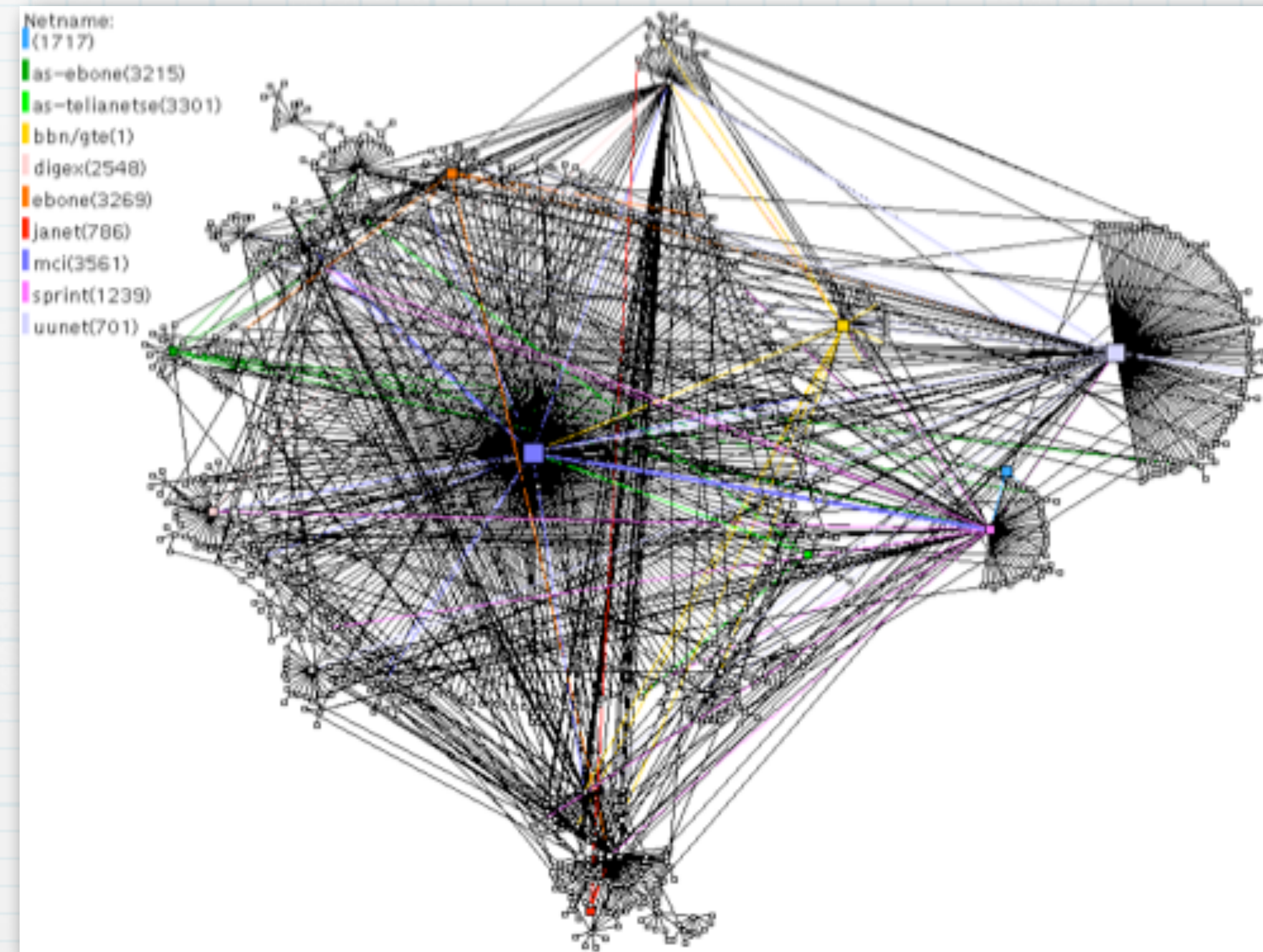
Python and NetworkX



```
K5 = nx.complete_graph(5)
K5.edges()
clique = nx.subgraph(K5, (0,1,2))
nx.is_tree(K5)
c = nx.cycle_graph(4)
nx.is_tree(c)
p = nx.path_graph(5)
nx.is_tree(p)
s = nx.star_graph(6)
nx.is_tree(s)
```


Centrality

Degree as a measure of influence, importance, prestige...

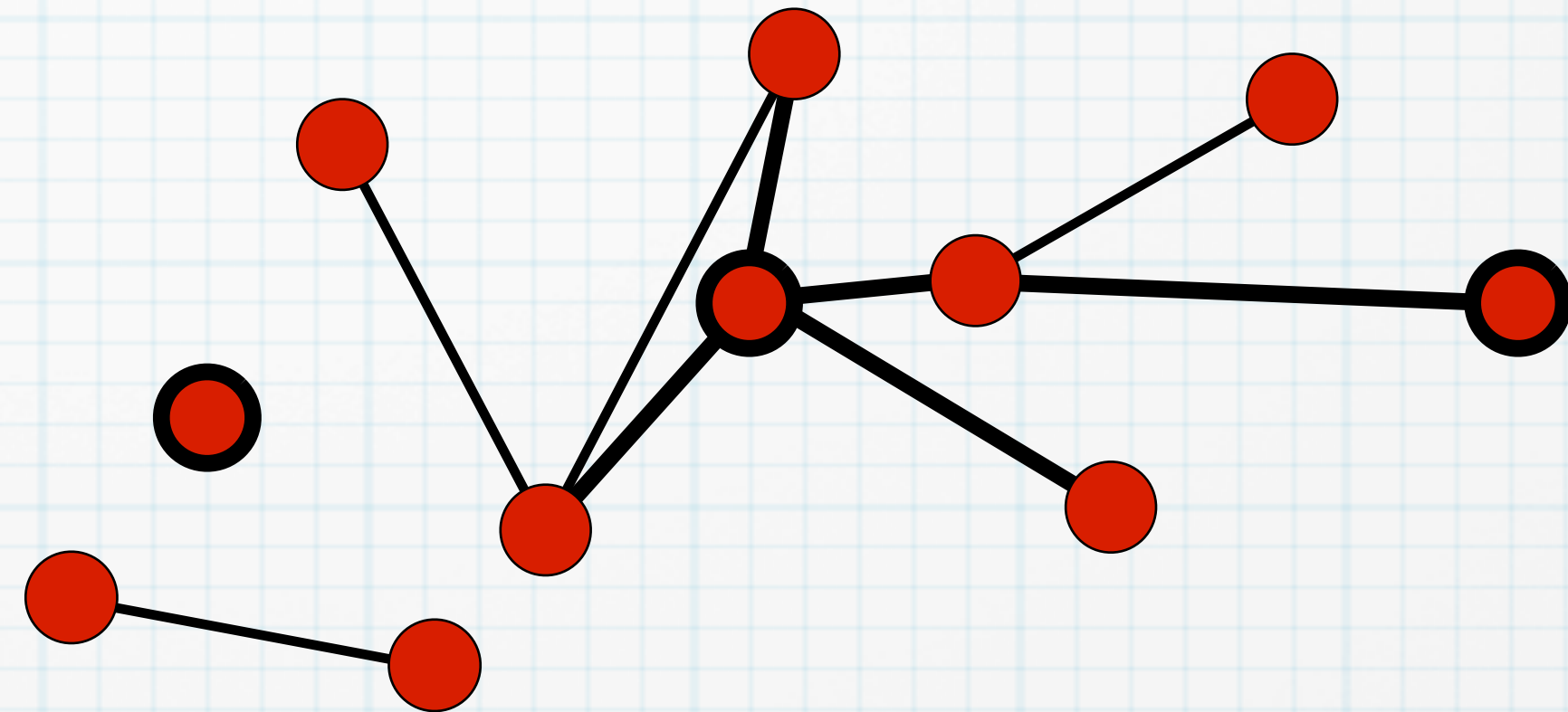


More definitions

- * degree k

- * indegree k_{in}

- * outdegree k_{out}

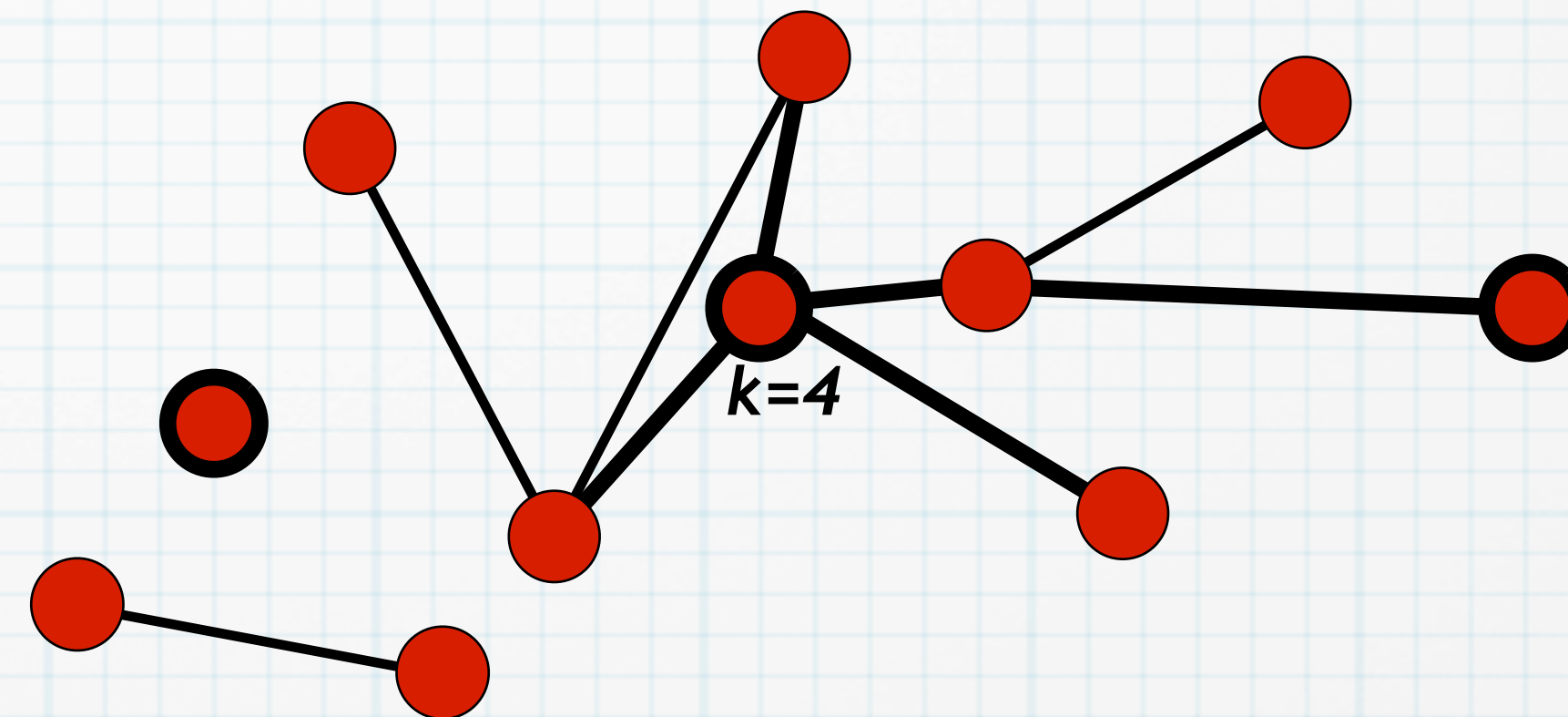


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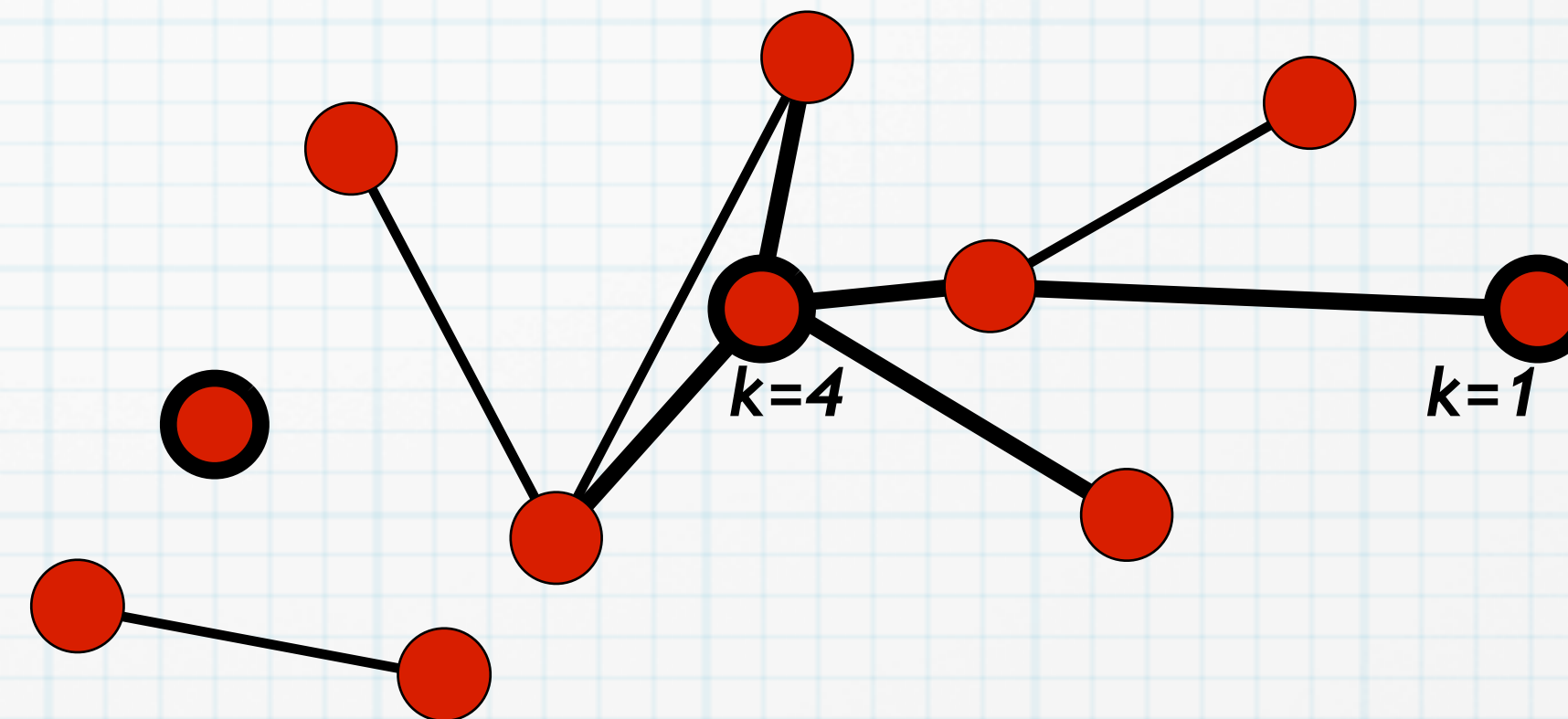


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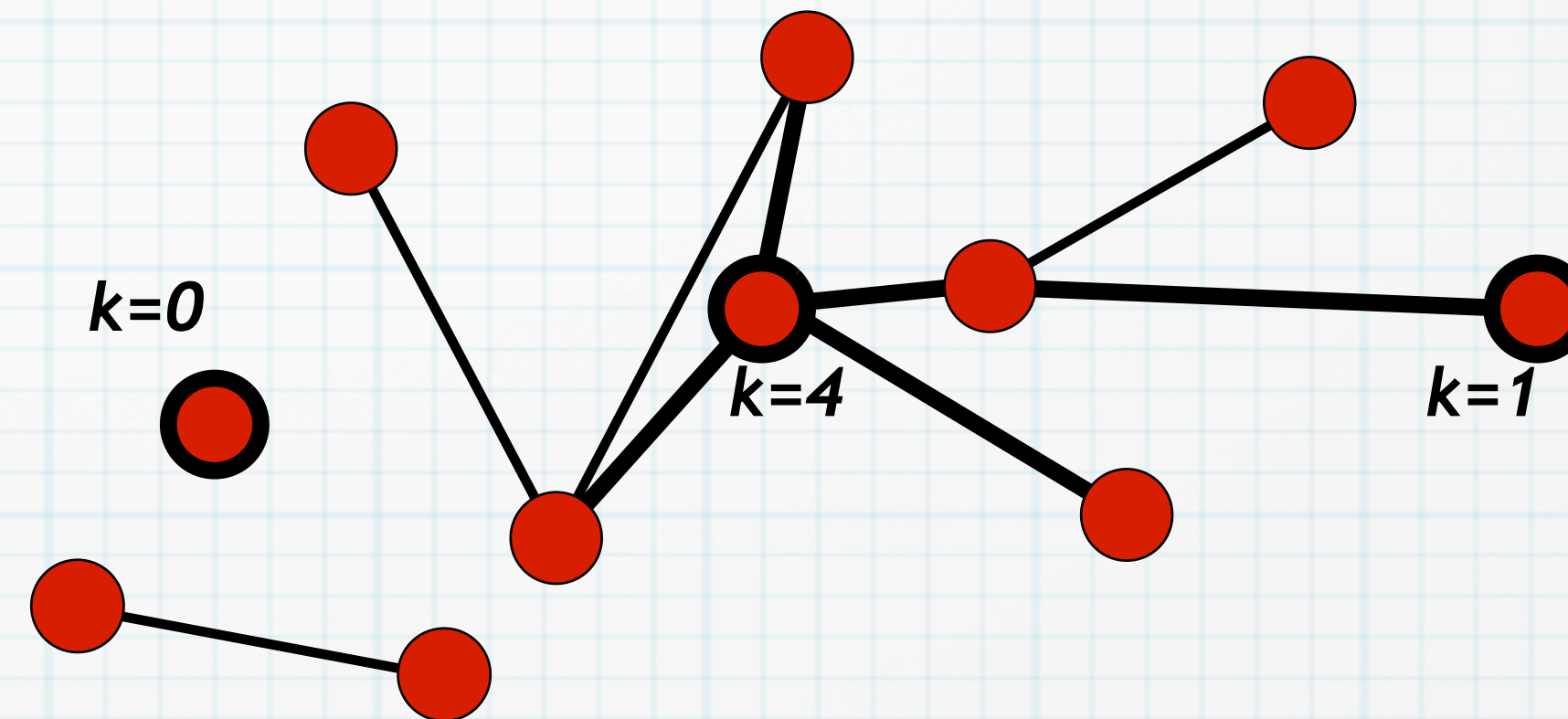


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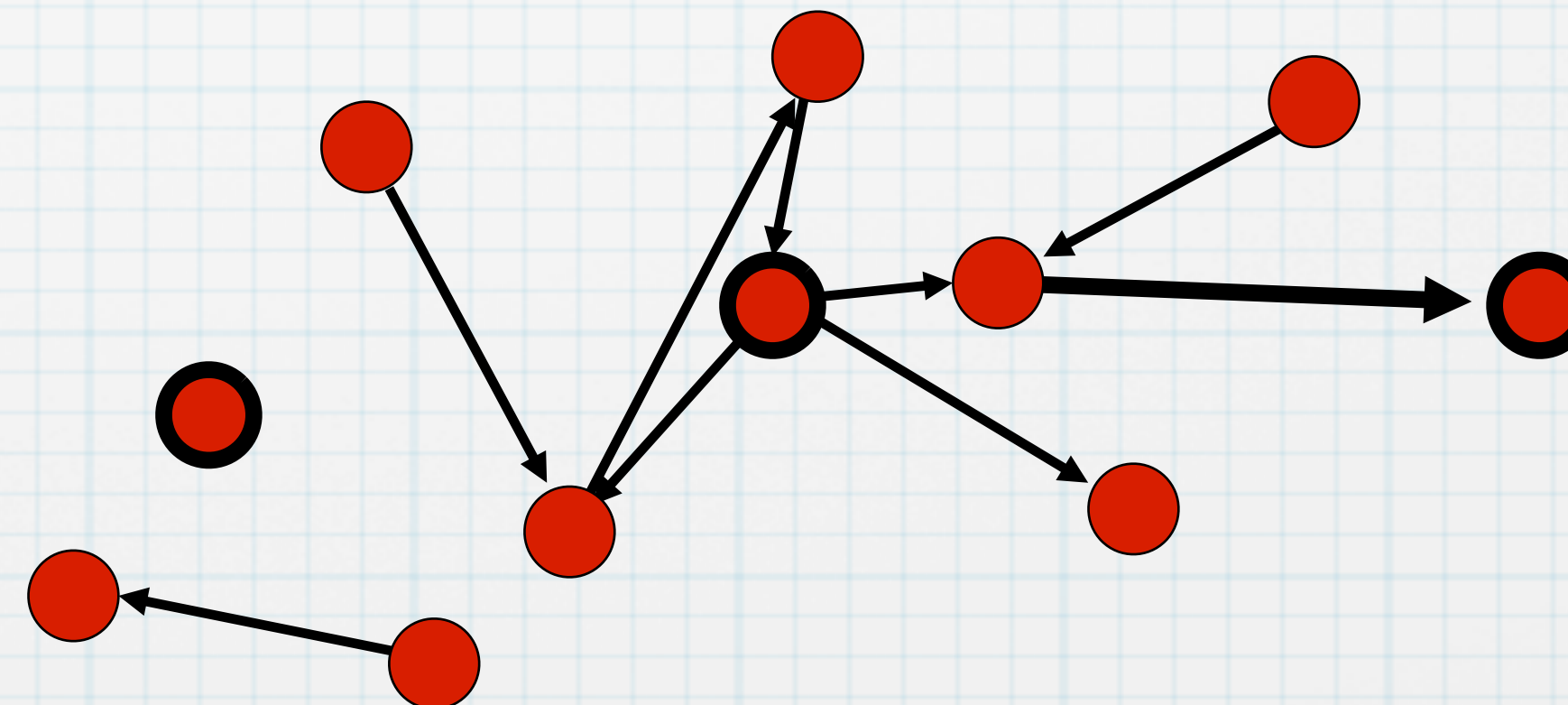
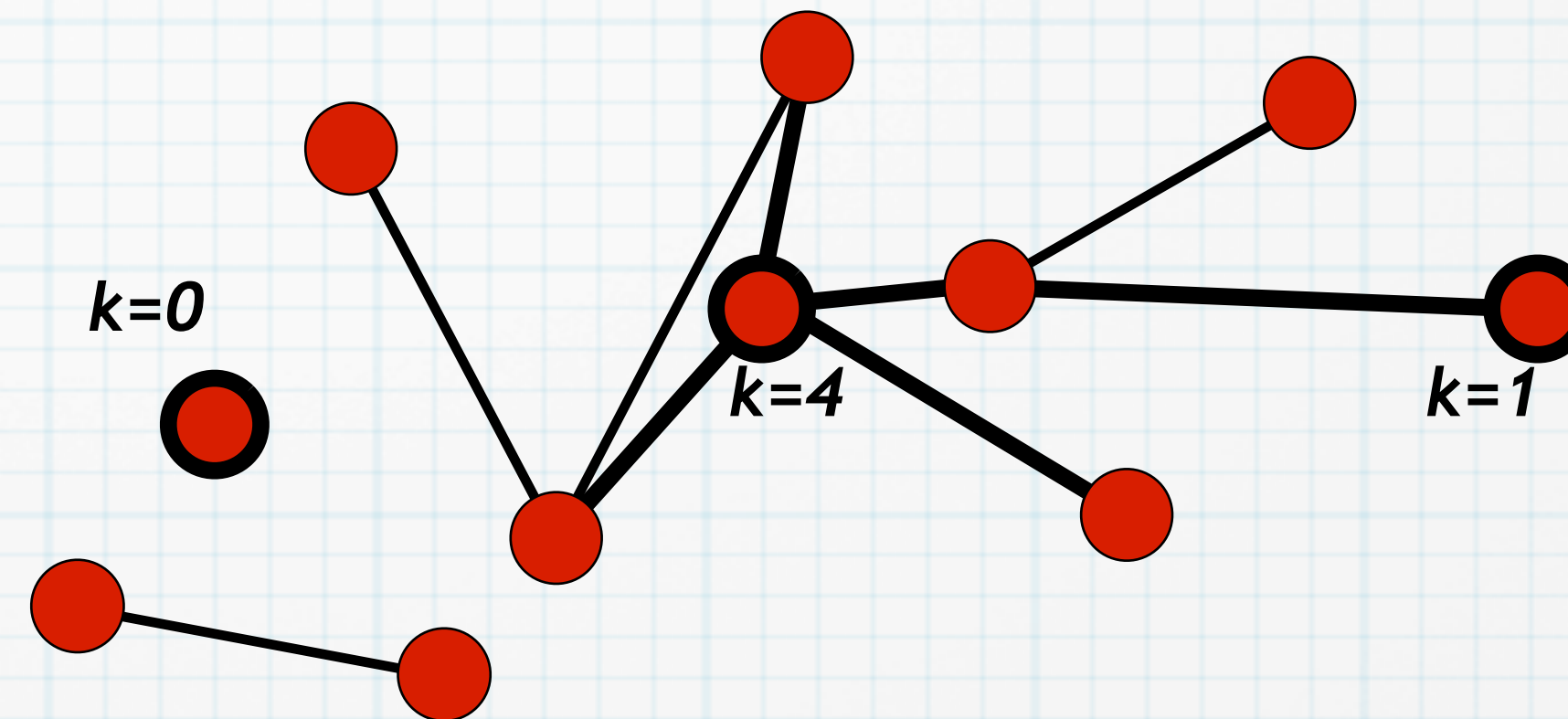


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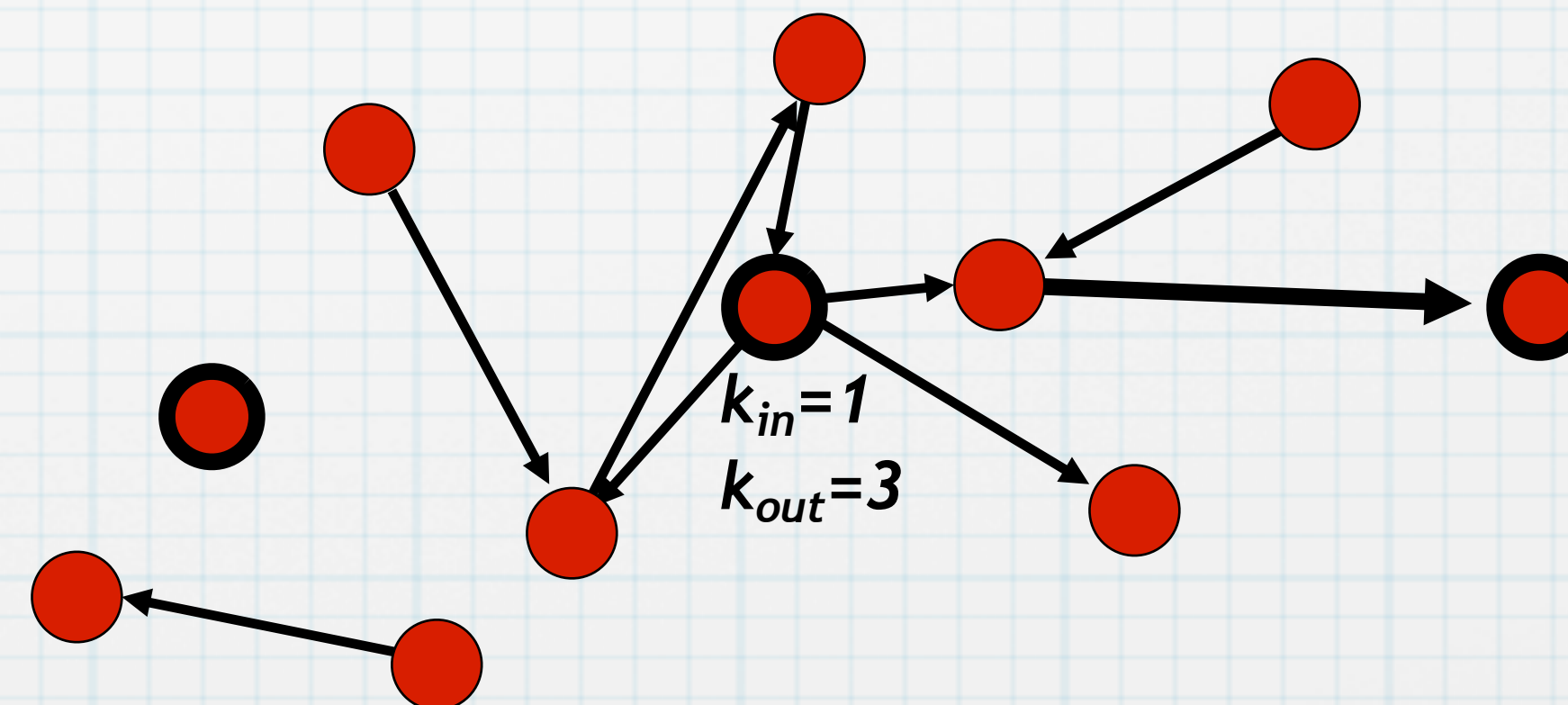
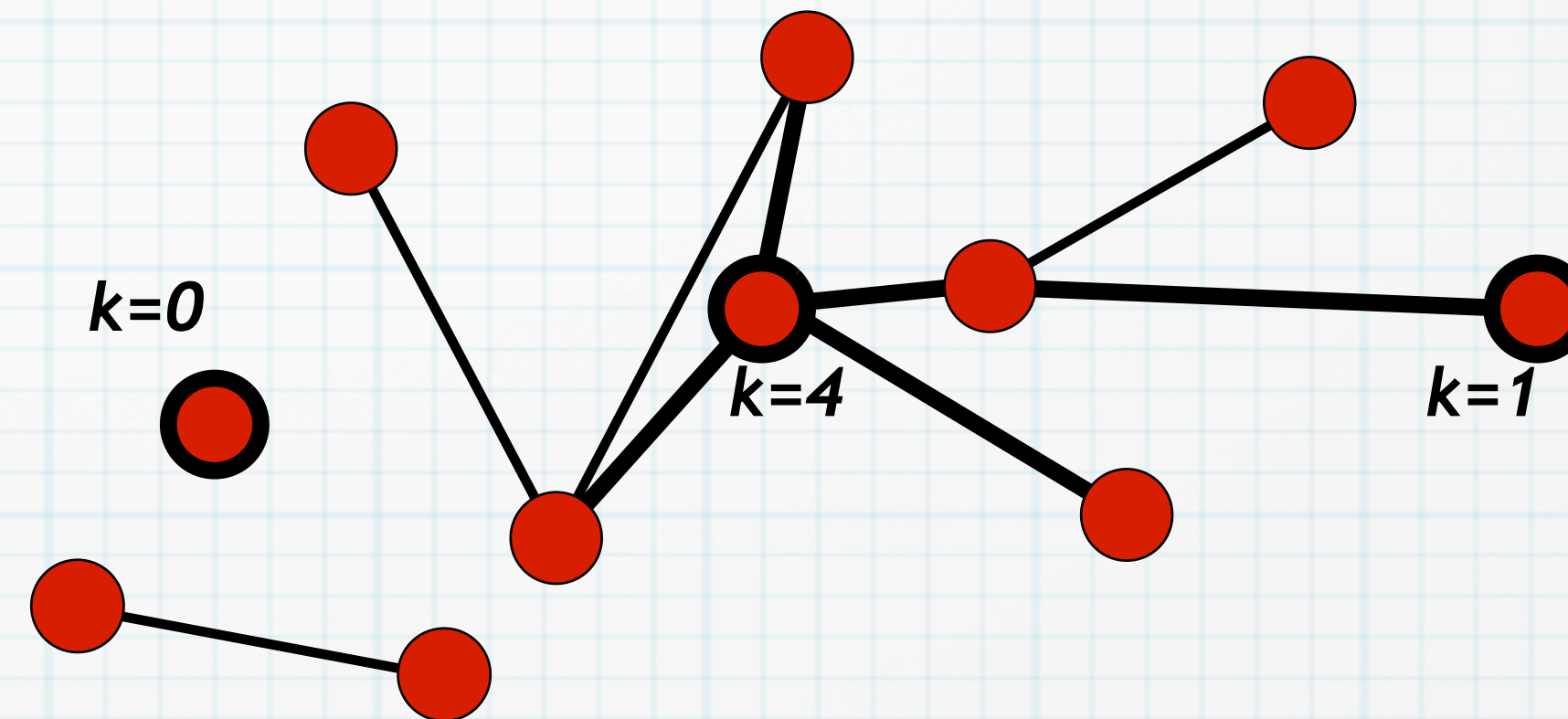


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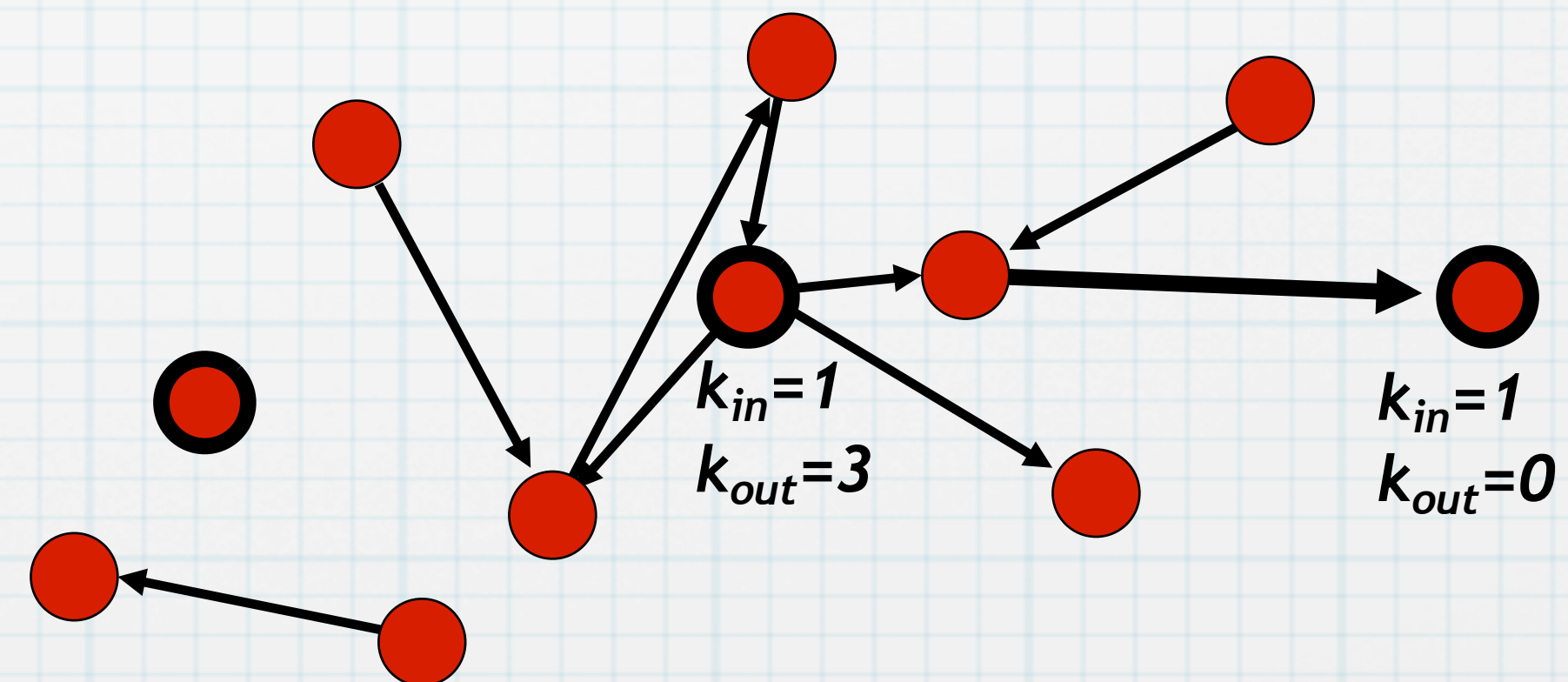
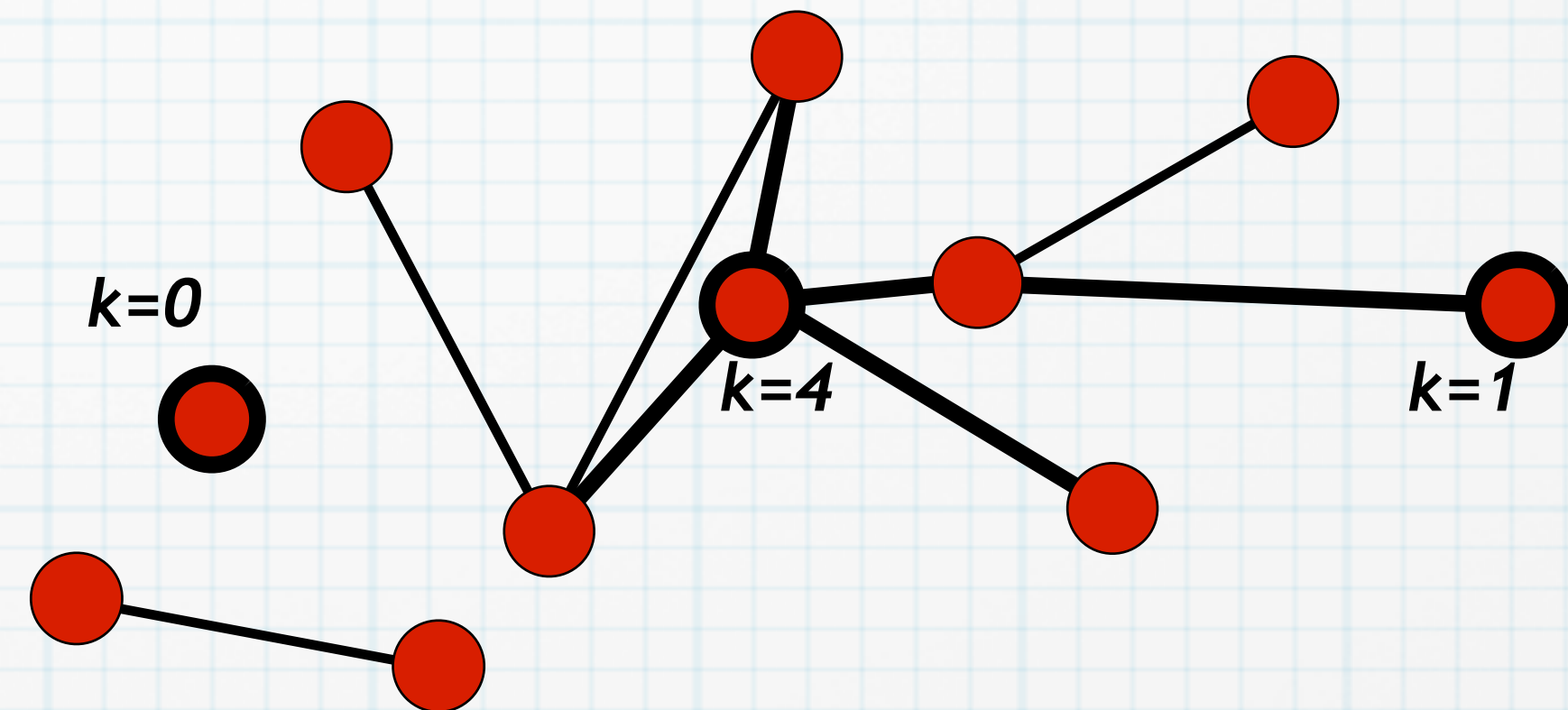


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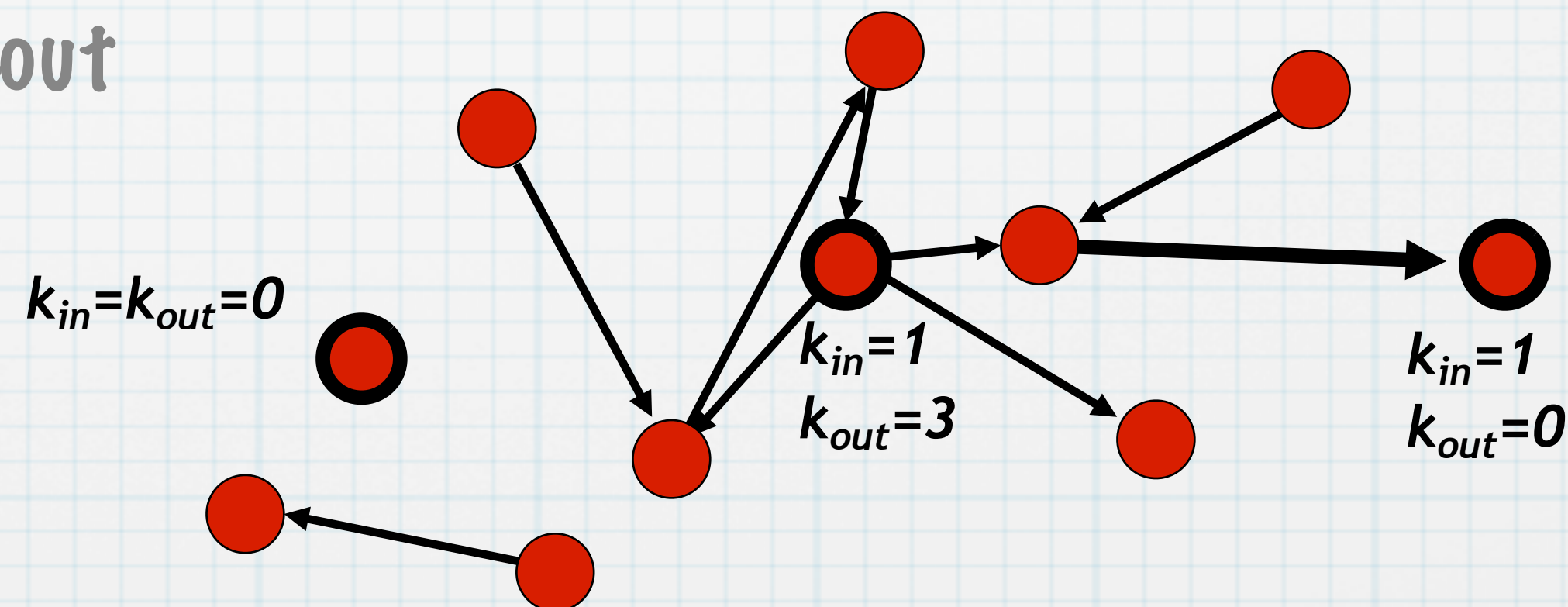
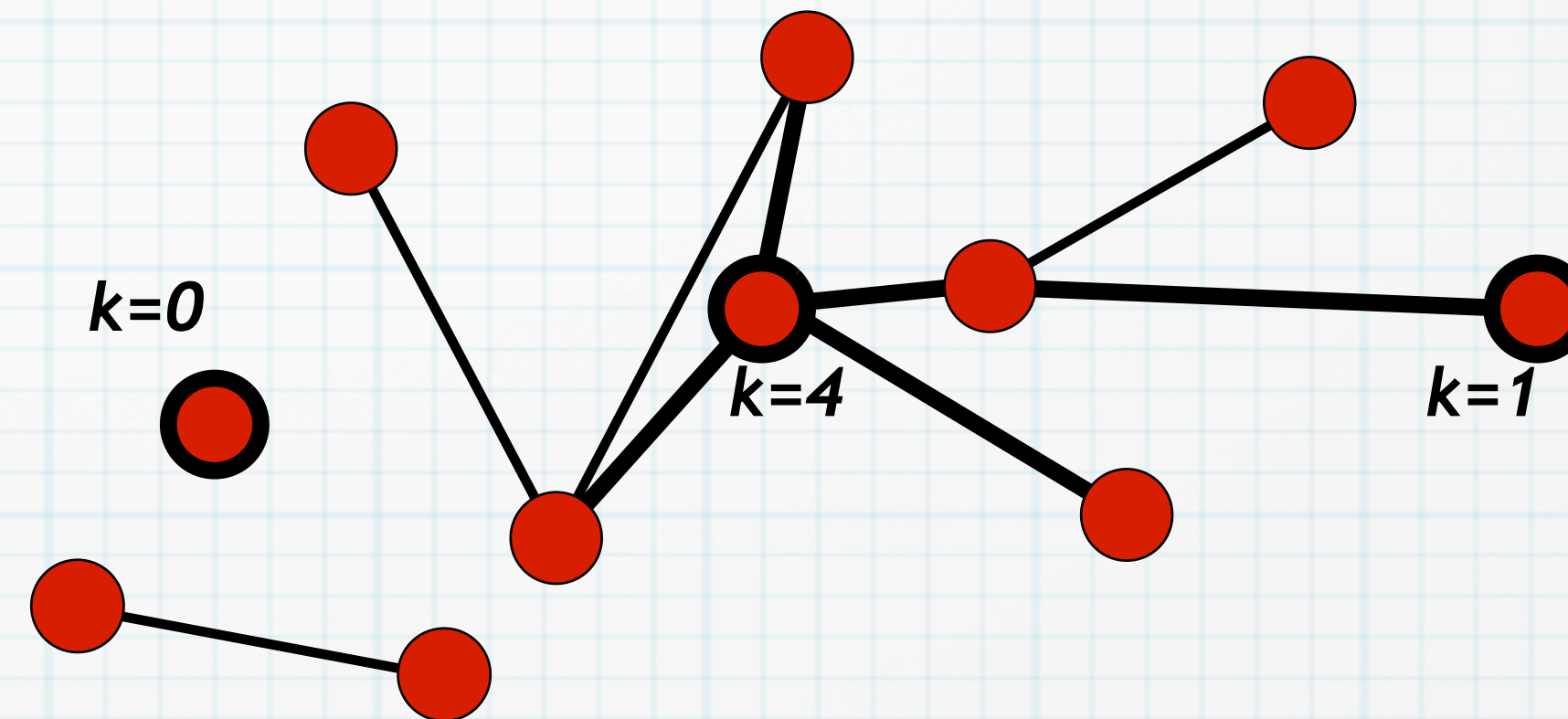


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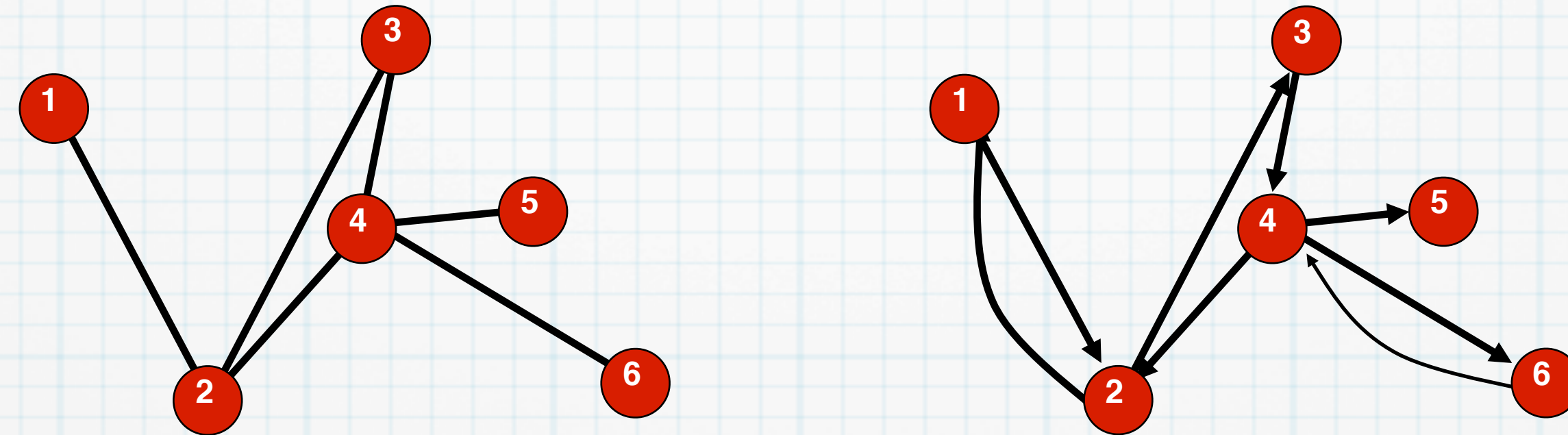
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Python and NetworkX



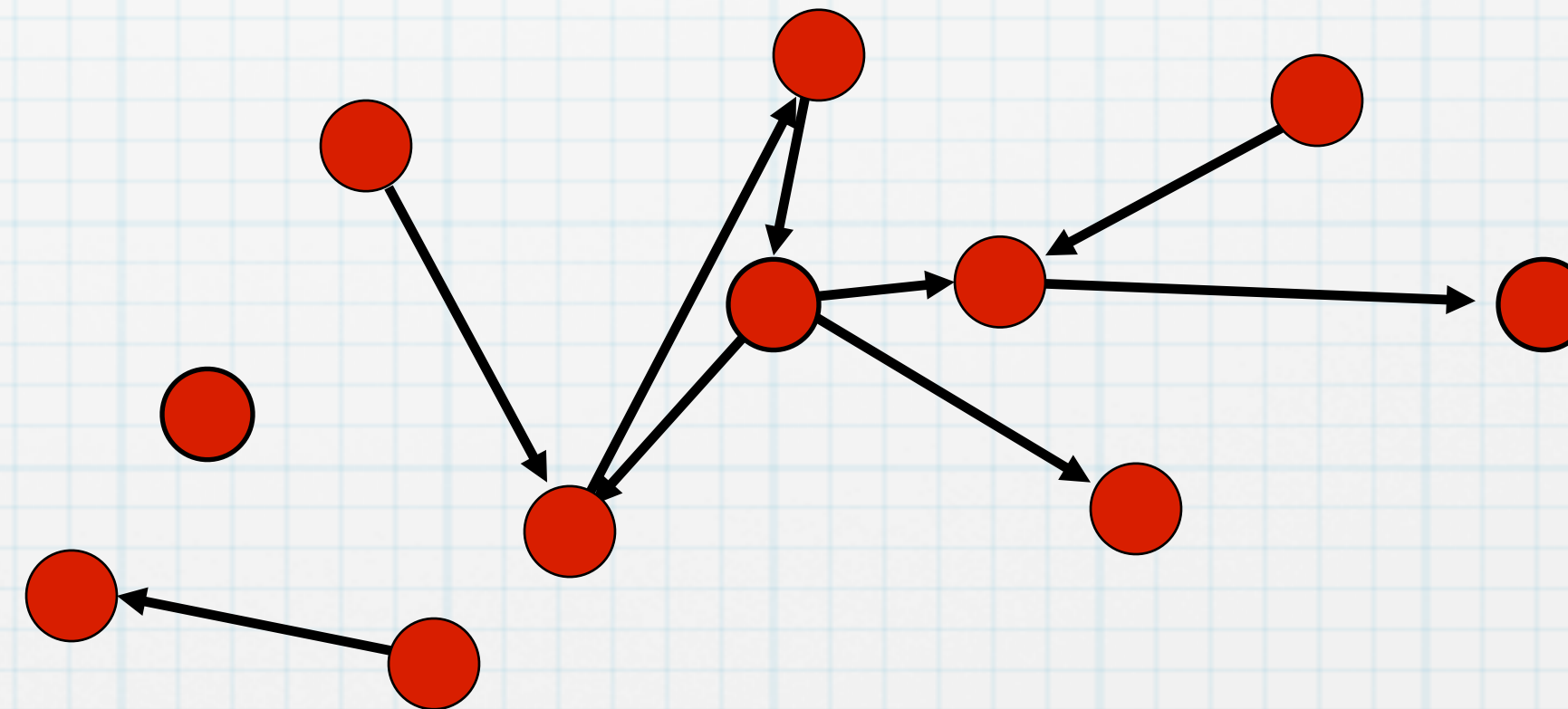
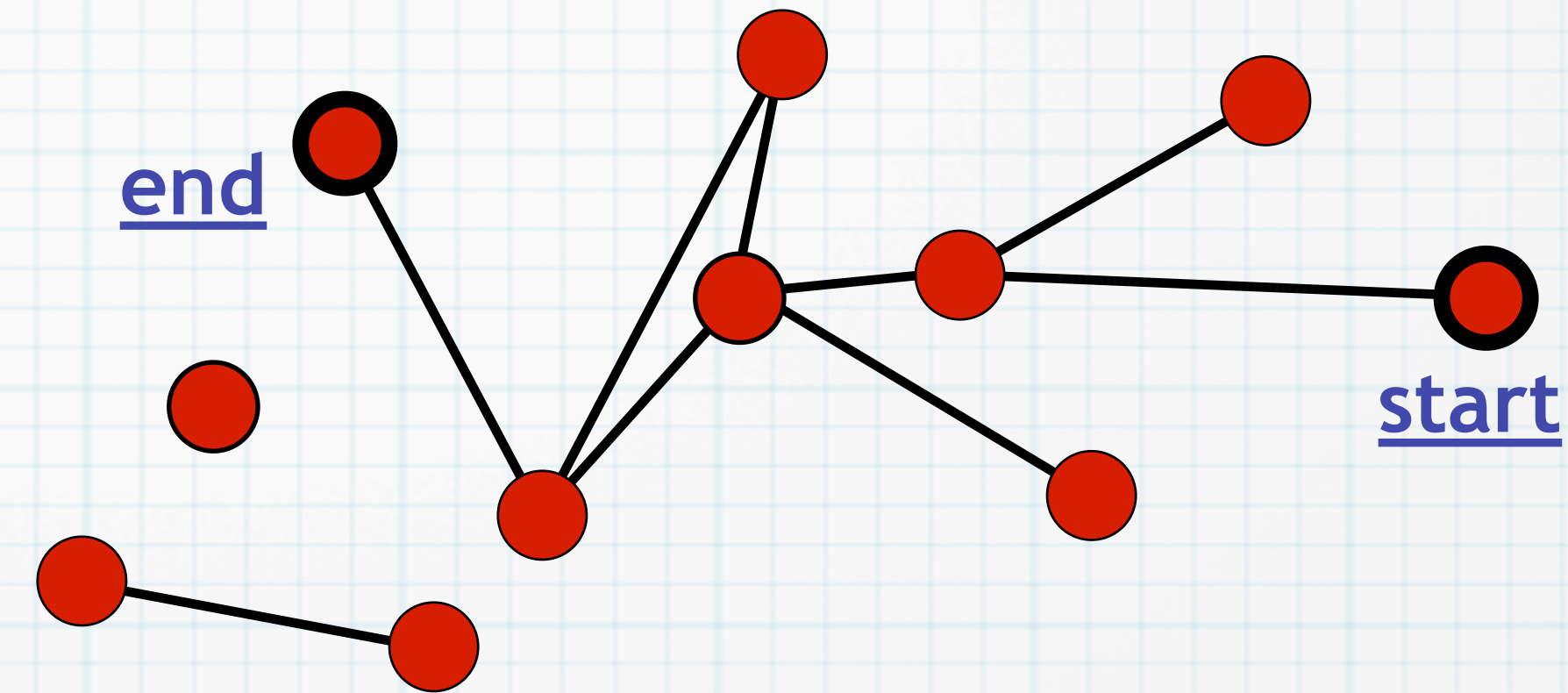
```
G.degree(2)
G.degree()
G.degree(4)

D.degree(4)
D.in_degree(4)
D.out_degree(4)
```


Distance

* path

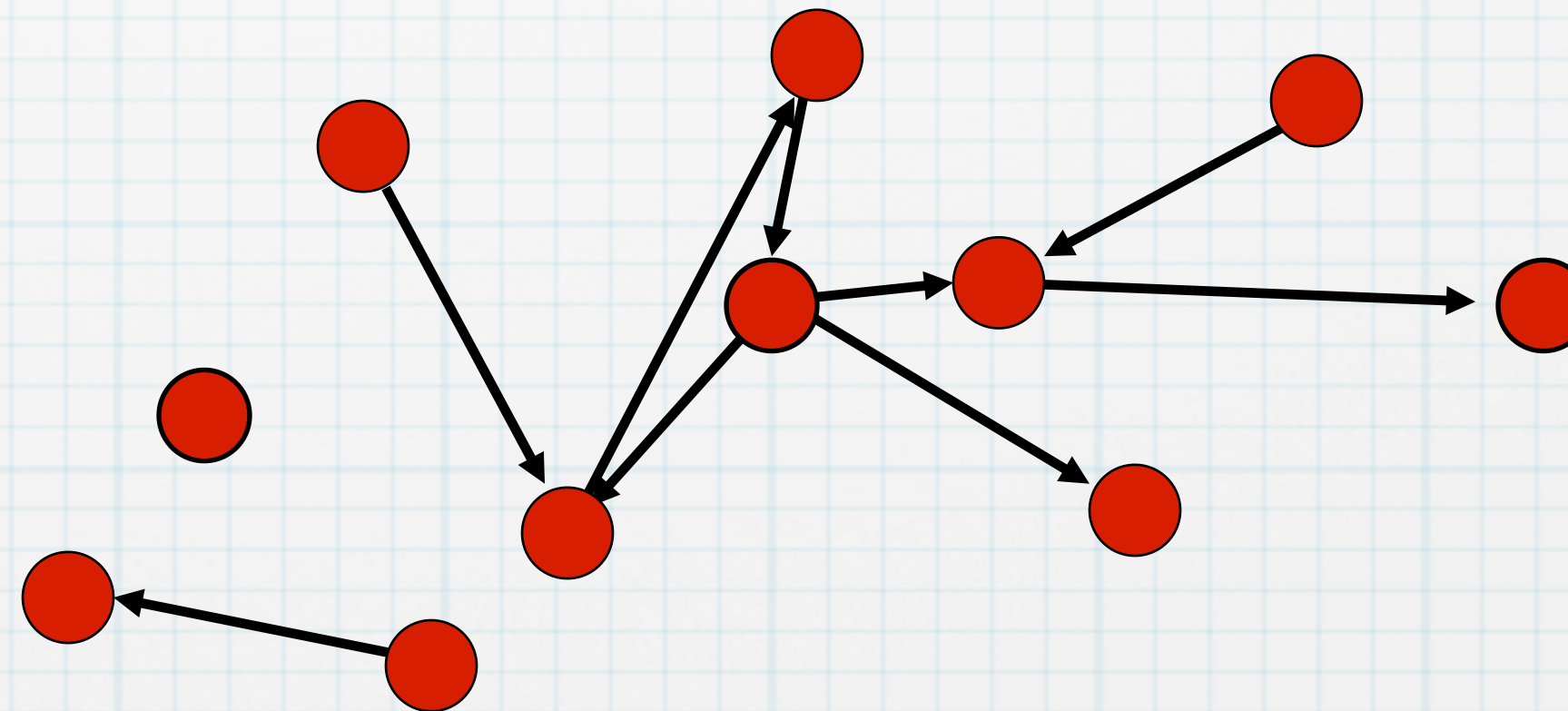
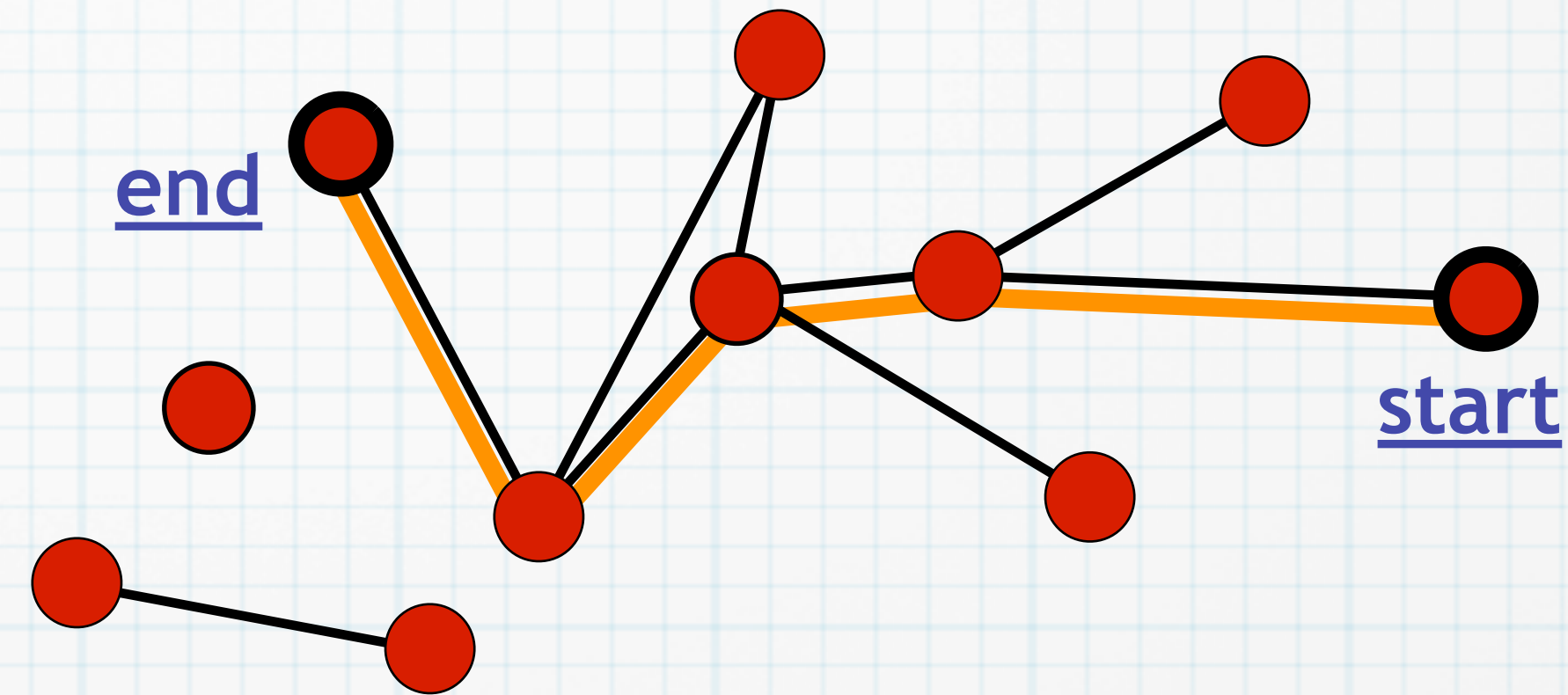
* shortest path



Distance

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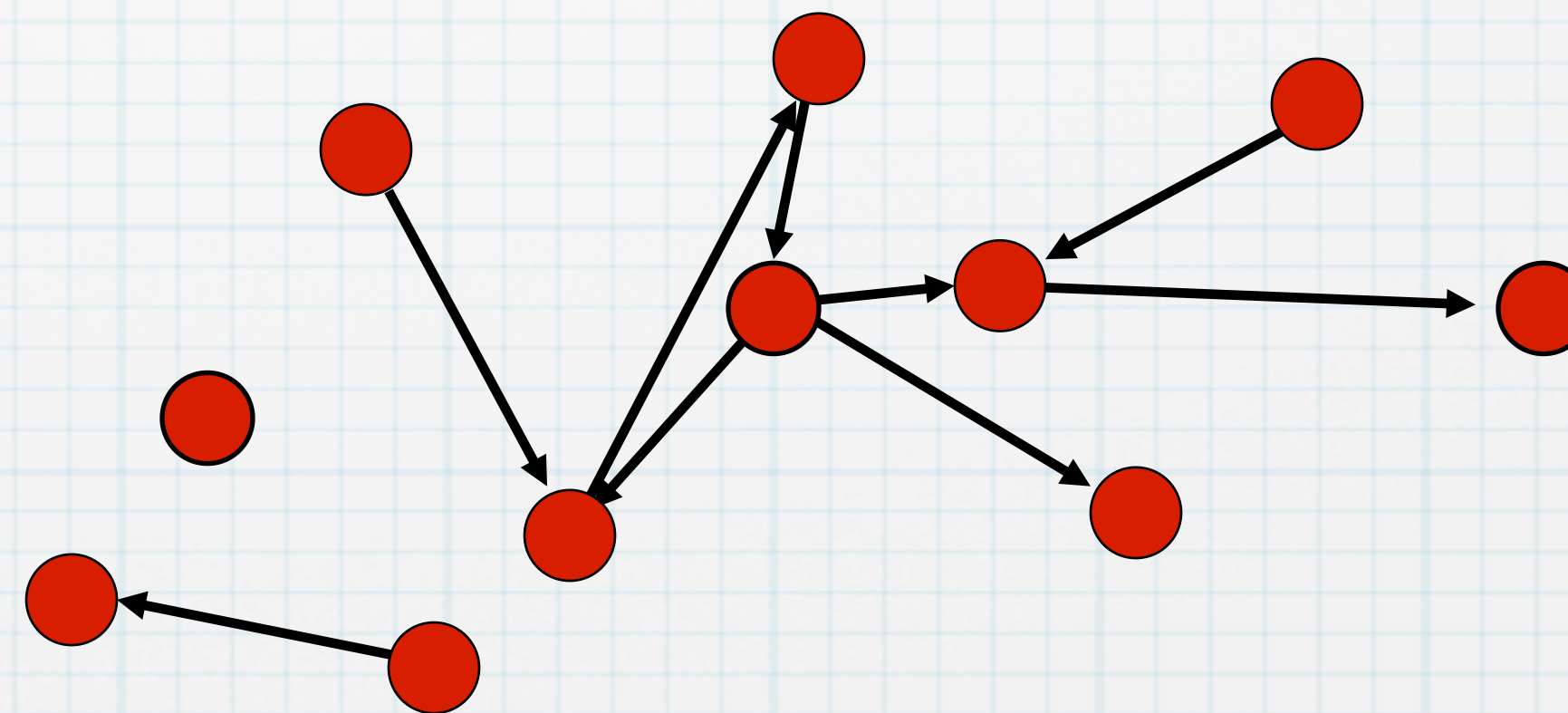
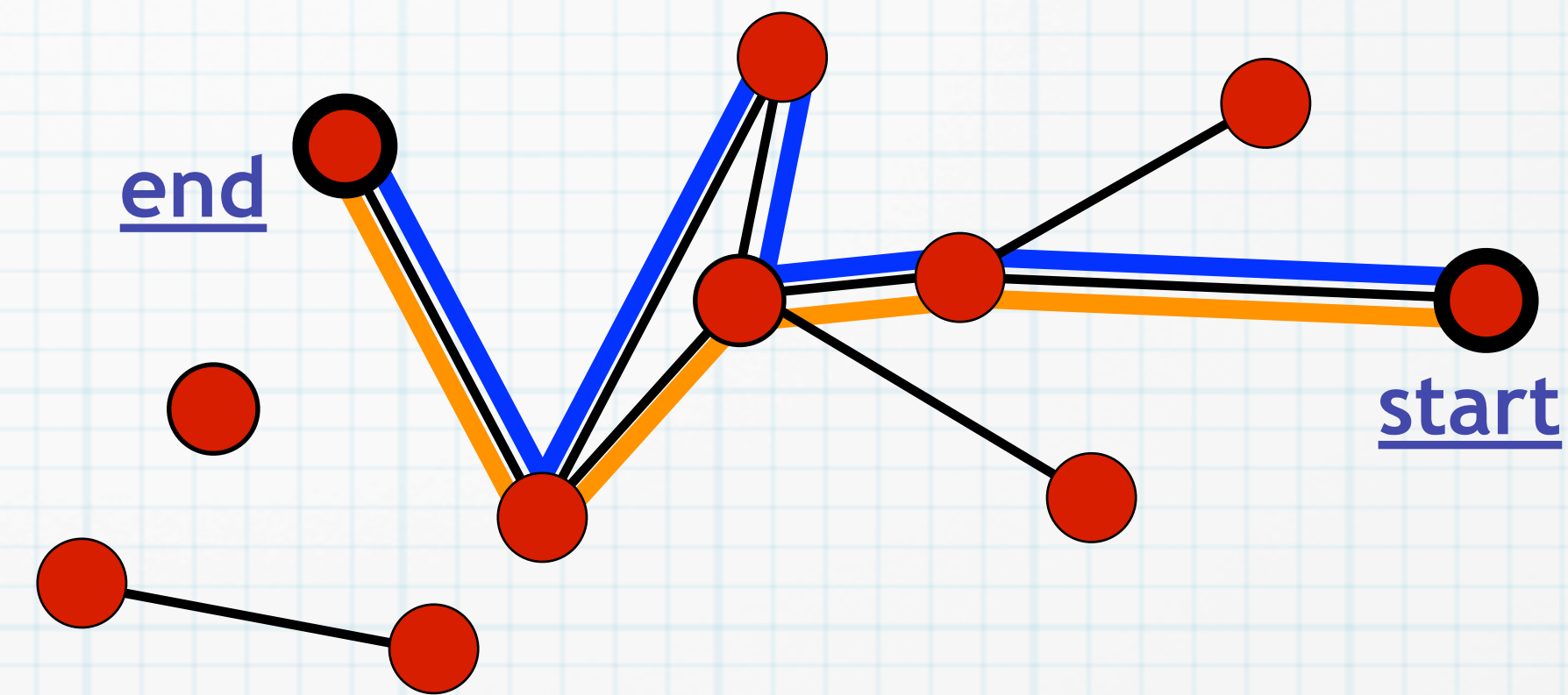
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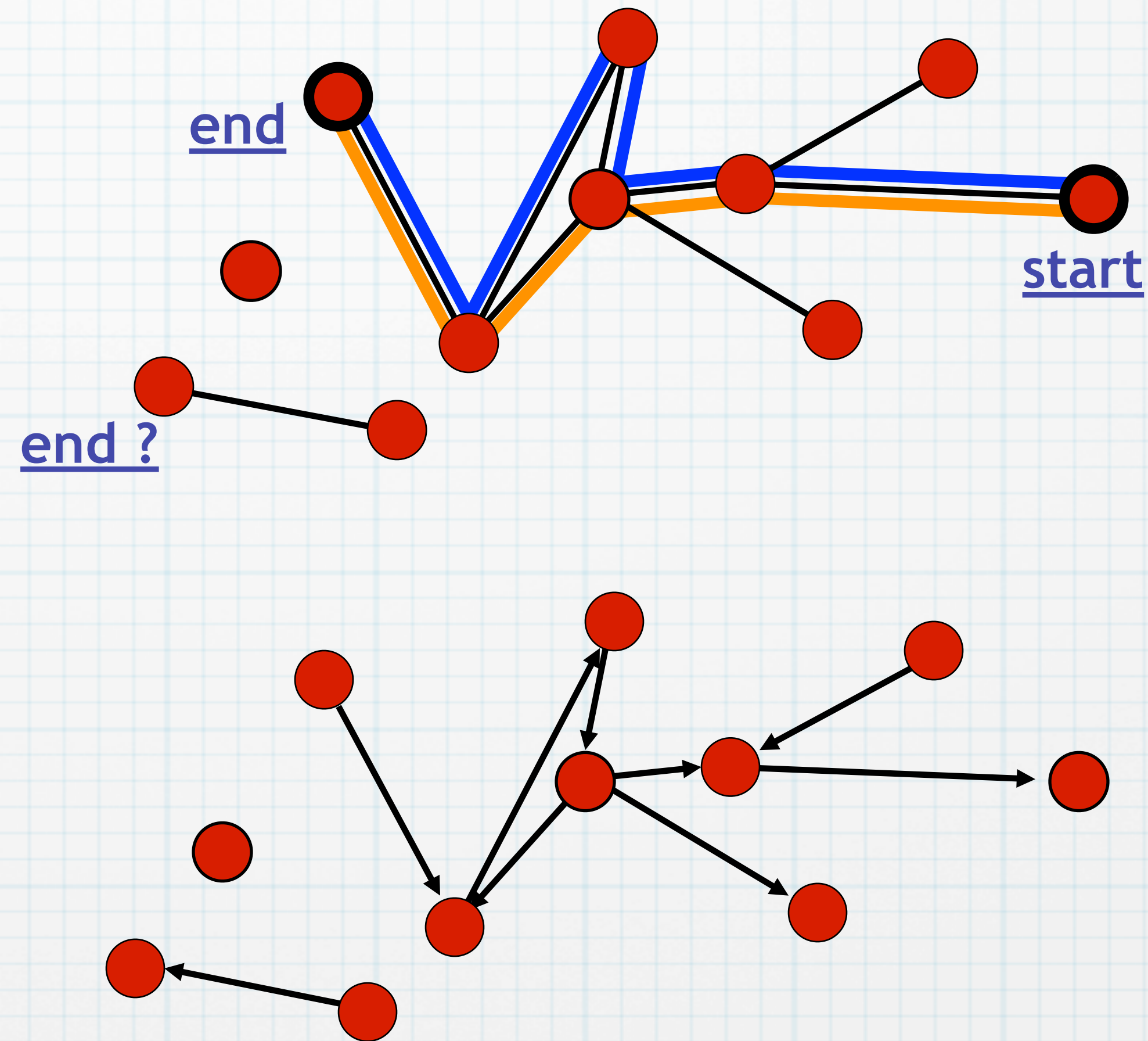
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Distance

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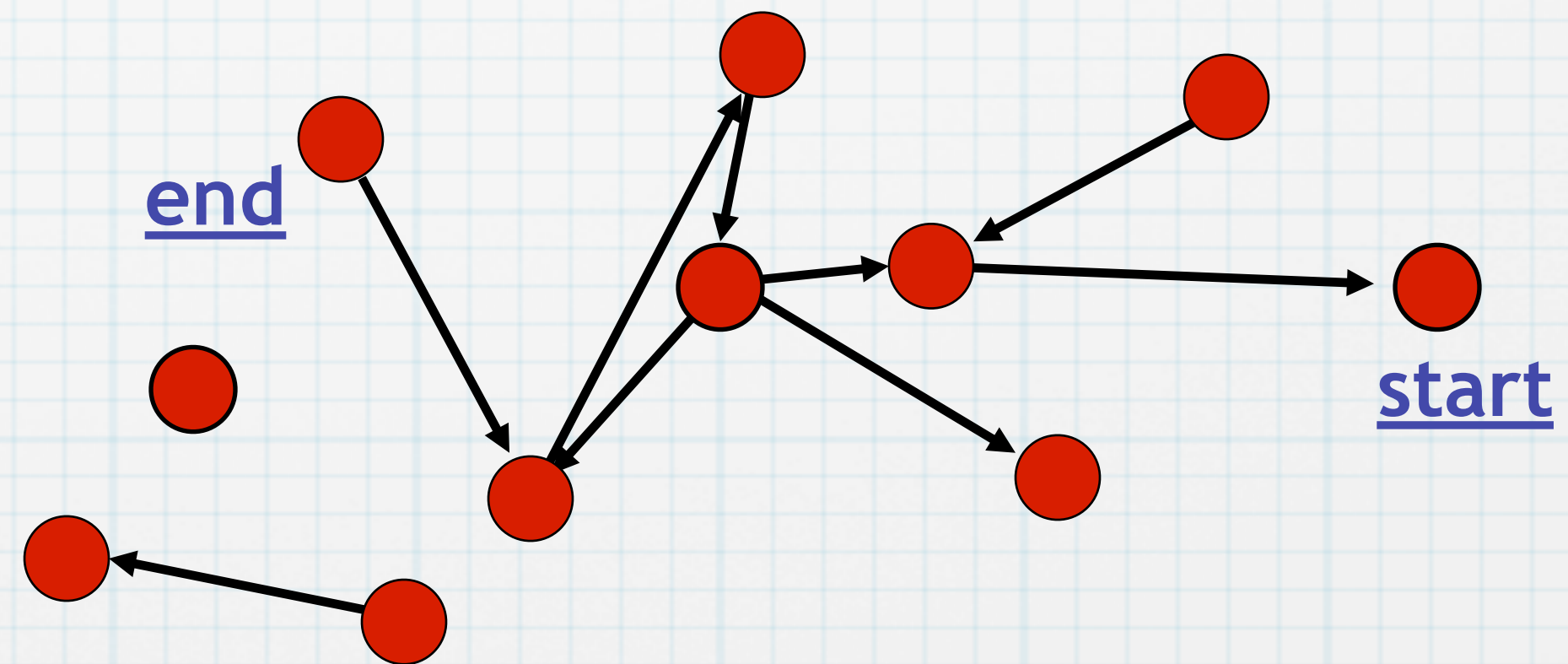
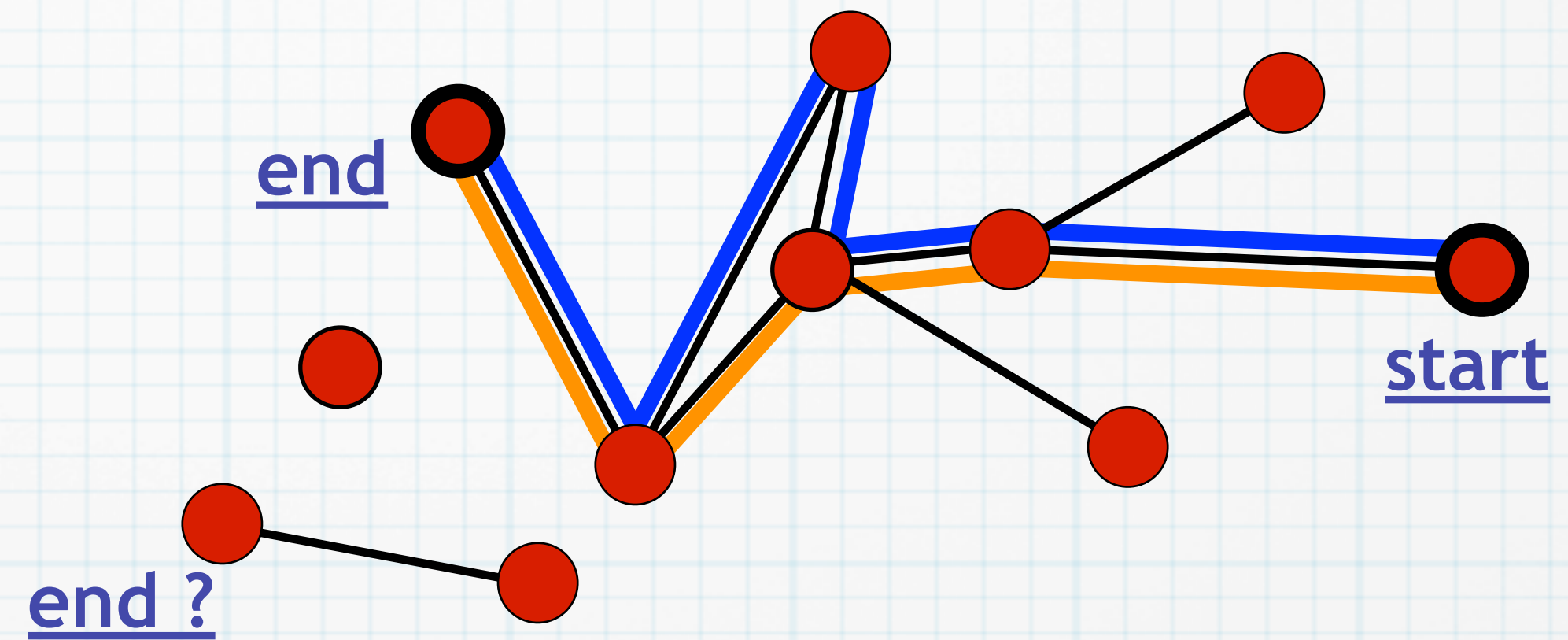
* shortest path



Distance

* path

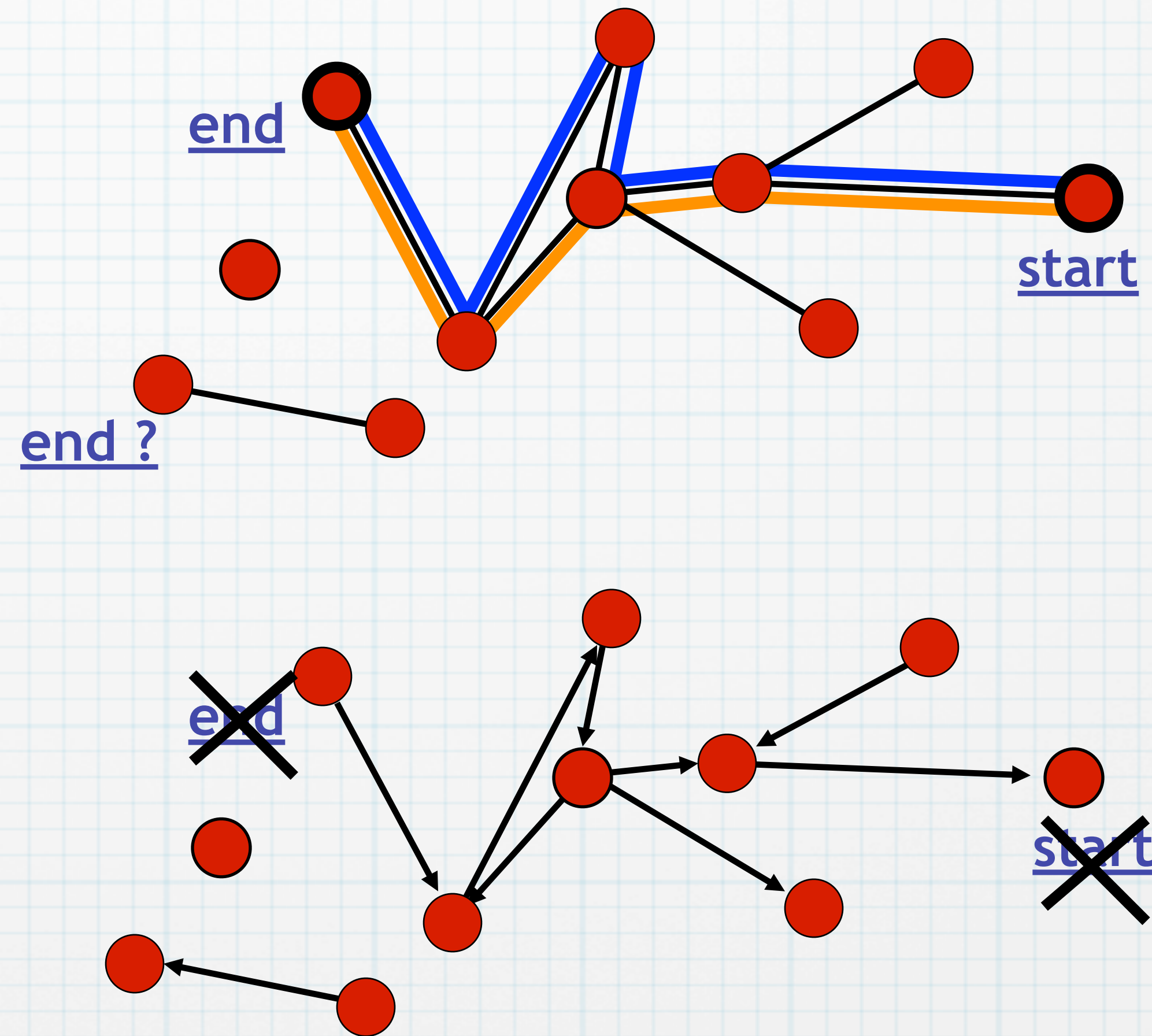
* shortest path



Distance

* path

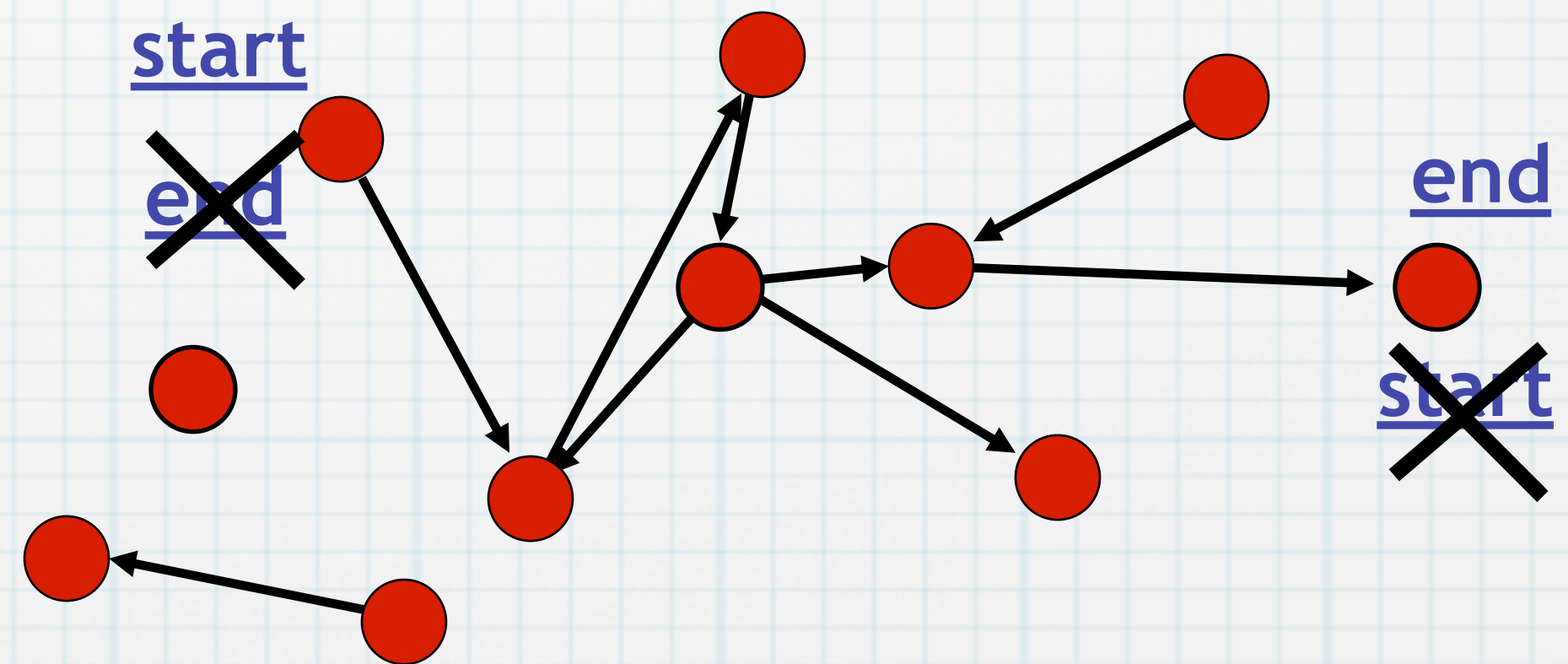
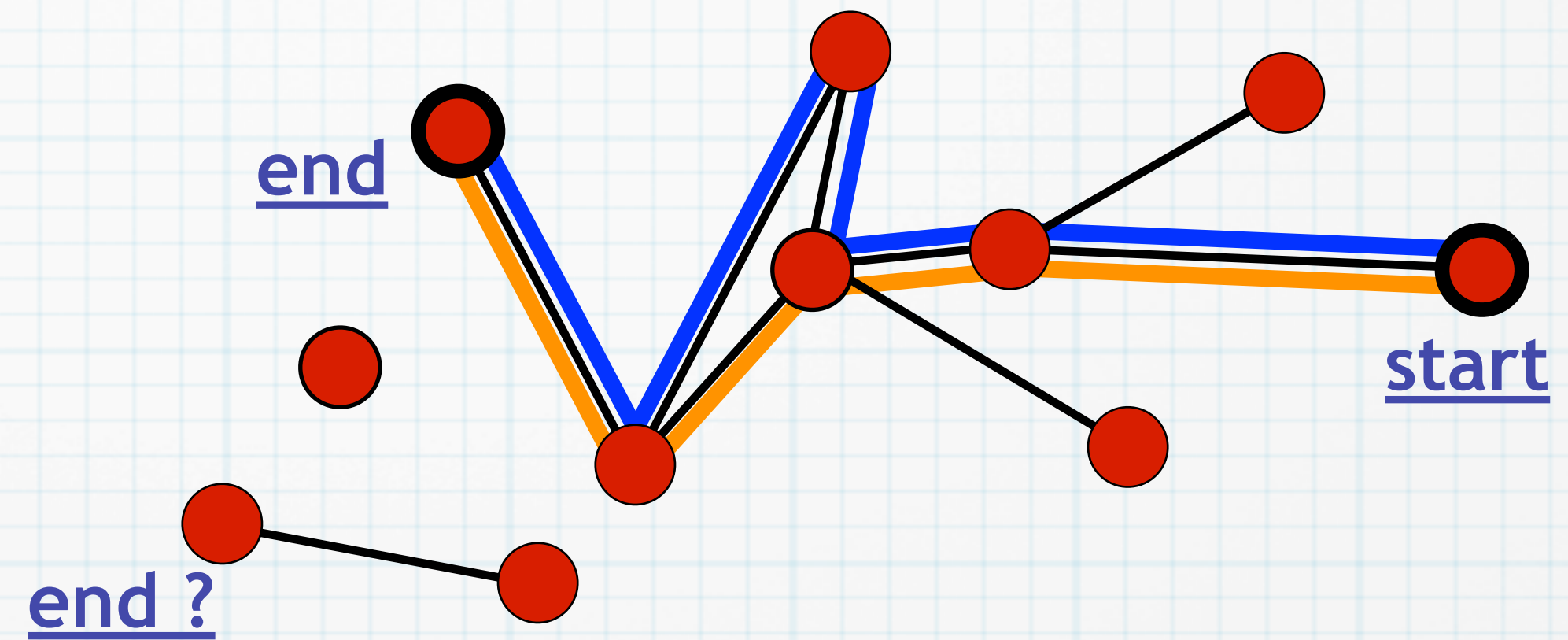
* shortest path



Distance

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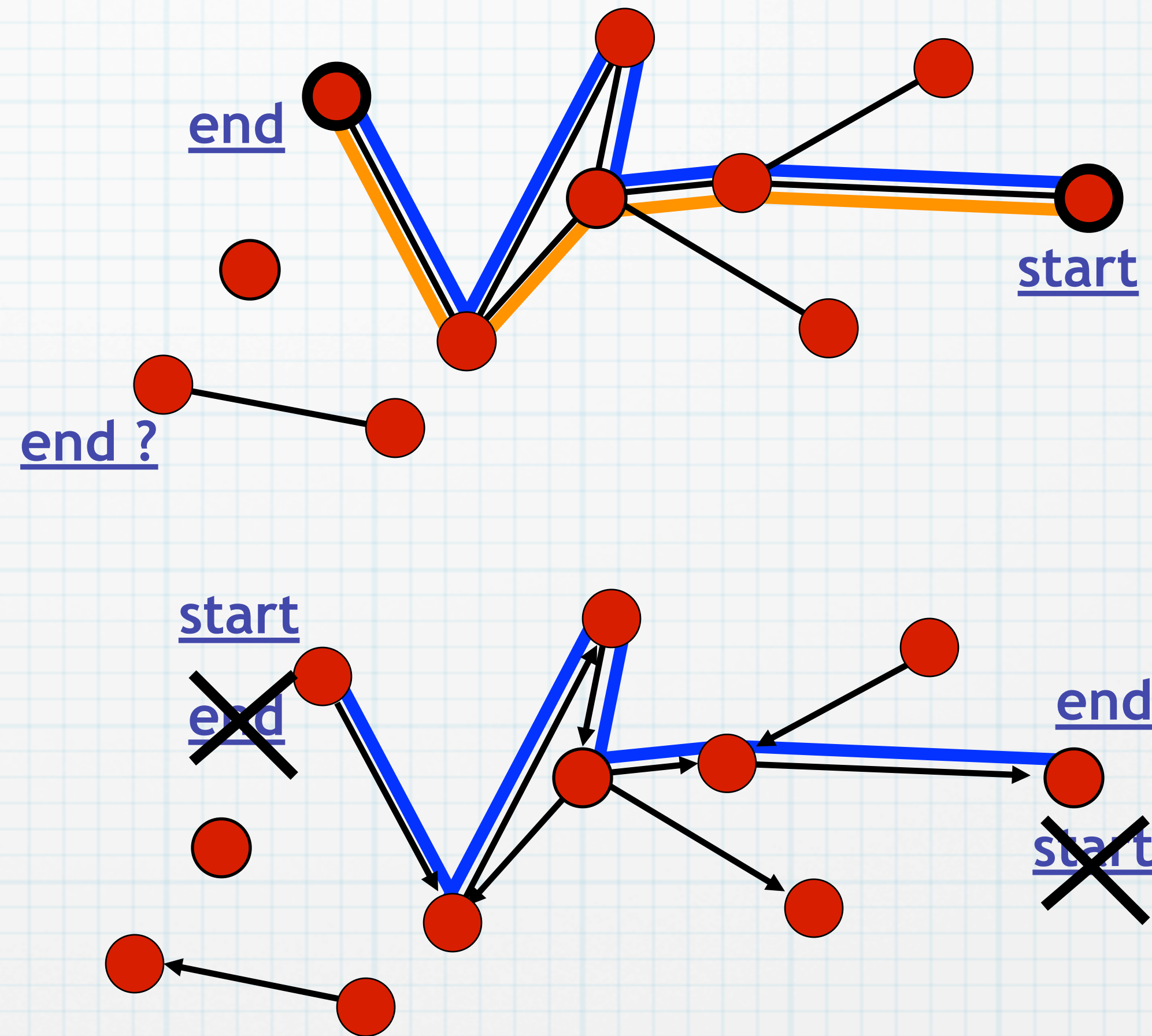
* shortest path



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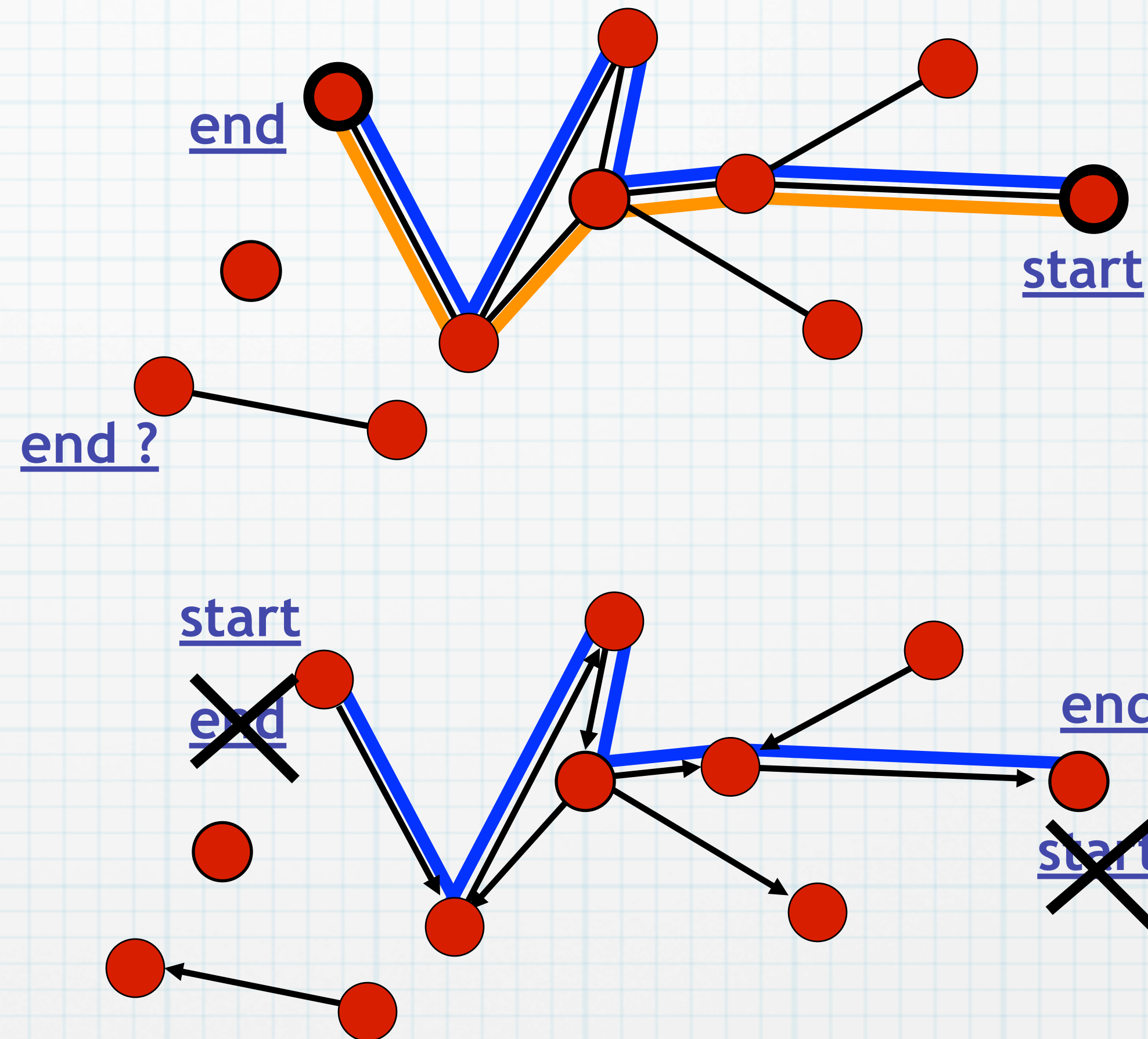
* path

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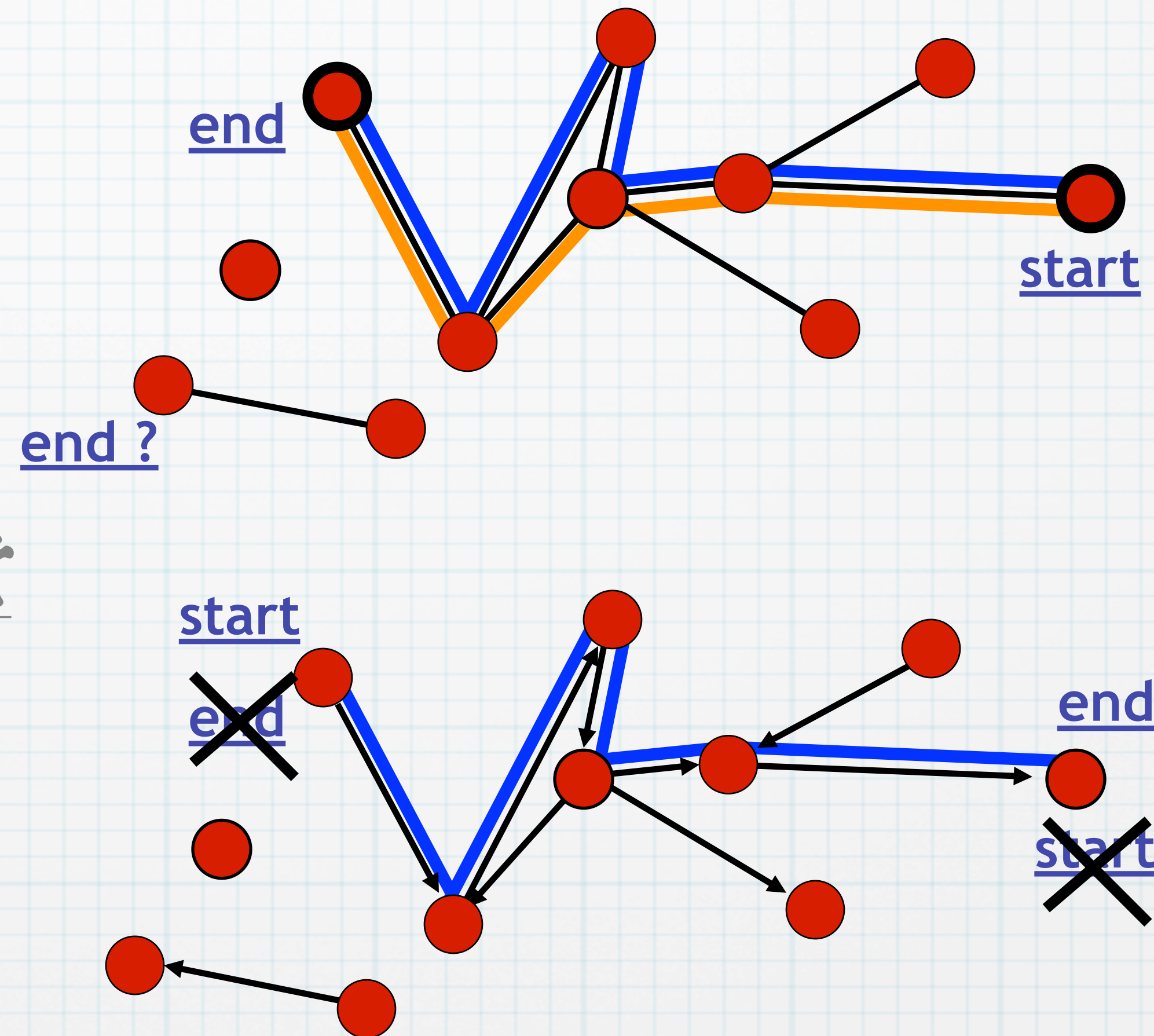
Distance

- * path
- * shortest path
- * tree: # paths?



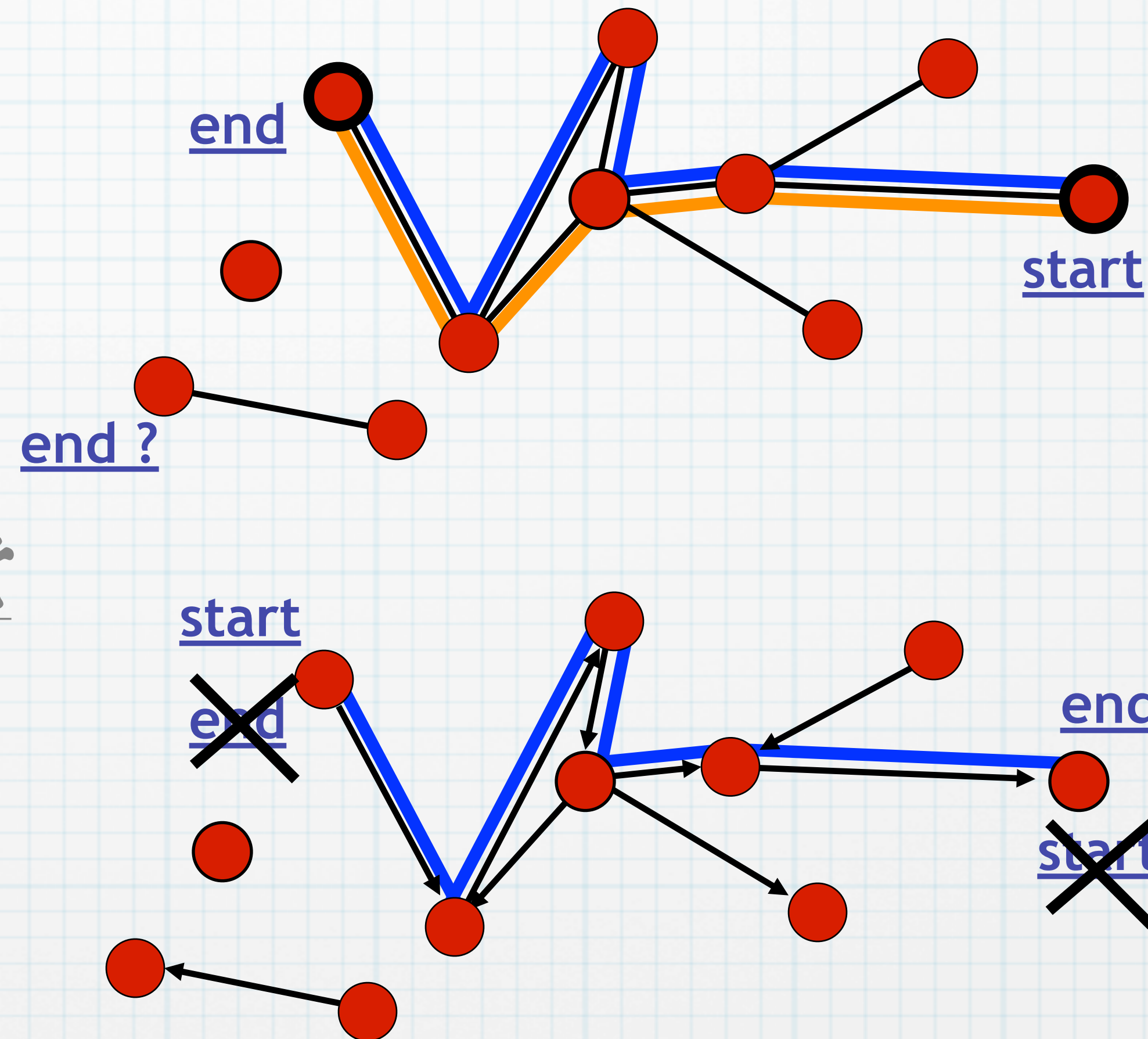
Distance

- * path
- * shortest path
- * tree: # paths?
- * Diameter: longest shortest path length

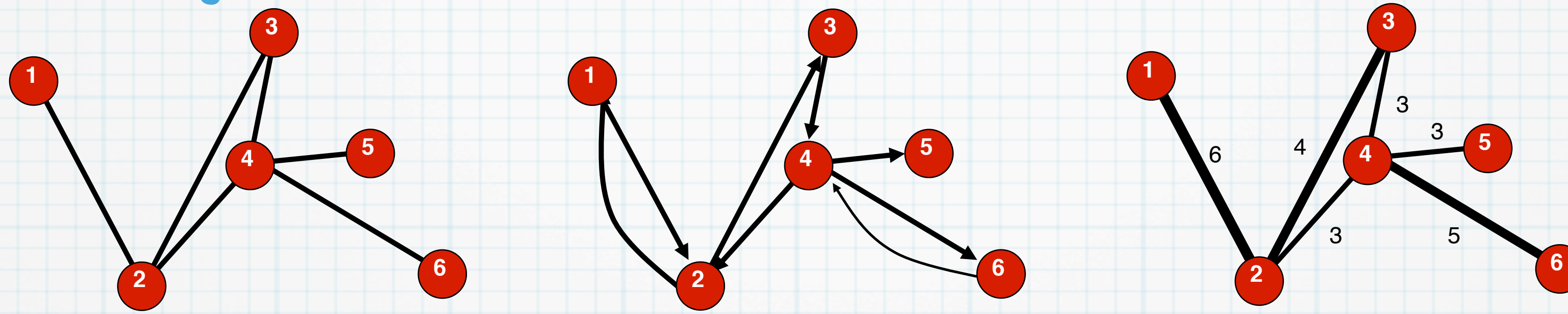


Distance

- * path
- * shortest path
- * tree: # paths?
- * Diameter: longest shortest path length
- * APL: average shortest path length



Python and NetworkX



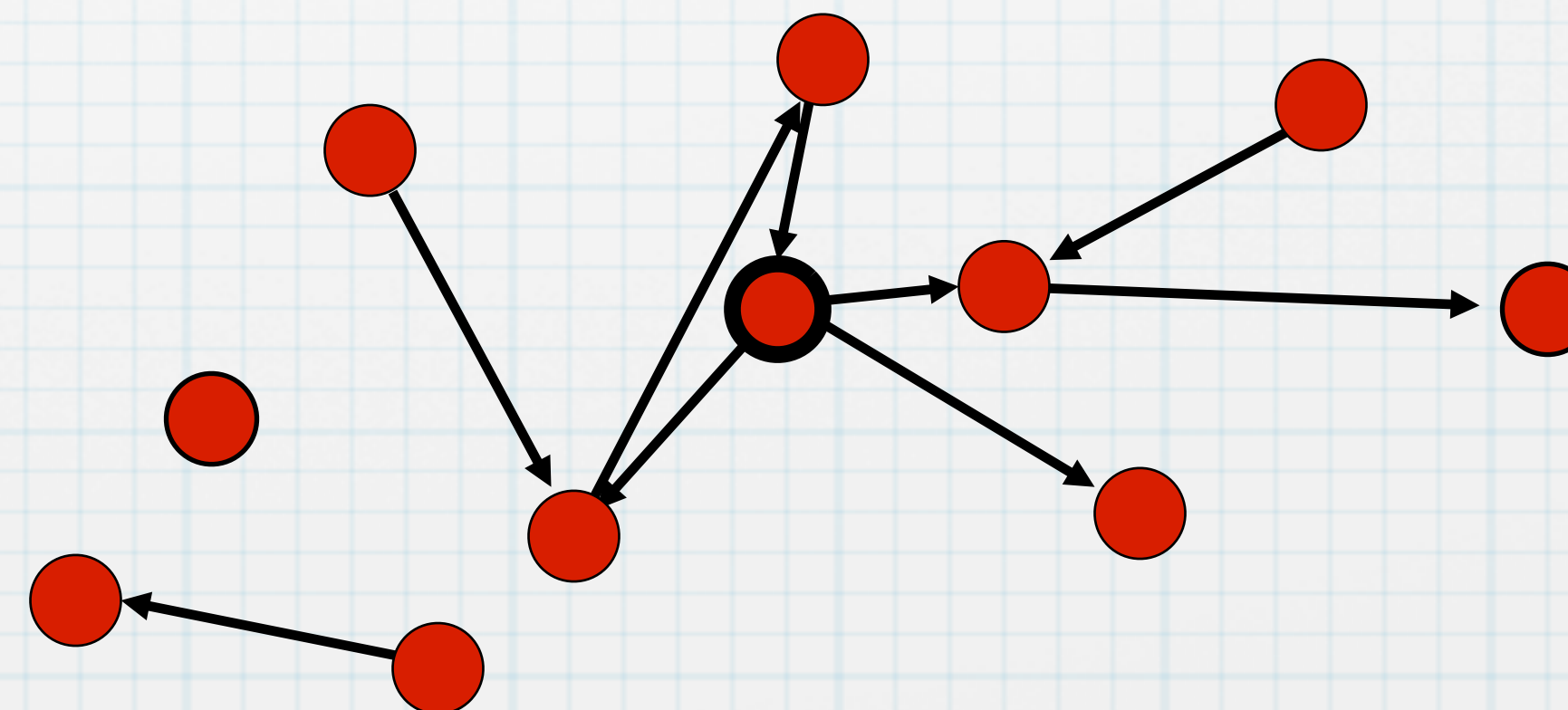
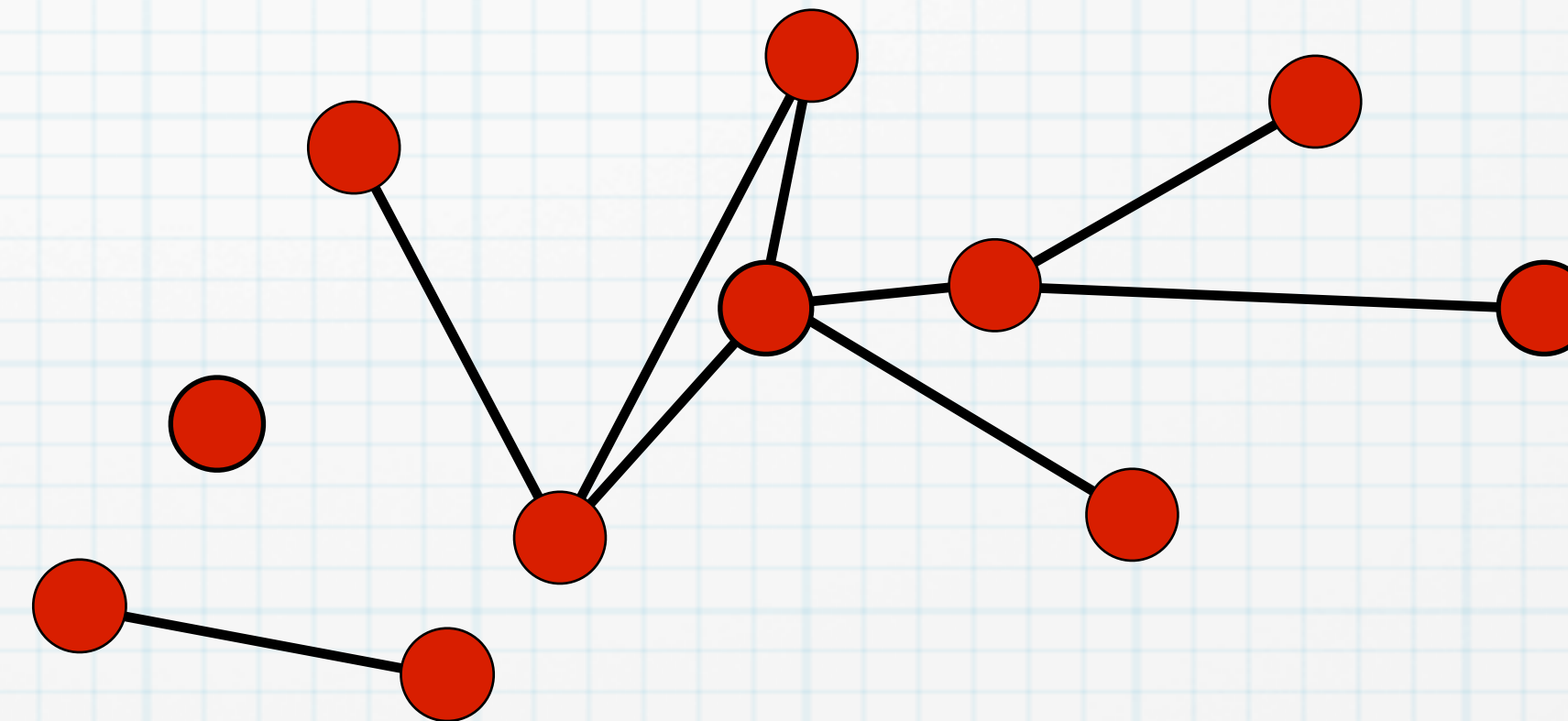
```
nx.has_path(G,1,6)
nx.shortest_path(G,1,6)
nx.shortest_path_length(G,1,6)
nx.shortest_path(G,1)
nx.shortest_path(G)
nx.average_shortest_path_length(G)

nx.shortest_path(D,1,6)
nx.has_path(D,1,5)
nx.has_path(D,5,1)

nx.shortest_path_length(W,1,6)
nx.shortest_path_length(W,1,6,'weight')
```

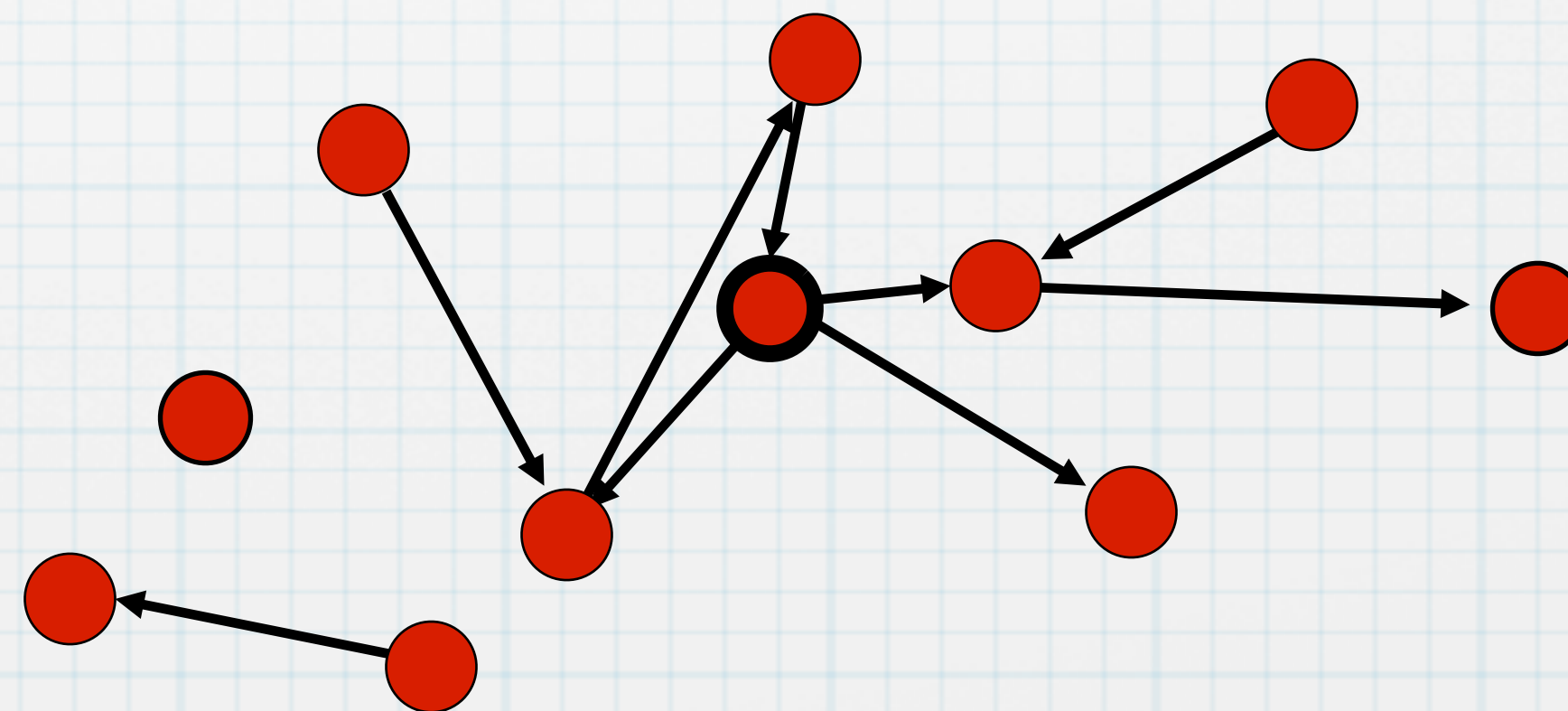
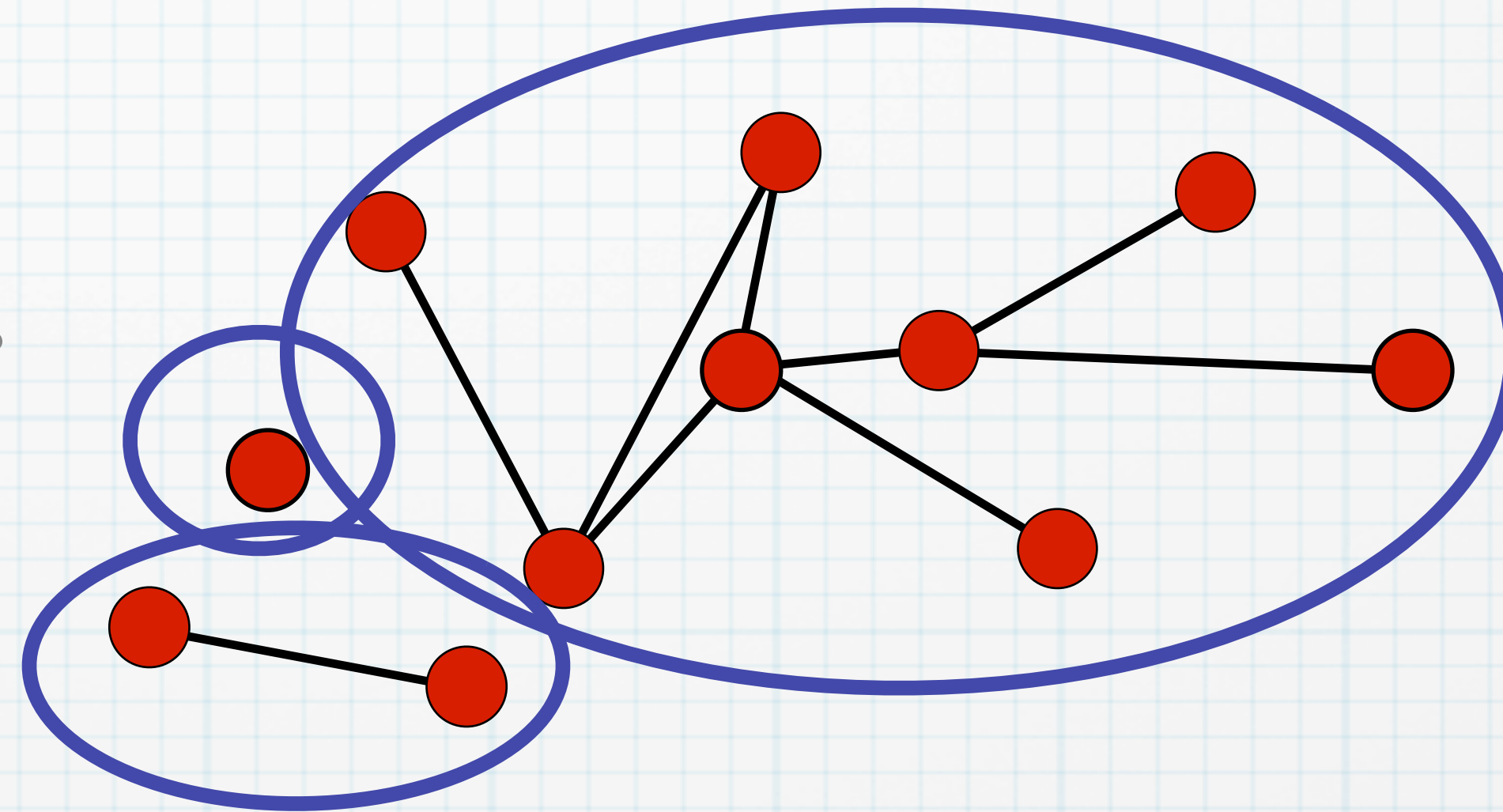

More definitions

- * connectedness
- * giant component
- * isolates
- * components
- * in- and out-
- * weak, strong



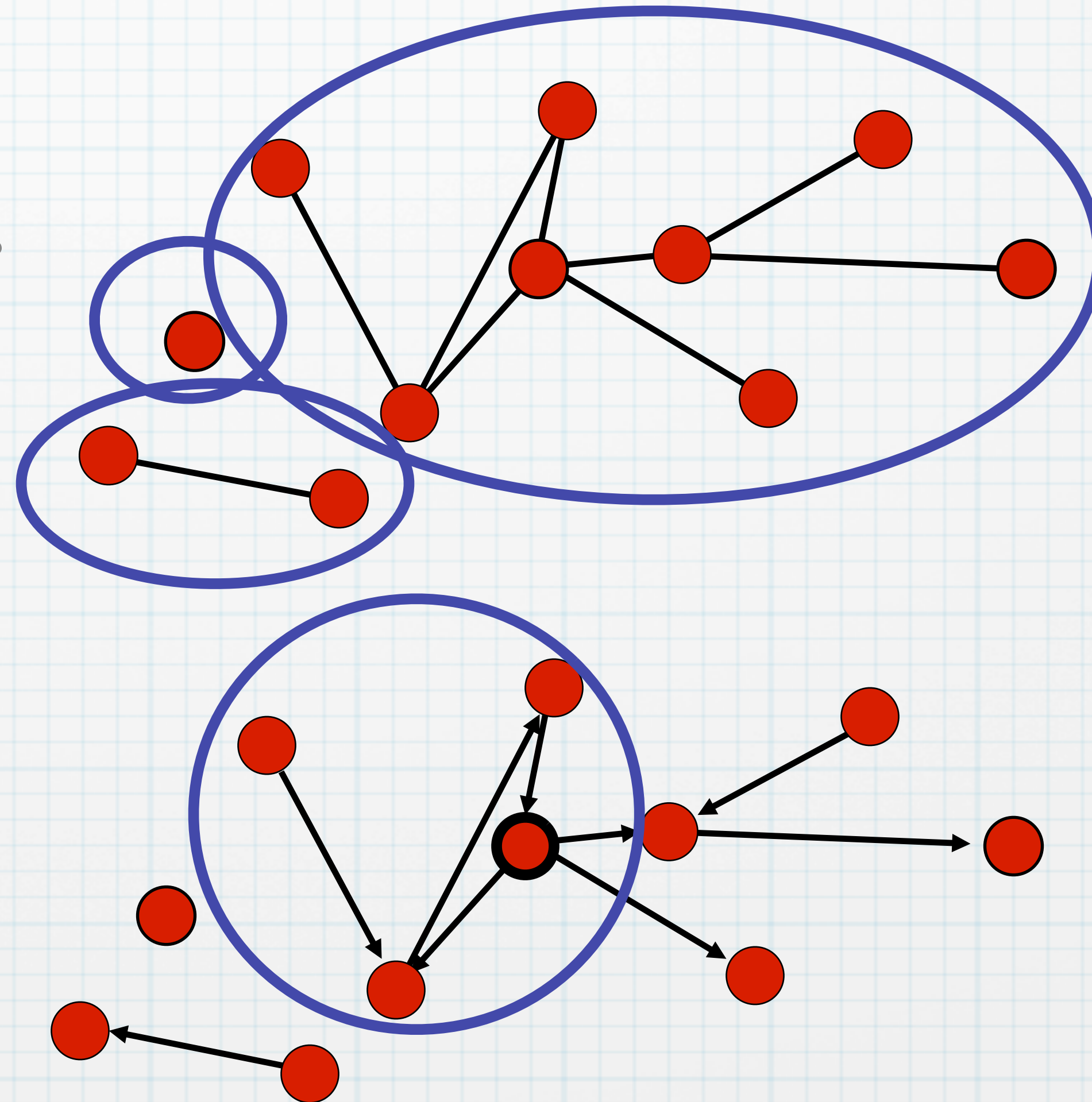
More definitions

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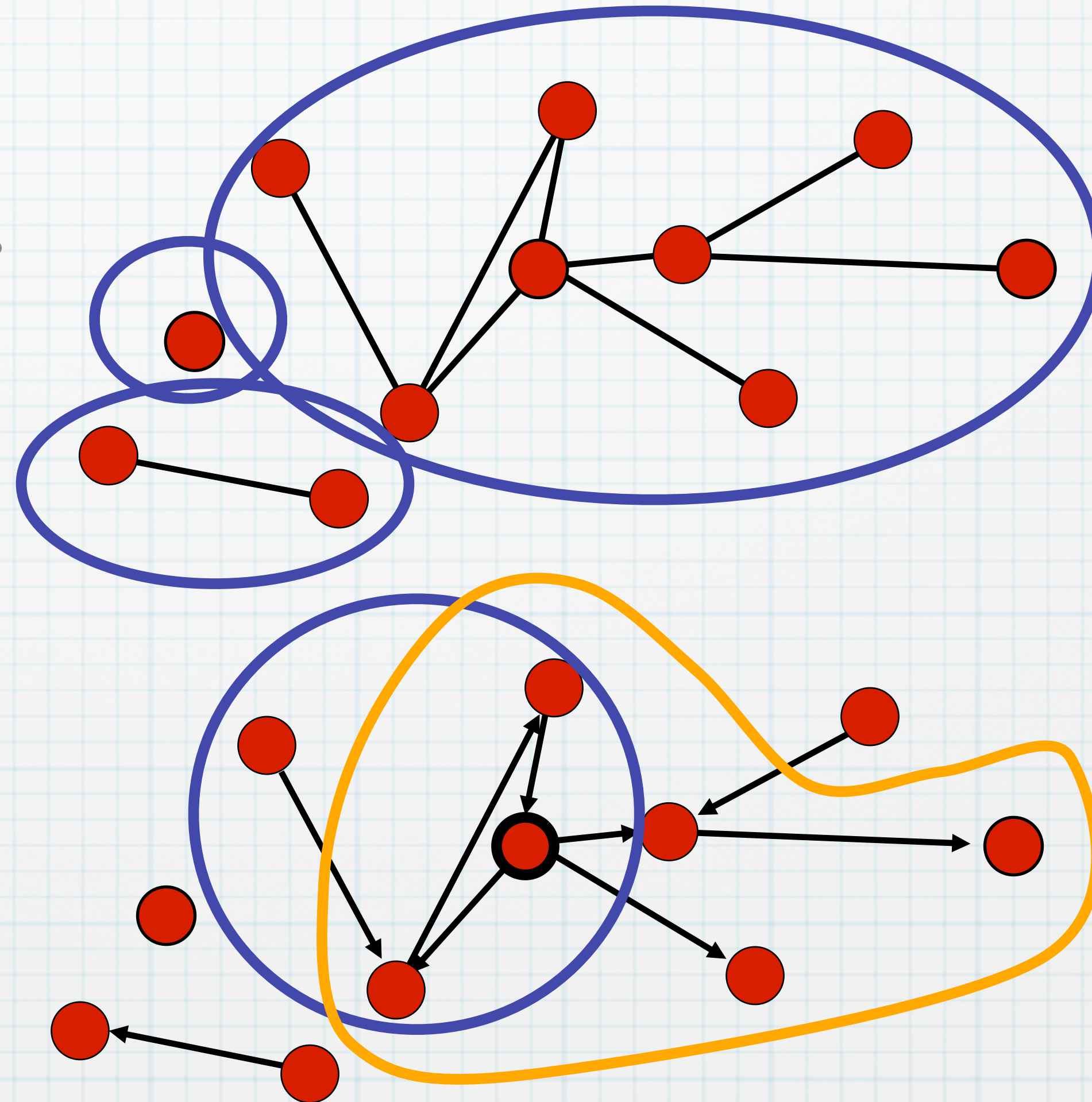
More definitions

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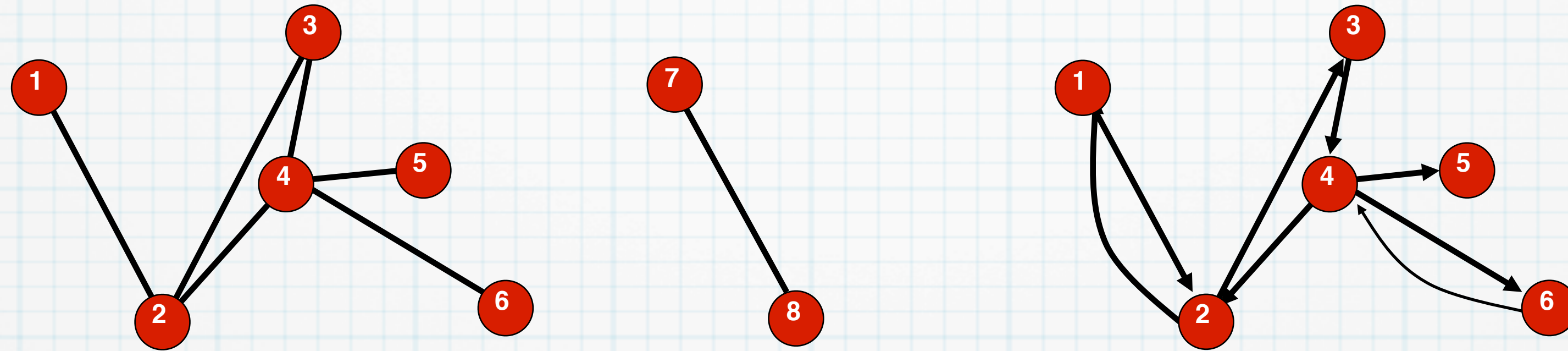


More definitions

- * connectedness
- * giant component
- * isolates
- * components
- * in- and out-
- * weak, strong



Python and NetworkX



```
nx.is_connected(G)
sorted(nx.connected_components(G))
G.add_edge(7,8)
nx.is_connected(G)
sorted(nx.connected_components(G))

nx.is_weakly_connected(D)
nx.is_strongly_connected(D)
sorted(nx.strongly_connected_components(D))
```

Basic measures

- * Networks:

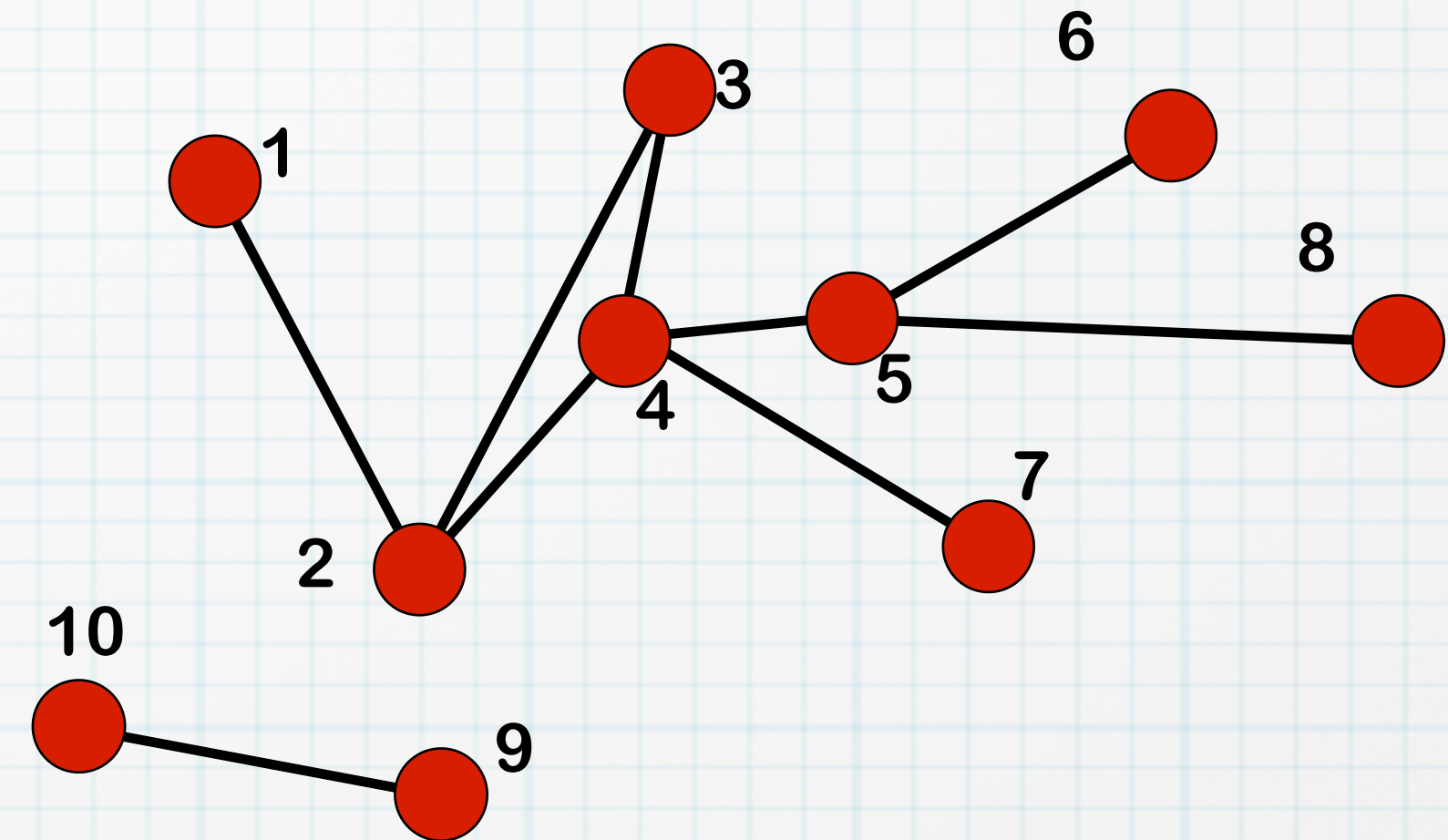
- * size N : # vertices

- * E : # edges

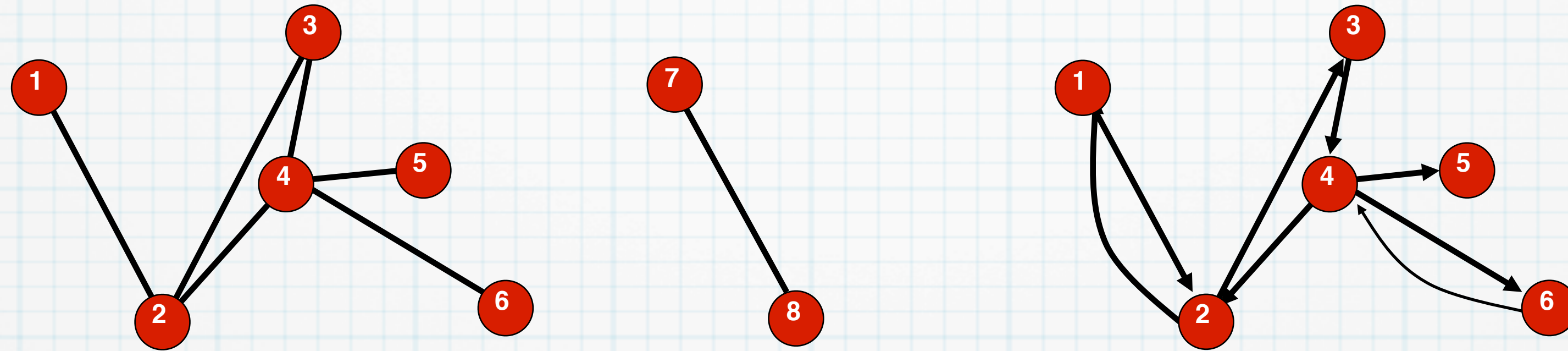
- * $E_{\max} = N(N-1)/2 \rightarrow$ complete graph

- * density: E/E_{\max}

- * if density $\ll 1$ (low) \rightarrow sparse graph



Python and NetworkX



```
G.number_of_nodes()  
G.number_of_edges()  
nx.density(G)  
nx.density(D)
```

Mathematical representation

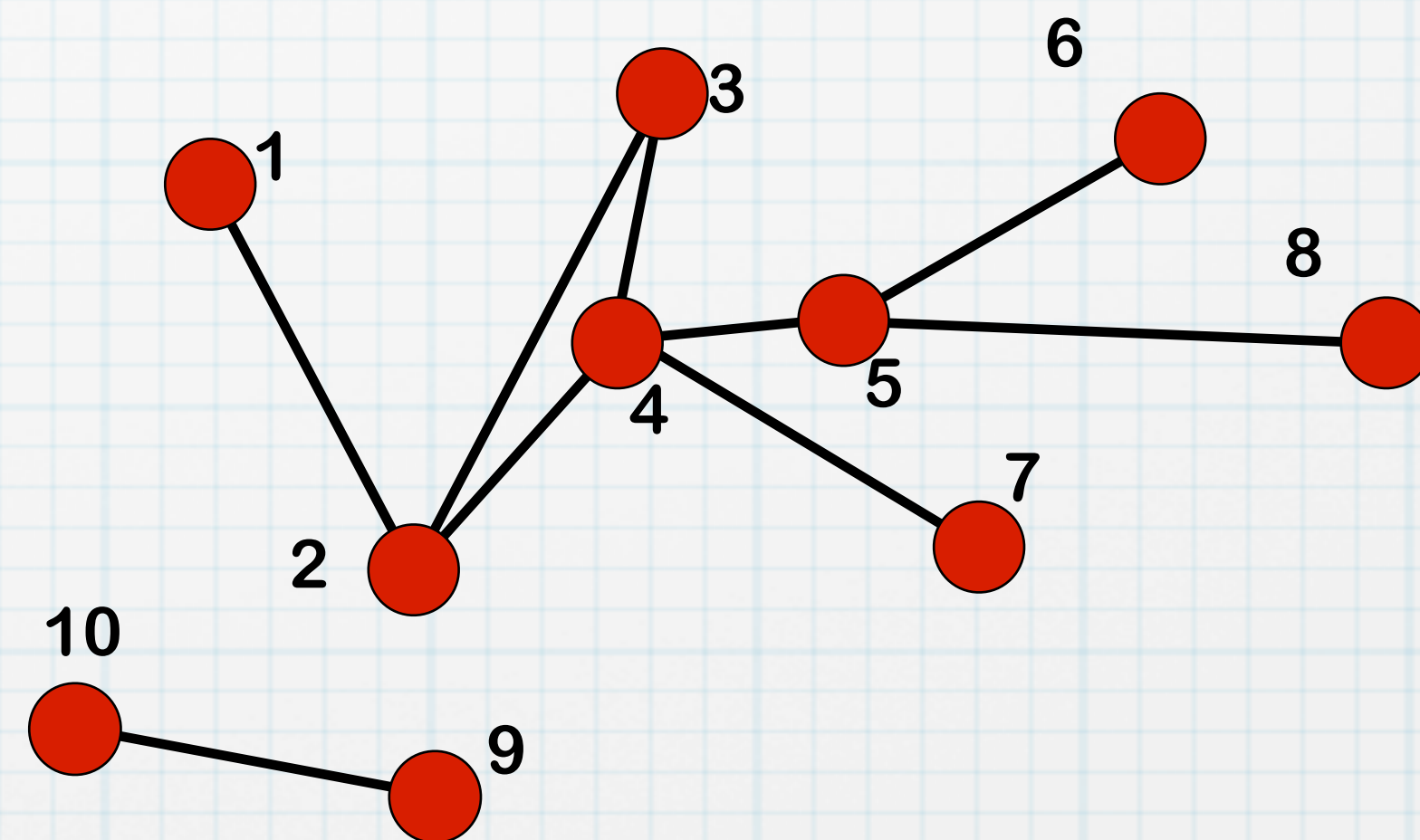
- * Adjacency matrix $A=(A_{ij})$

- * $A_{ij} = 1$ if i linked to j

- * $A_{ij} = 0$ otherwise

- * $A_{47} = ?$ $A_{74} = ?$

- * $A_{13} = ?$

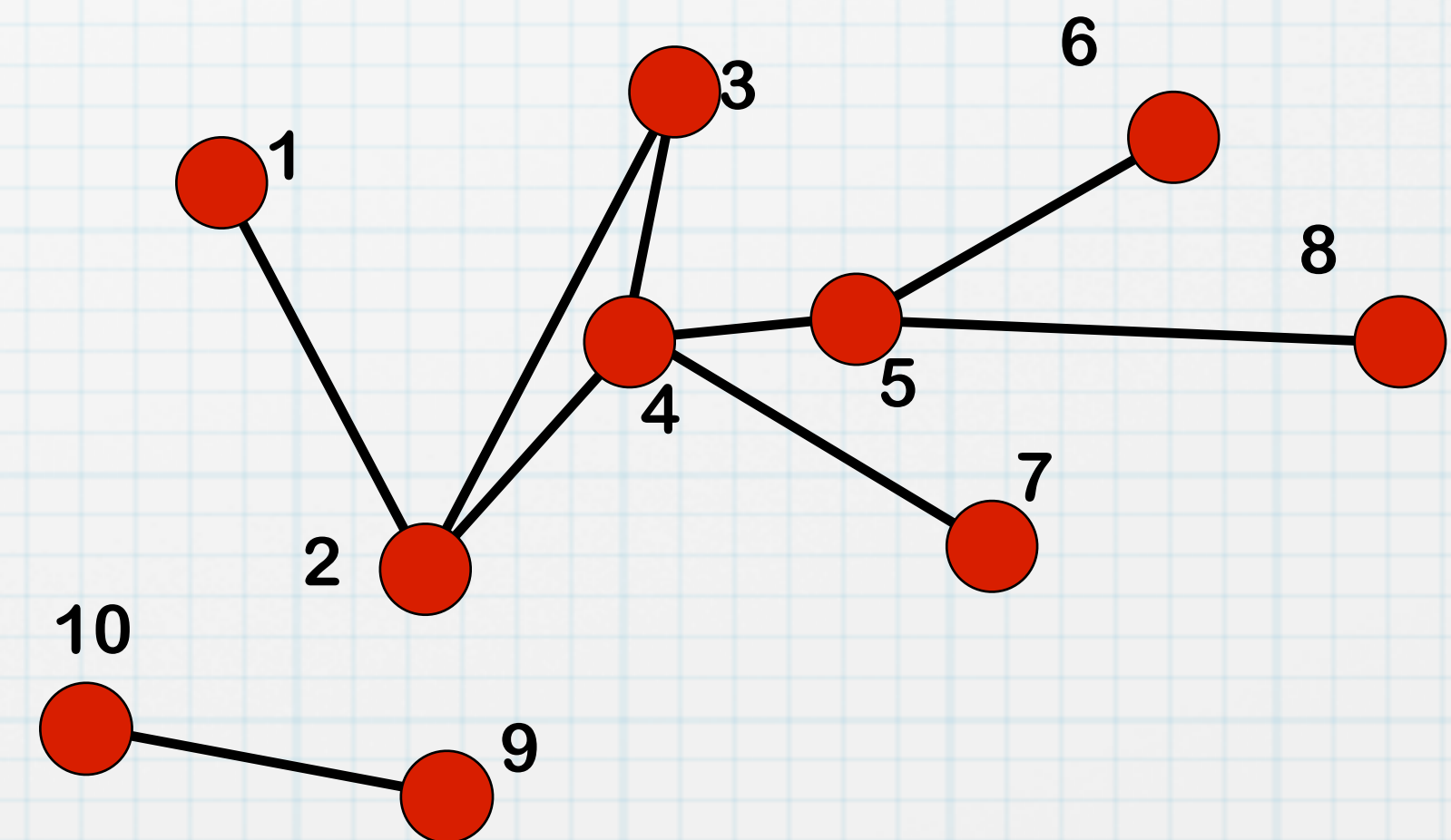


- * a.k.a. sociomatrix, connectivity matrix

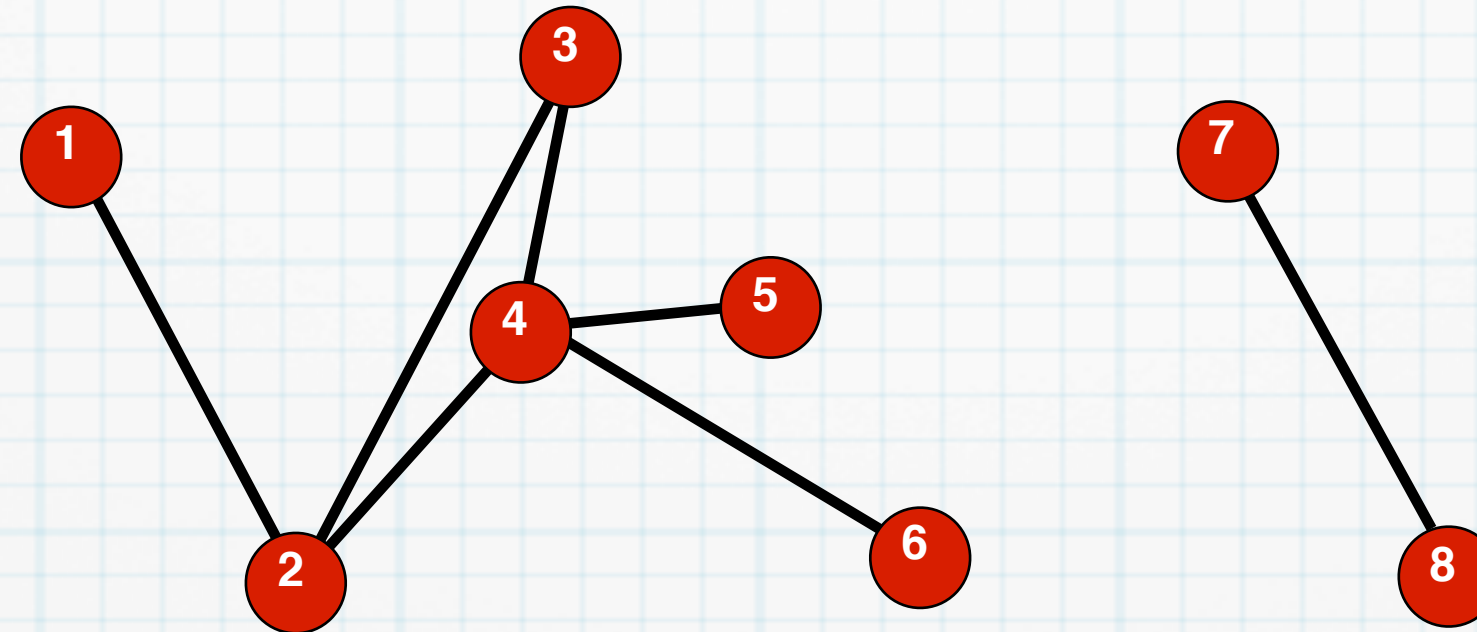
Mathematical representation

- * no self-loops: $A_{ii} = 0$
- * undirected network: $A_{ij} = A_{ji}$
- * \rightarrow symmetric matrix
- * no multiple edges: $A_{ij} = 0$ or 1

	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



Python and NetworkX

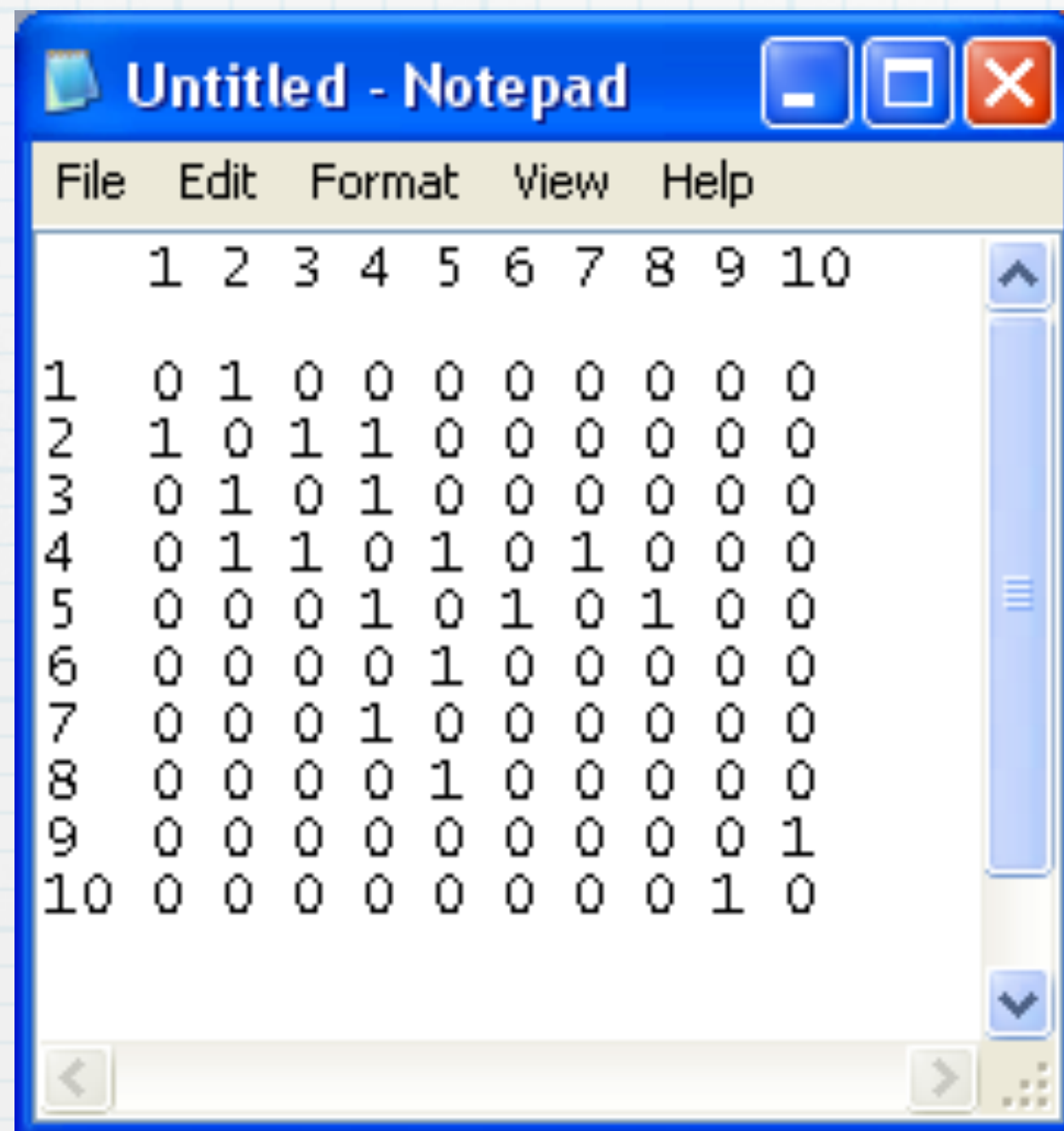


```
nx.adjacency_matrix(G)
print(nx.adjacency_matrix(G))
G.edge[3][4]
G.edge[3][4]['color']='blue'
G.edge[3][4]
G.edge[4]
```


Local measures

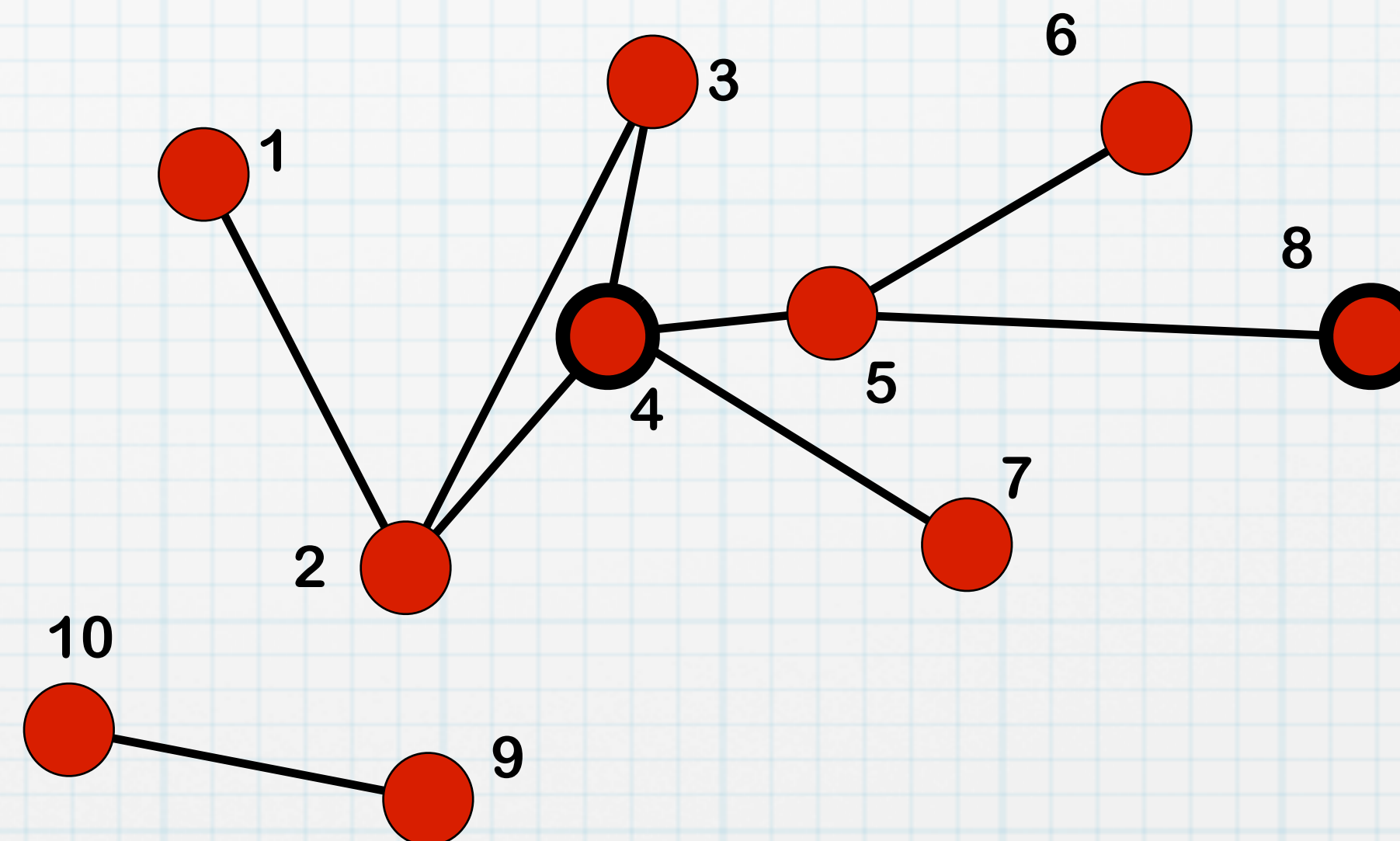
Degree: # links departing from the node

$$k_i = A_{i1} + A_{i2} + \dots + A_{iN} = \sum_j A_{ij}$$



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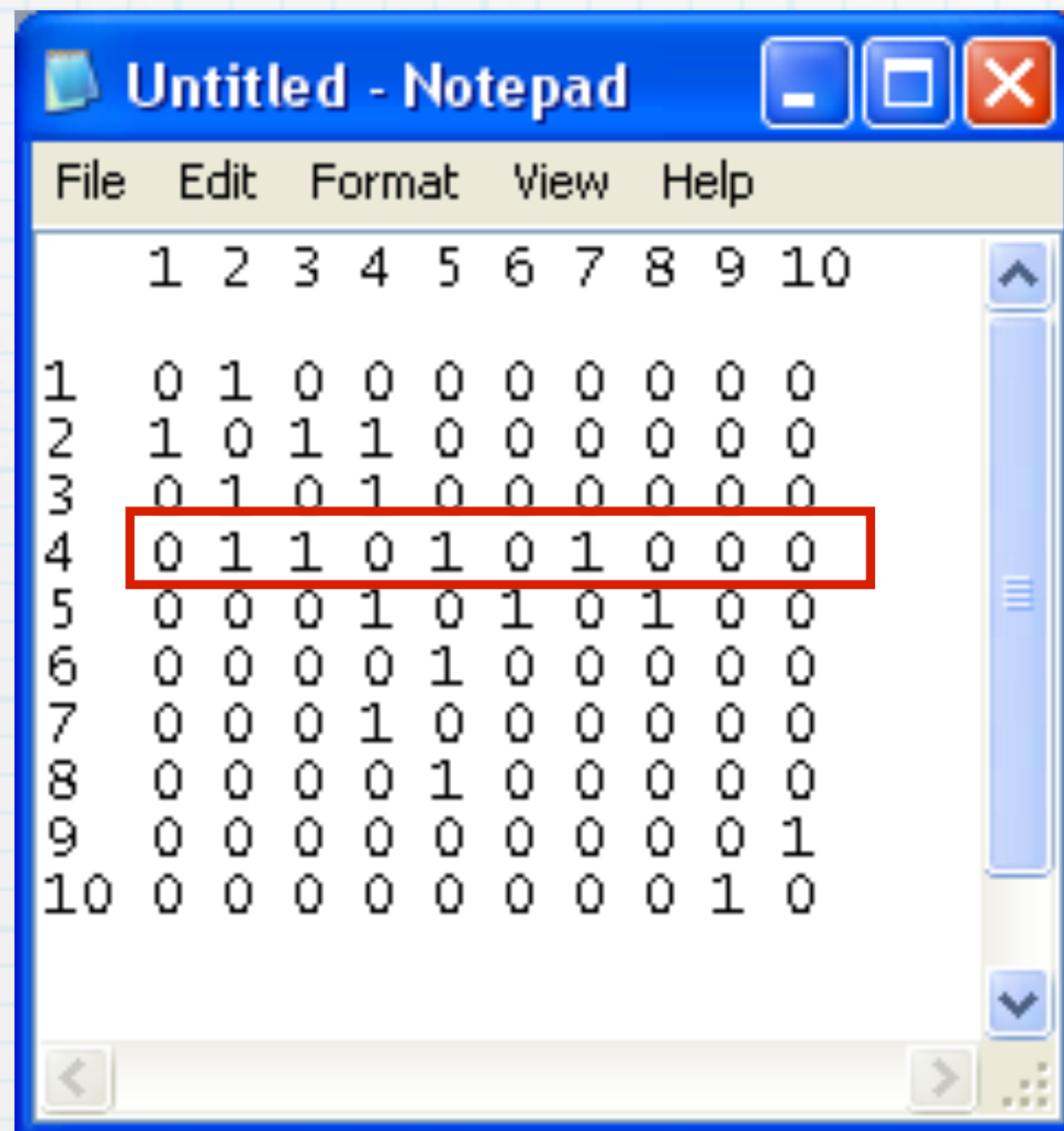
	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



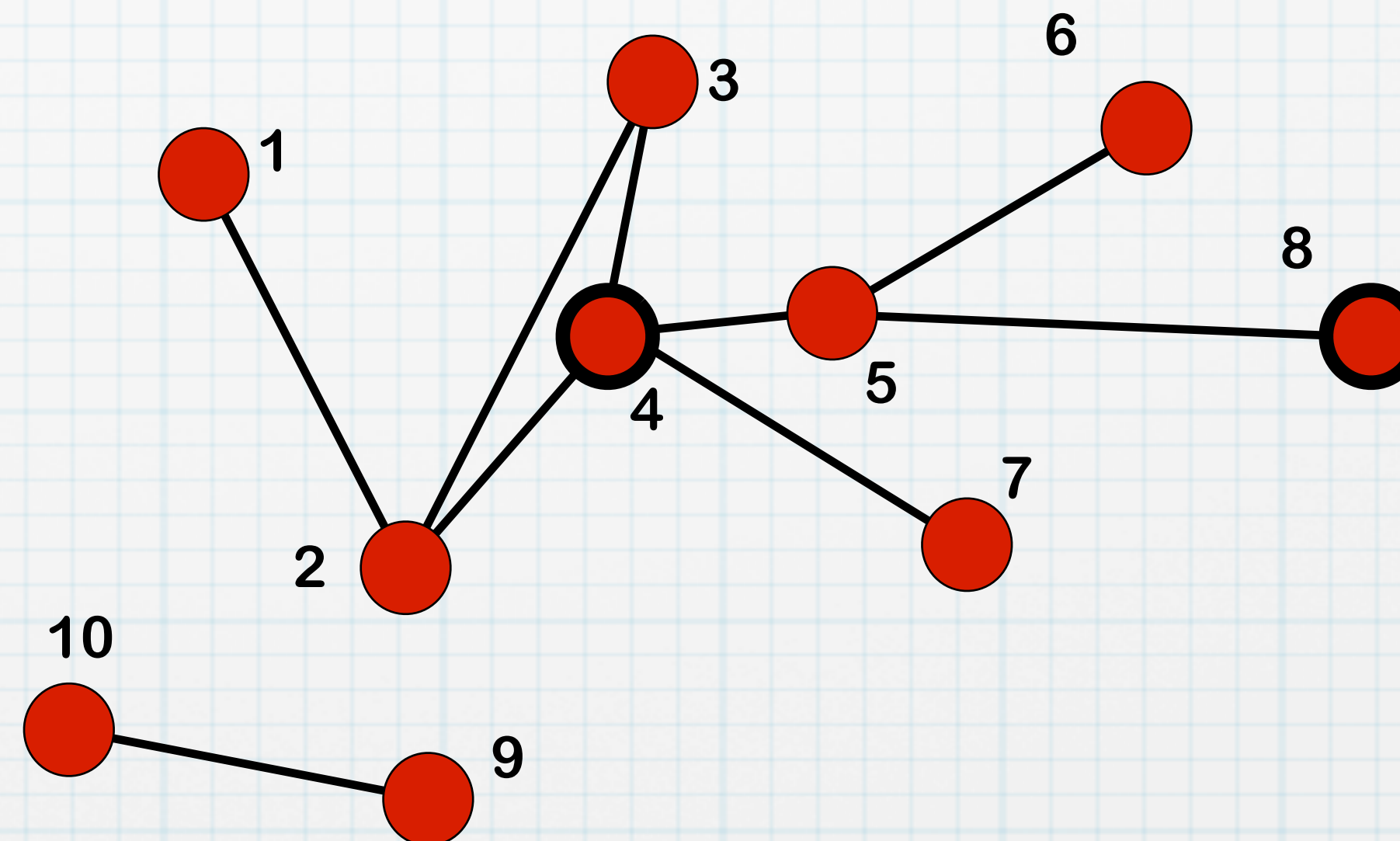
Local measures

Degree: # links departing from the node

$$k_i = A_{i1} + A_{i2} + \dots + A_{iN} = \sum_j A_{ij}$$



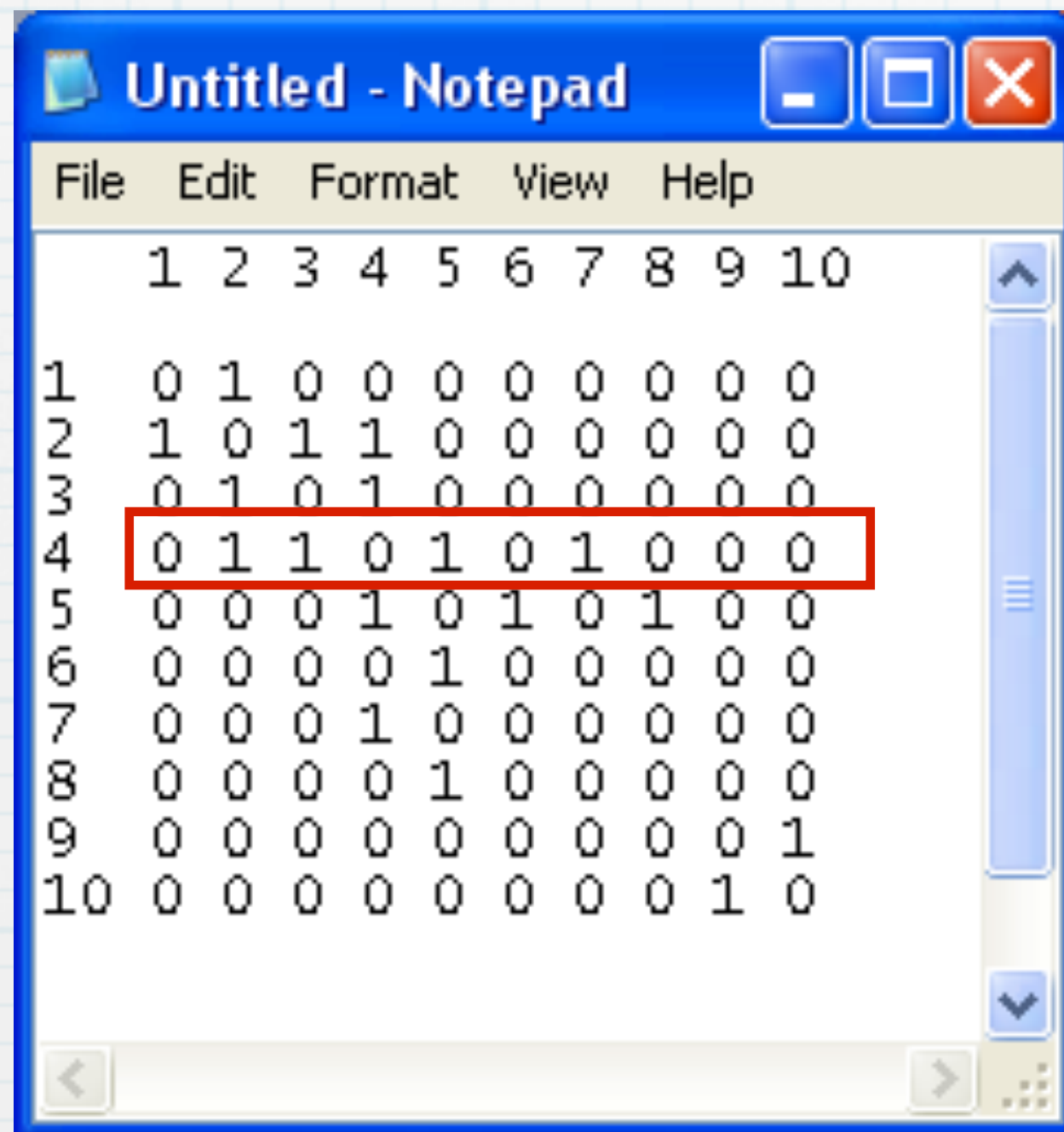
	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



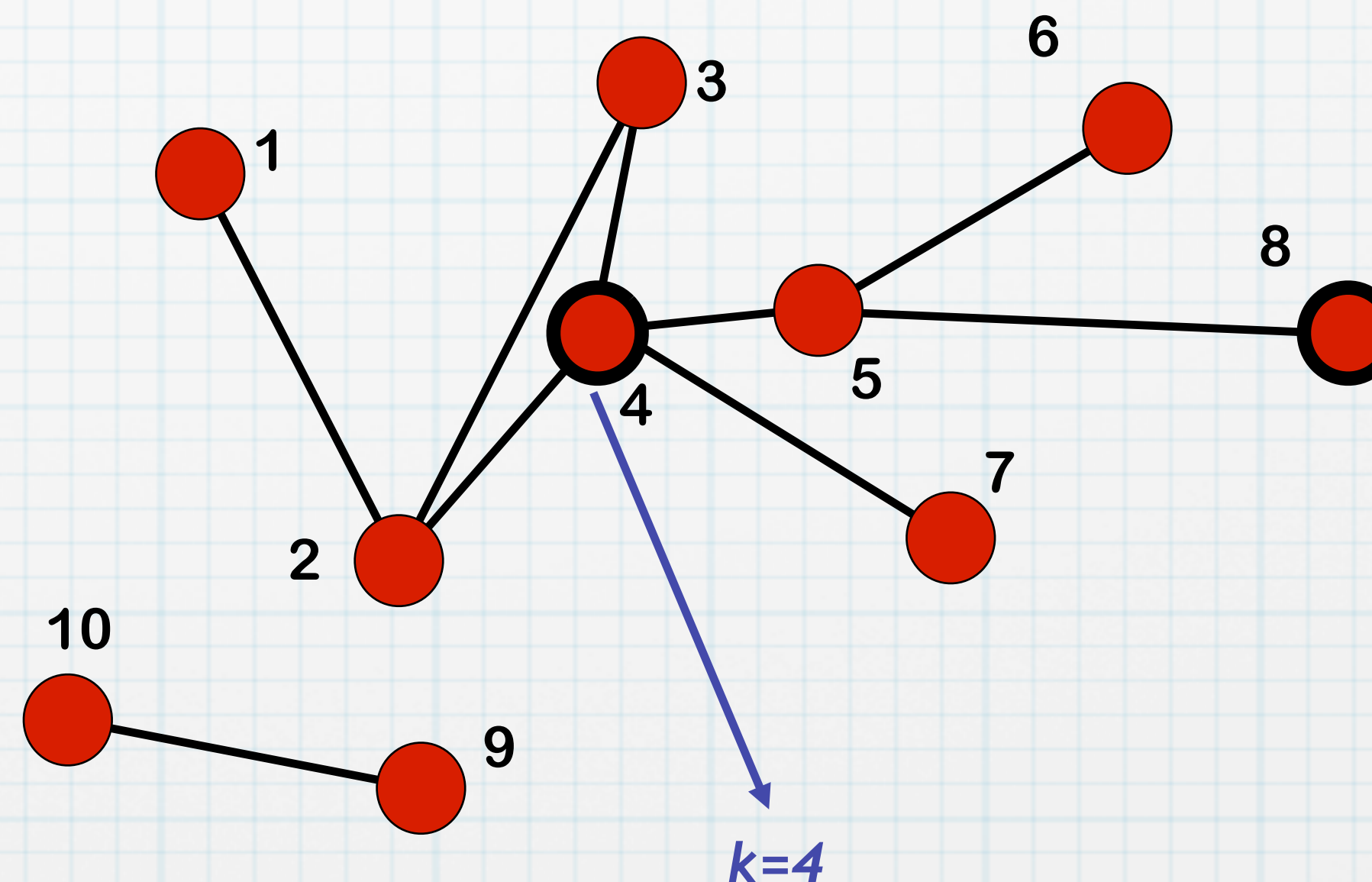
Local measures

Degree: # links departing from the node

$$k_i = A_{i1} + A_{i2} + \dots + A_{iN} = \sum_j A_{ij}$$



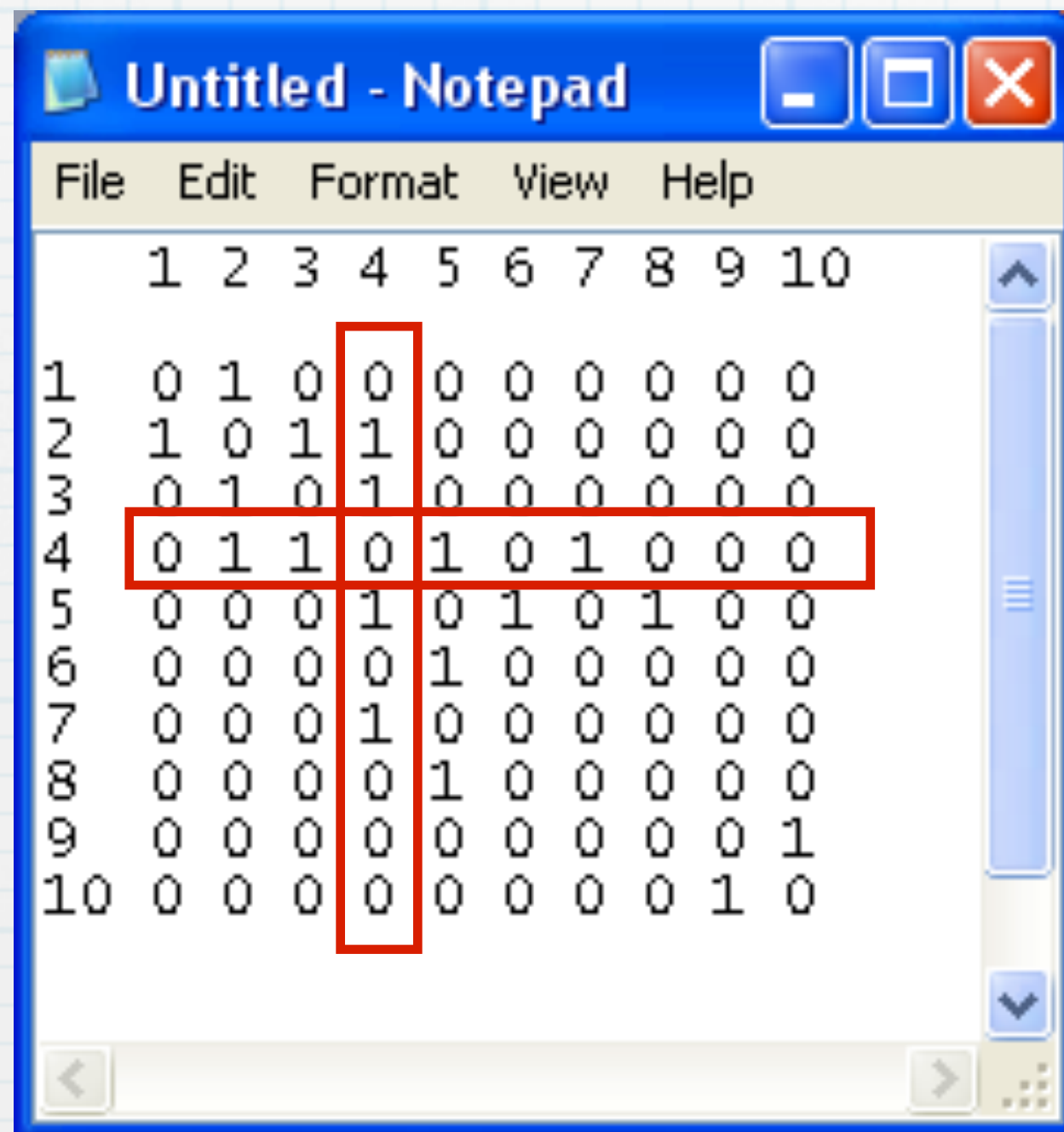
	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



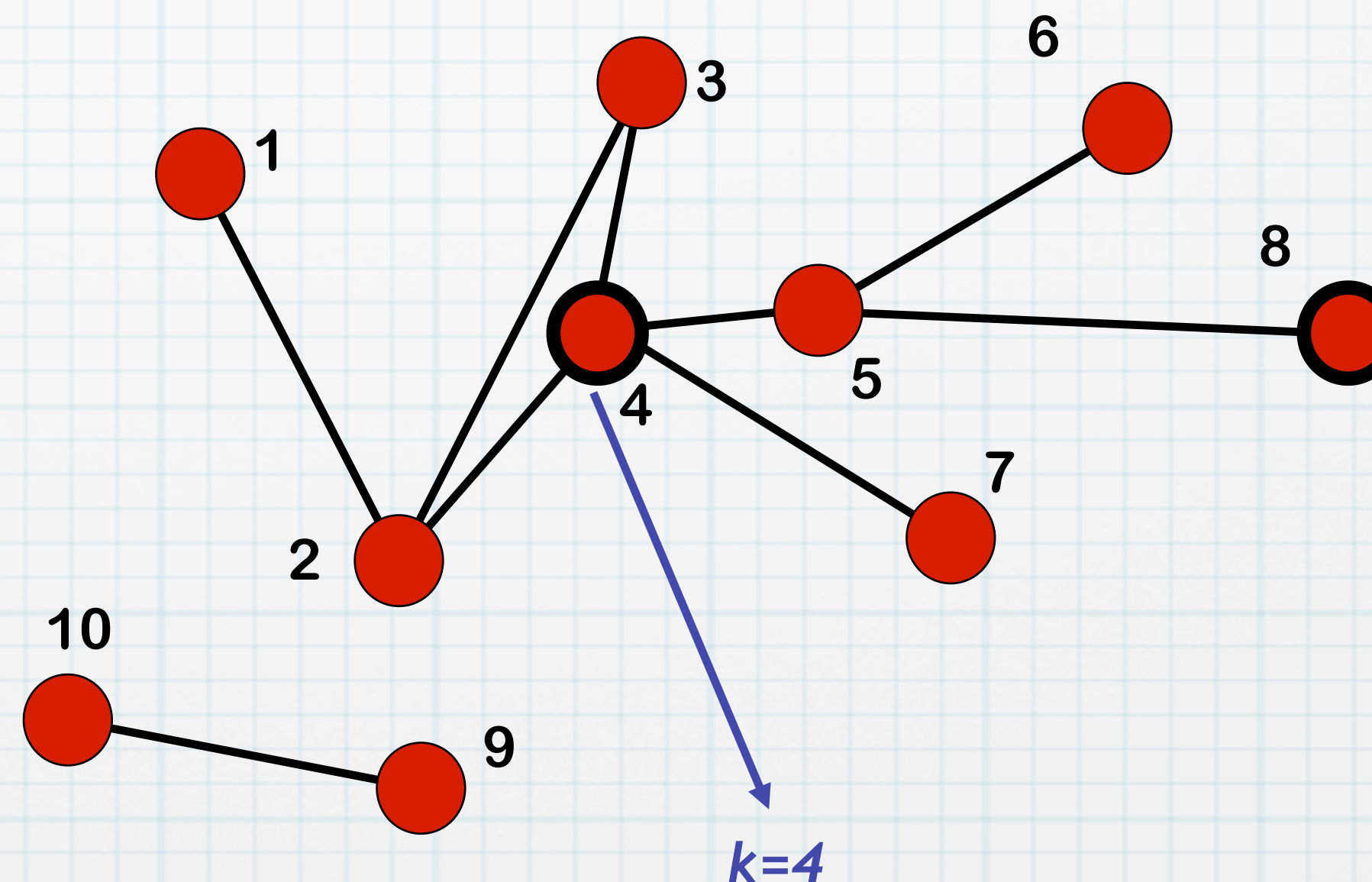
Local measures

Degree: # links departing from the node

$$k_i = A_{i1} + A_{i2} + \dots + A_{iN} = \sum_j A_{ij}$$



	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



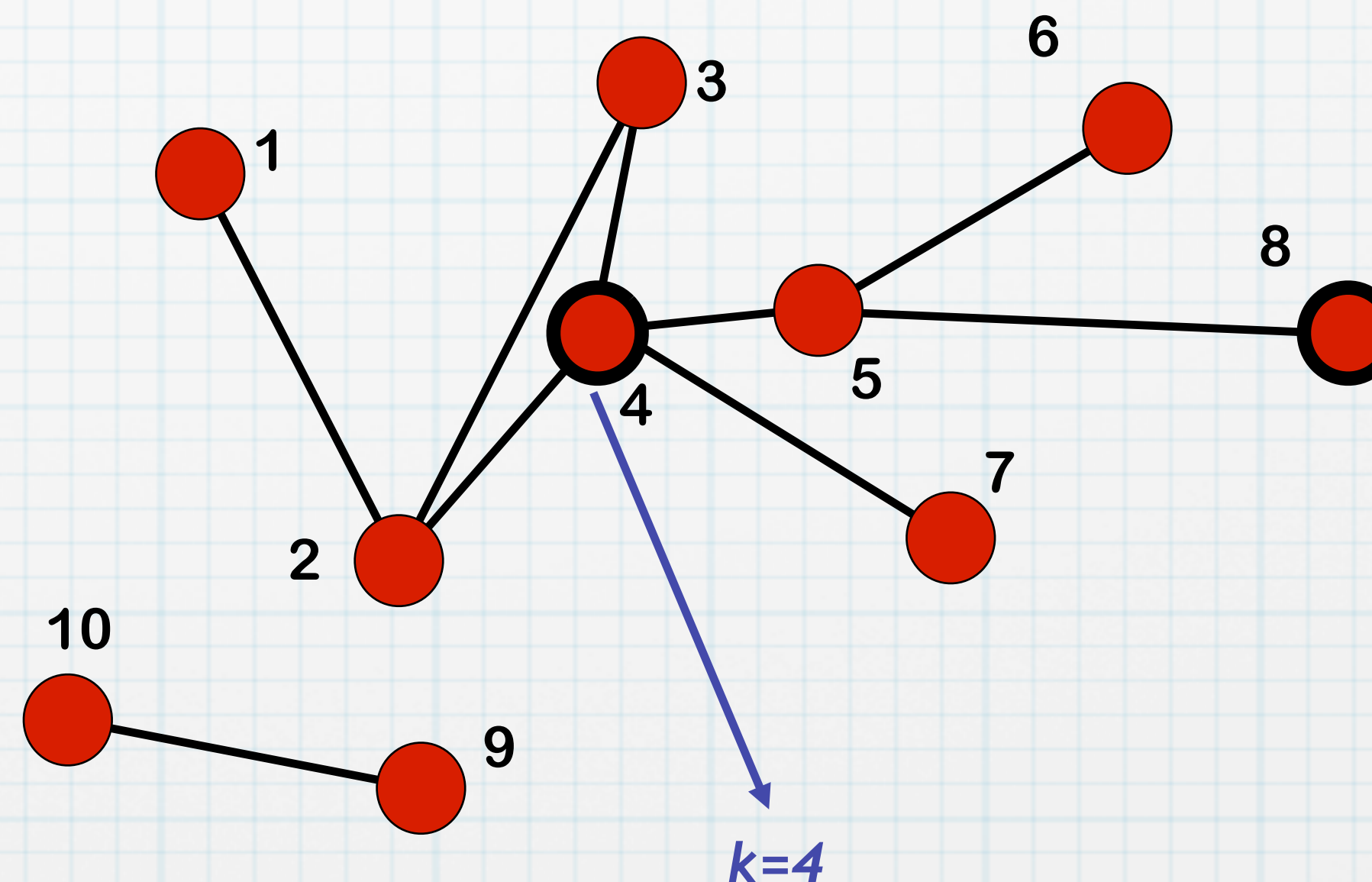
Local measures

Degree: # links departing from the node

$$k_i = A_{i1} + A_{i2} + \dots + A_{iN} = \sum_j A_{ij}$$

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	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



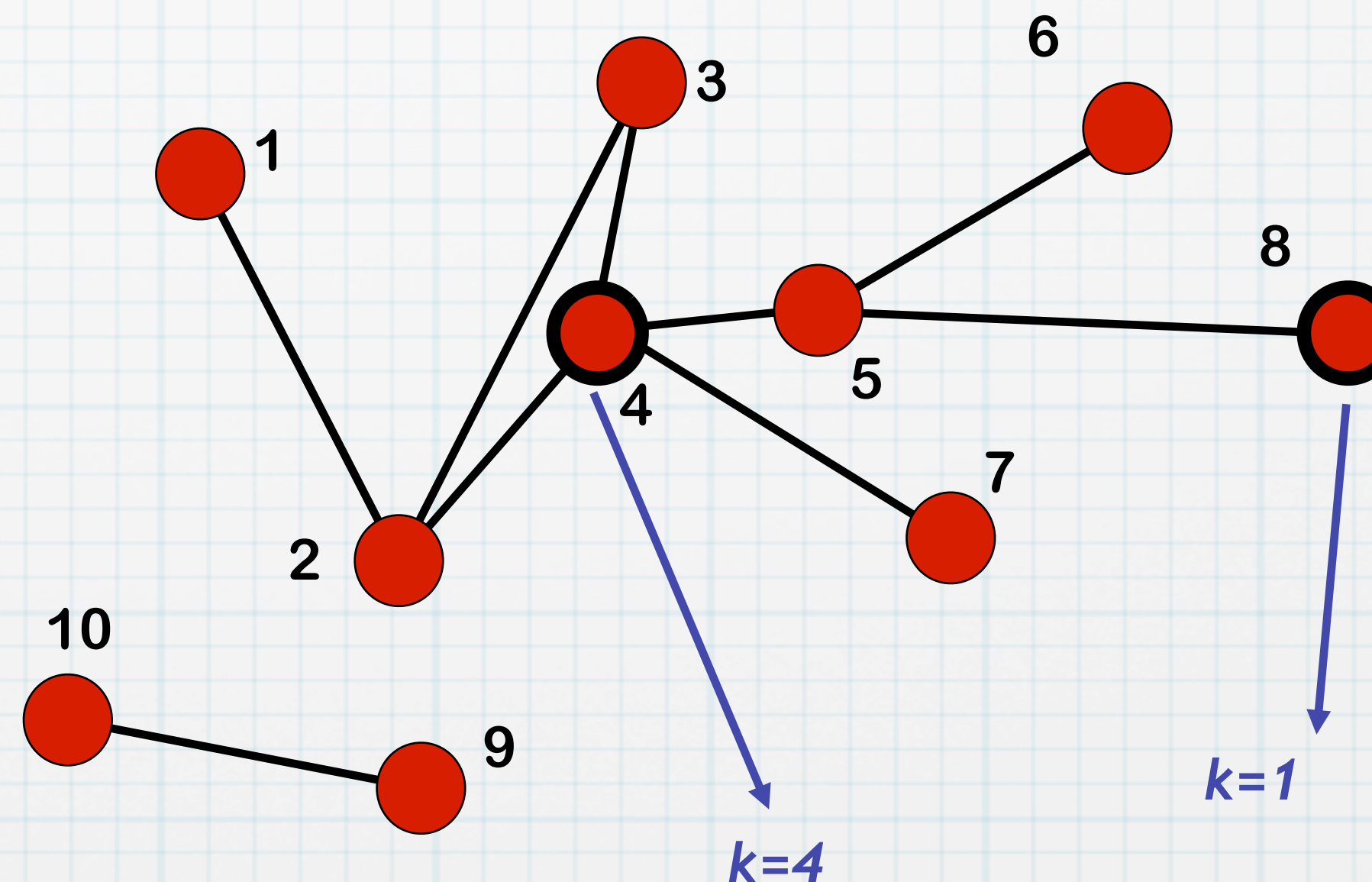
Local measures

Degree: # links departing from the node

$$k_i = A_{i1} + A_{i2} + \dots + A_{iN} = \sum_j A_{ij}$$

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	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	1	0	0	0	0	0	0
3	0	1	0	1	0	0	0	0	0	0
4	0	1	1	0	1	0	1	0	0	0
5	0	0	0	1	0	1	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	1	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	1
10	0	0	0	0	0	0	0	0	1	0



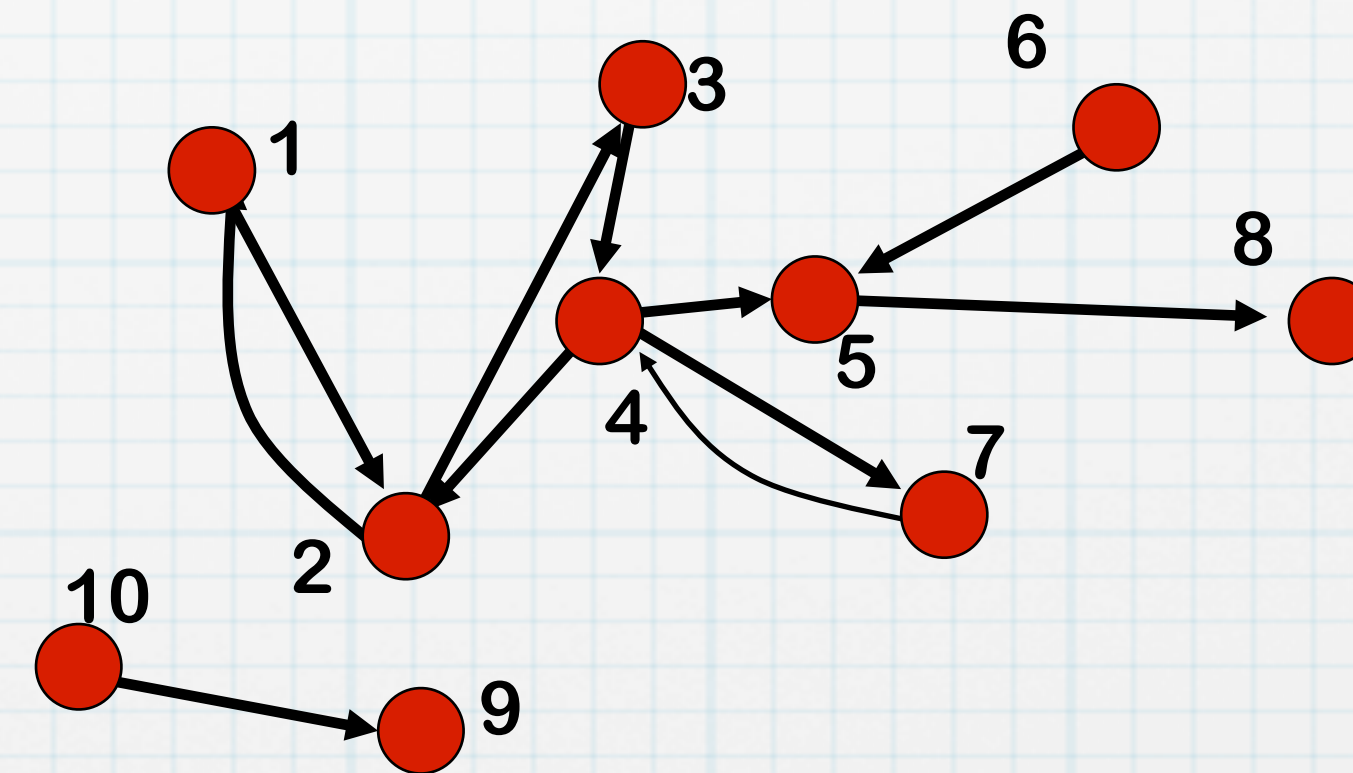
- Attendance
- Schedule update
- Review

Mathematical representation

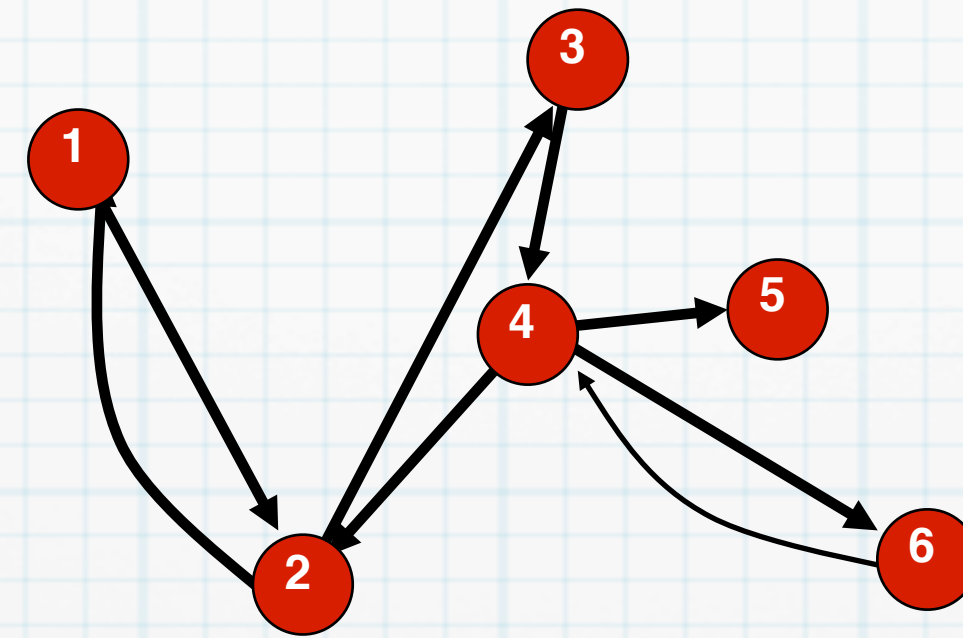
* directed networks:
not symmetric !!!

* $i \rightarrow j \quad A_{ij} = 1$

	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	0	0	0	0	0	0	0
3	0	0	0	1	0	0	0	0	0	0
4	0	1	0	0	1	0	1	0	0	0
5	0	0	0	0	0	0	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0
10	0	0	0	0	0	0	0	0	1	0



Python and NetworkX



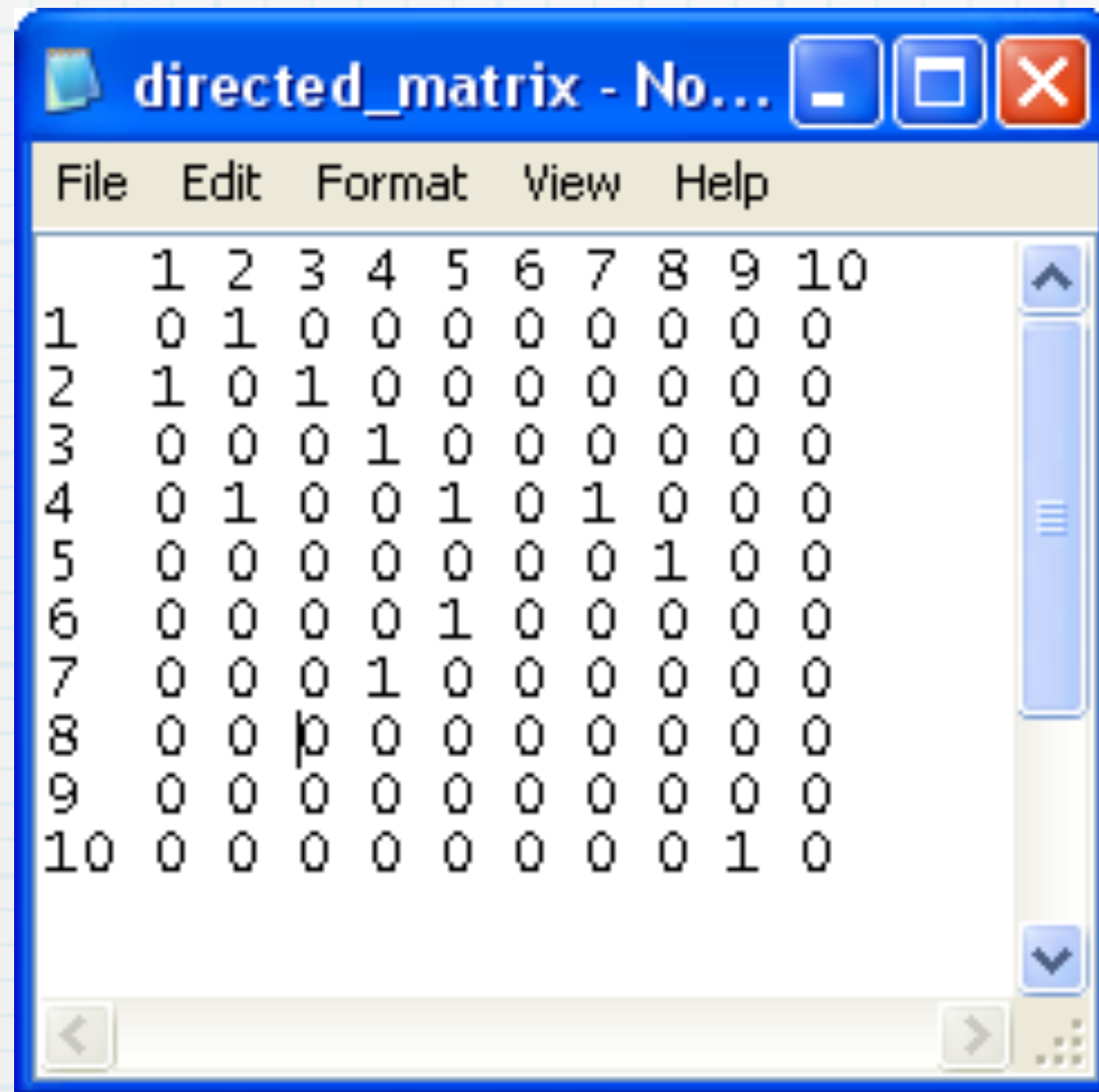
```
print(nx.adjacency_matrix(D))  
D.edge[3][4]  
D.edge[4][3]  
D.edge[4]
```

More local measures

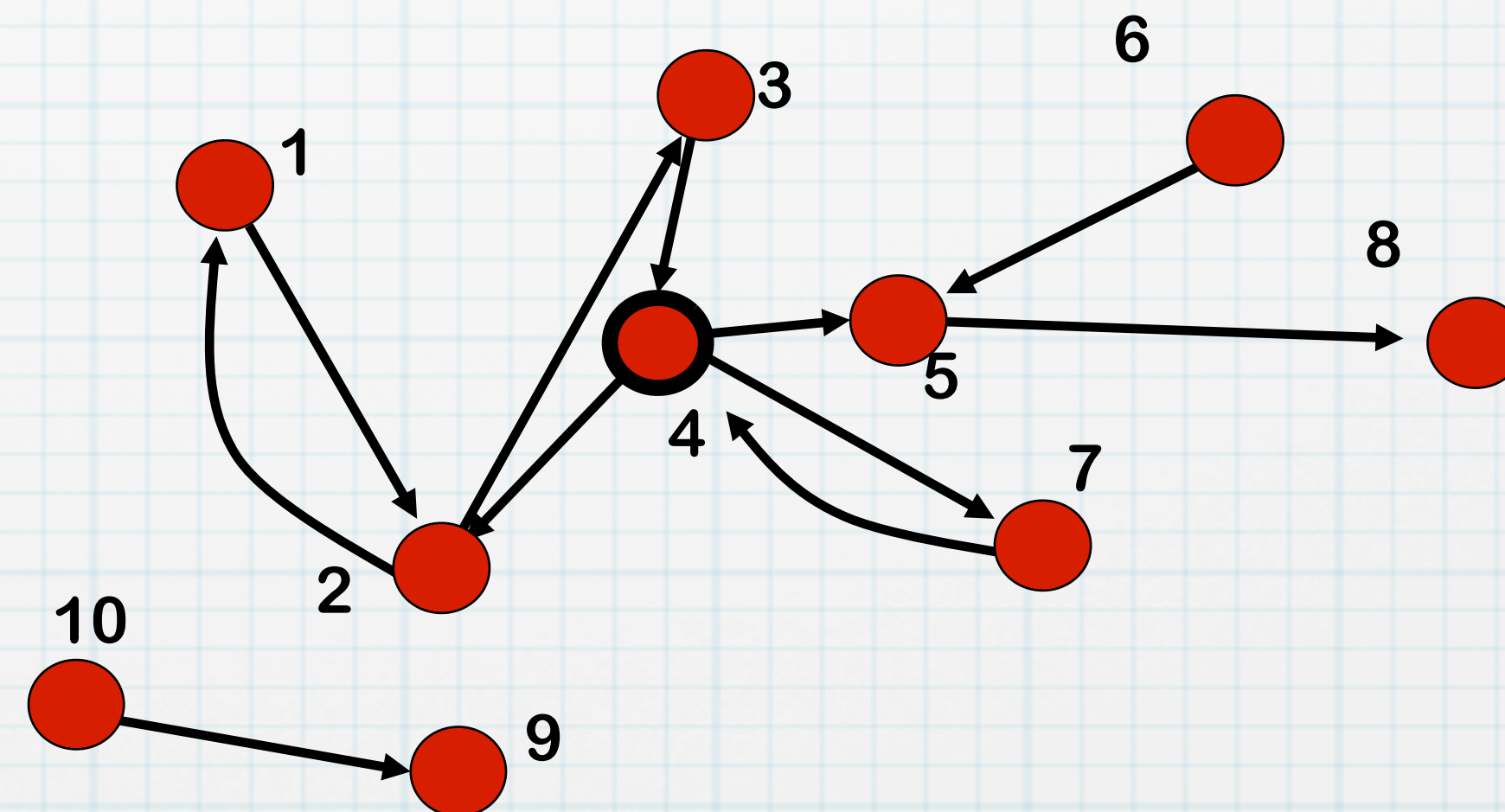
Directed networks have in-degree and out-degree:

$$k_{in}(i) = \sum_j A_{ji}$$

$$k_{out}(i) = \sum_j A_{ij}$$



	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	0	0	0	0	0	0	0
3	0	0	0	1	0	0	0	0	0	0
4	0	1	0	0	1	0	1	0	0	0
5	0	0	0	0	0	0	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0
10	0	0	0	0	0	0	0	0	1	0

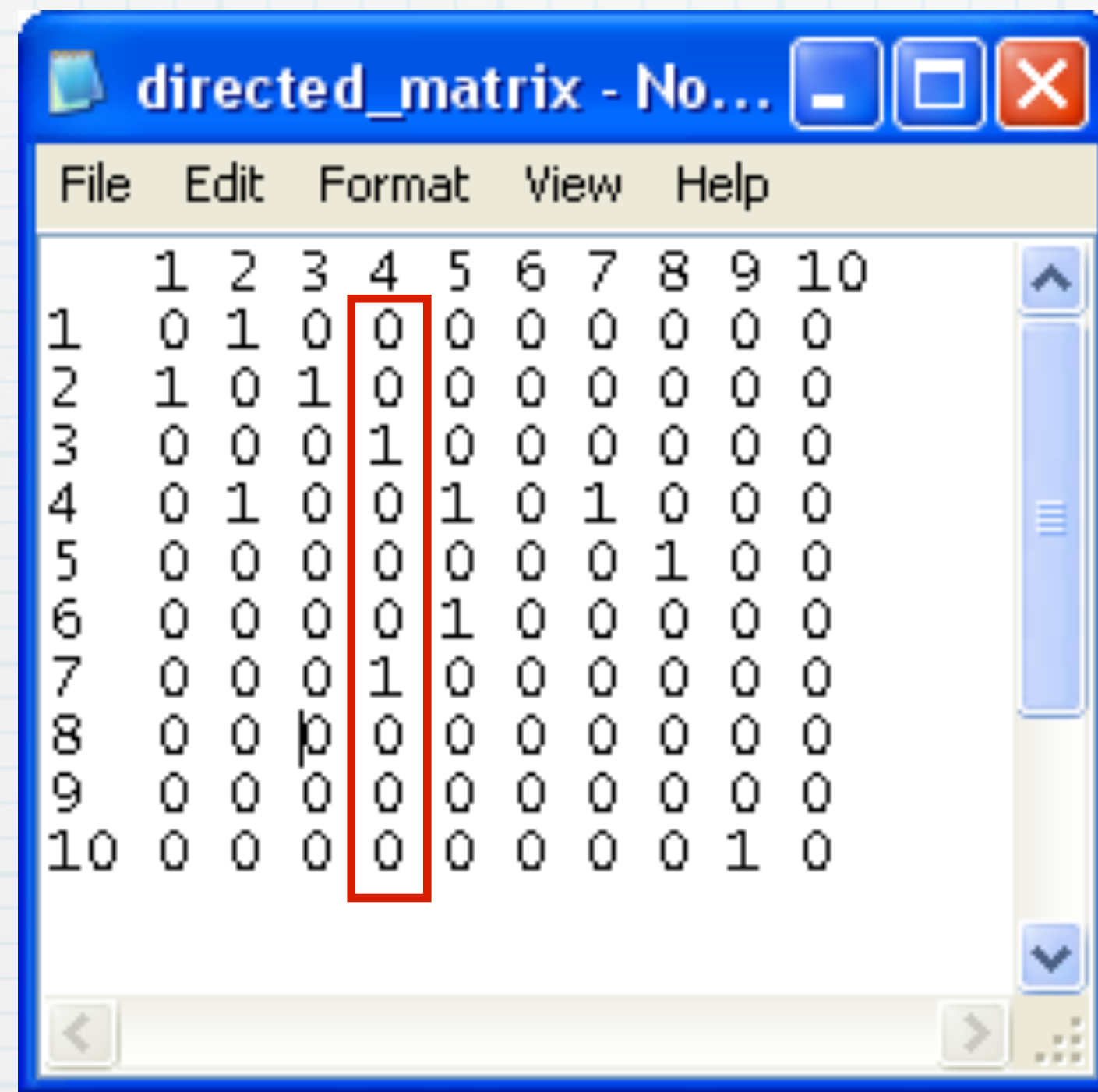


More local measures

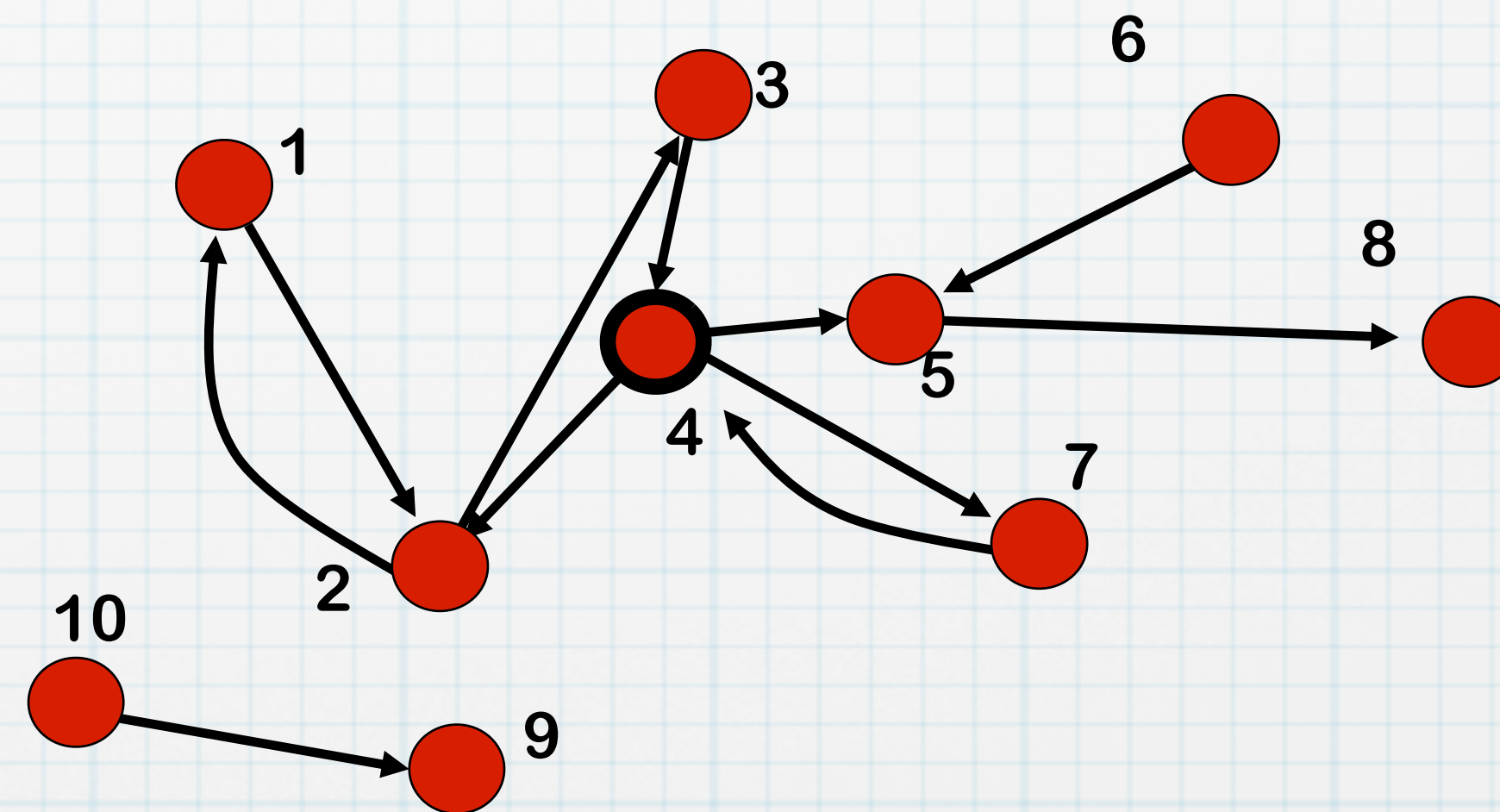
Directed networks have in-degree and out-degree:

$$k_{in}(i) = \sum_j A_{ji}$$

$$k_{out}(i) = \sum_j A_{ij}$$



	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	0	0	0	0	0	0	0
3	0	0	0	1	0	0	0	0	0	0
4	0	1	0	0	1	0	1	0	0	0
5	0	0	0	0	0	0	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0
10	0	0	0	0	0	0	0	0	1	0



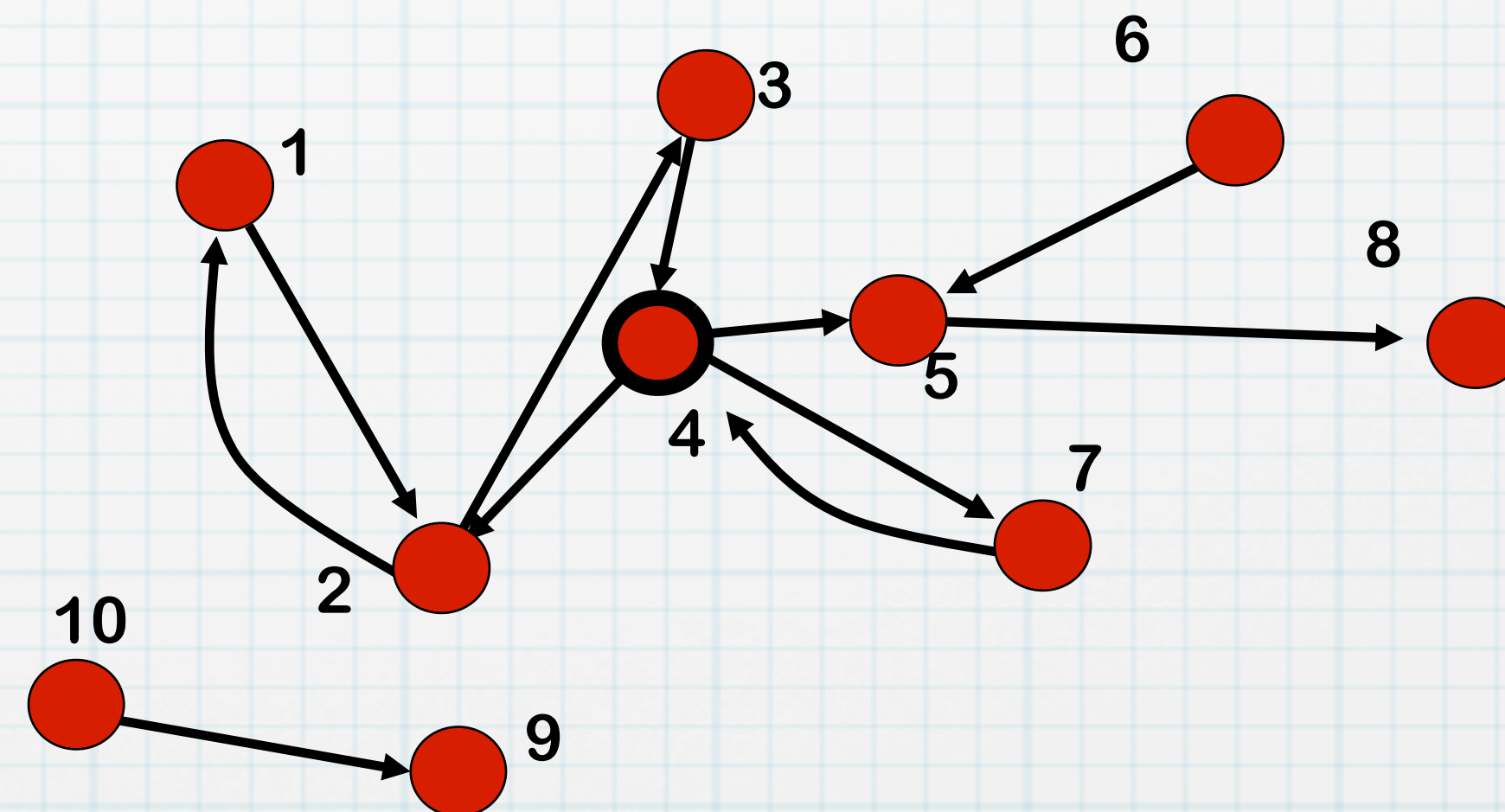
More local measures

Directed networks have in-degree and out-degree:

$$k_{in}(i) = \sum_j A_{ji}$$

$$k_{out}(i) = \sum_j A_{ij}$$

	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	0	0	0	0	0	0	0
3	0	0	0	1	0	0	0	0	0	0
4	0	1	0	0	1	0	1	0	0	0
5	0	0	0	0	0	0	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0
10	0	0	0	0	0	0	0	0	1	0



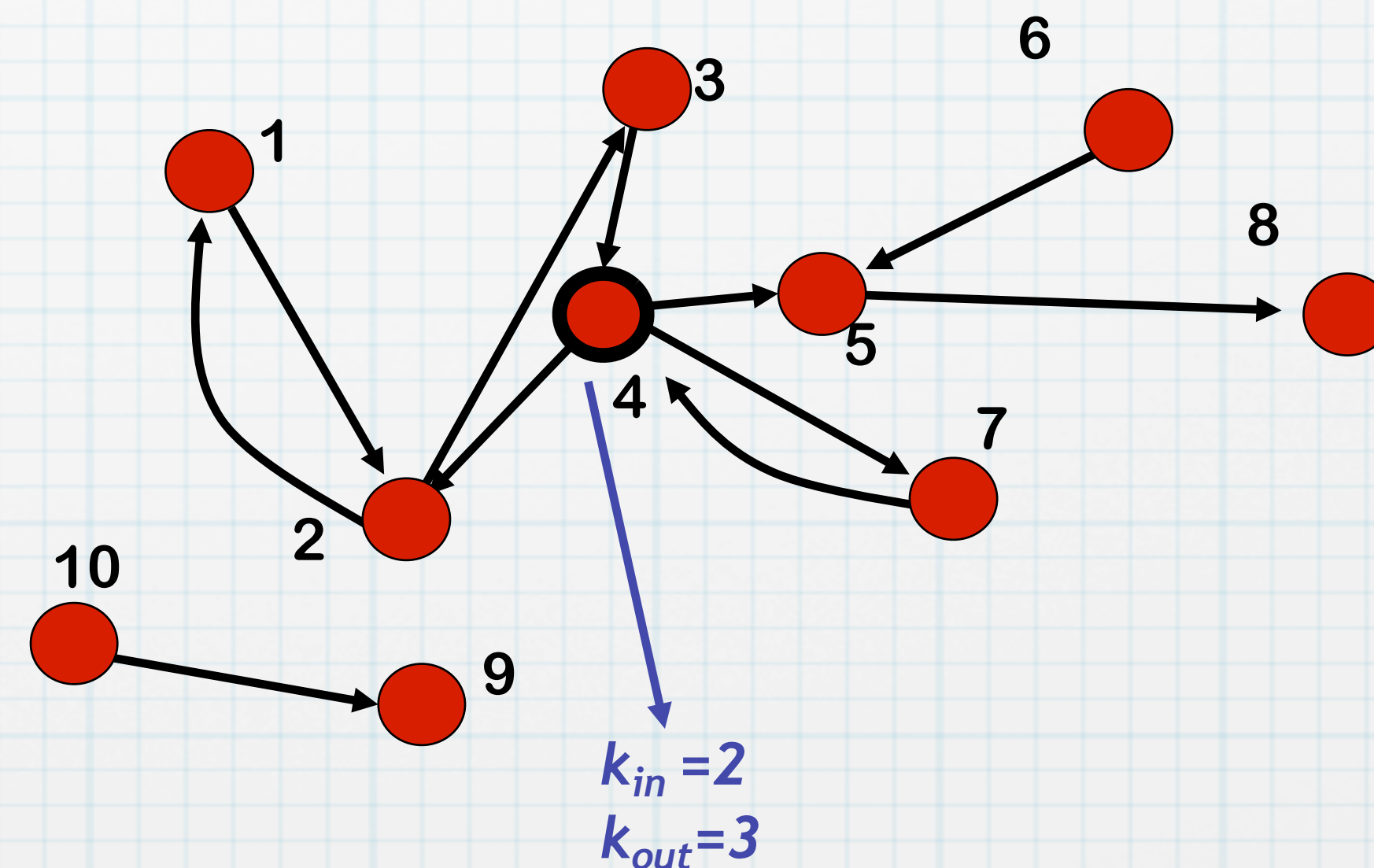
More local measures

Directed networks have in-degree and out-degree:

$$k_{in}(i) = \sum_j A_{ji}$$

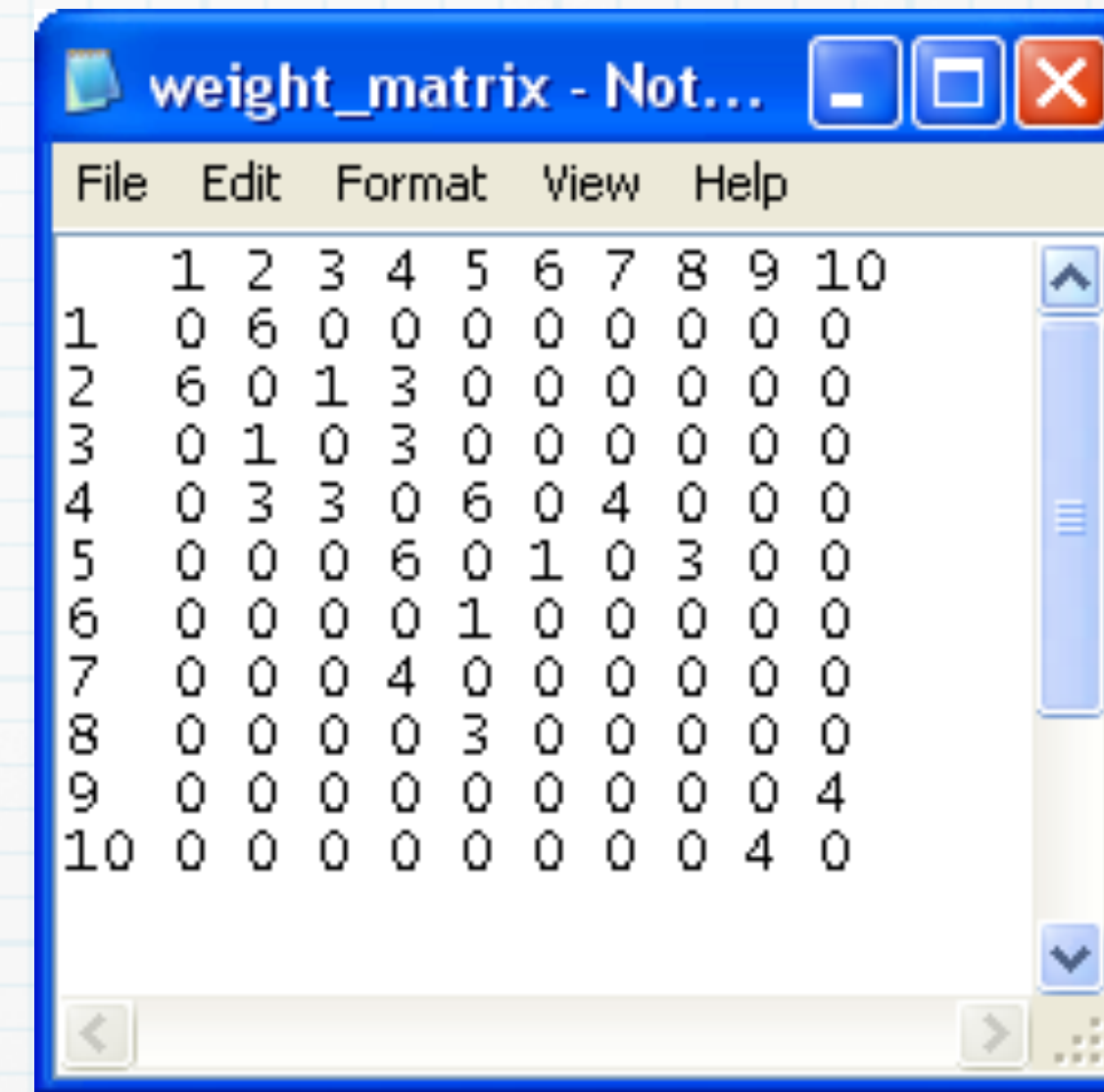
$$k_{out}(i) = \sum_j A_{ij}$$

	1	2	3	4	5	6	7	8	9	10
1	0	1	0	0	0	0	0	0	0	0
2	1	0	1	0	0	0	0	0	0	0
3	0	0	0	1	0	0	0	0	0	0
4	0	1	0	0	1	0	1	0	0	0
5	0	0	0	0	0	0	0	1	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	1	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0
10	0	0	0	0	0	0	0	0	1	0

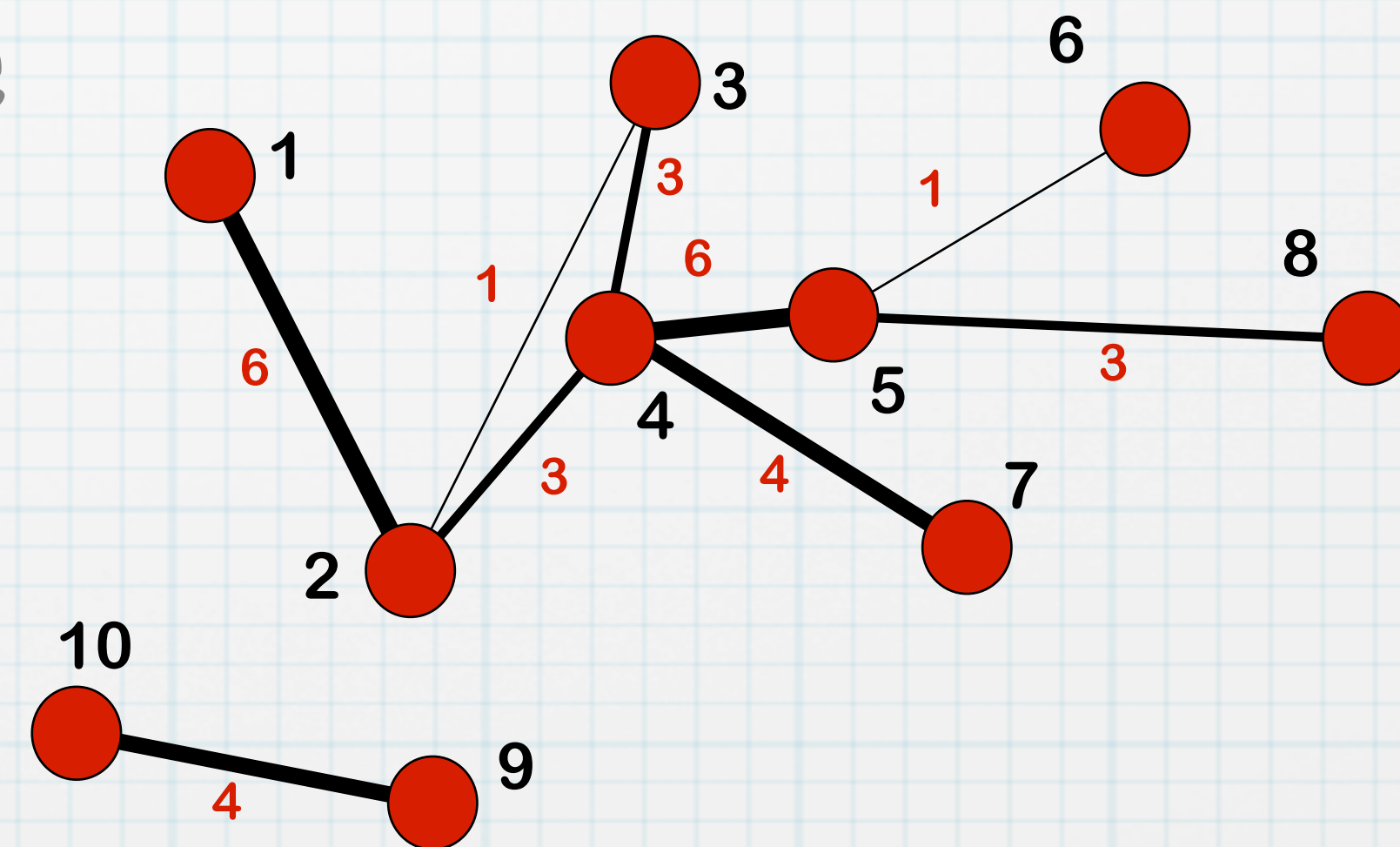


Mathematical representation

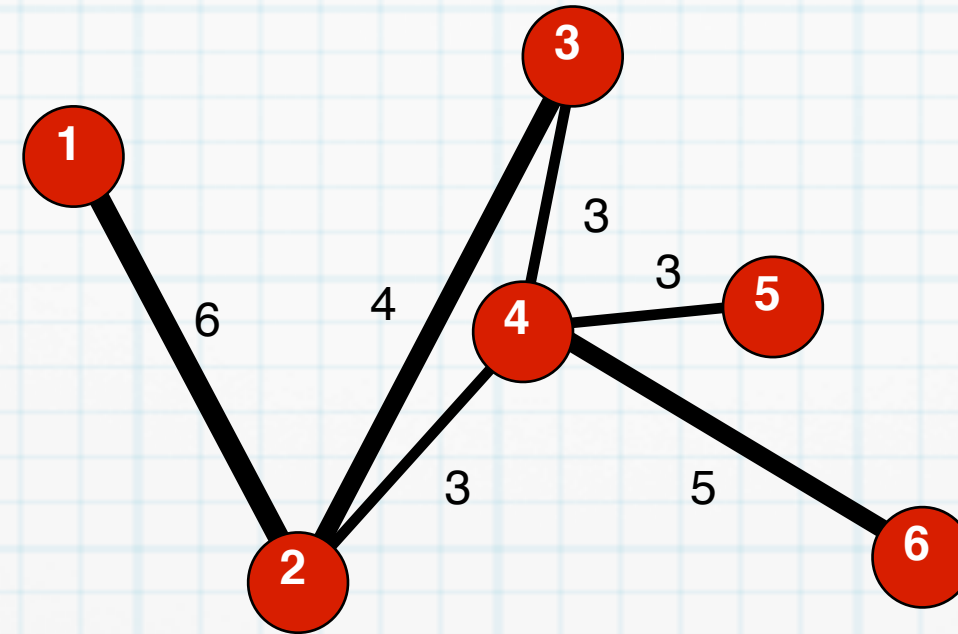
- * weighted networks:
weight matrix
- * undirected \rightarrow symmetric
- * directed \rightarrow asymmetric
- * w_{ij} = weight of link $i \rightarrow j$
- * $w_{ij} = 0$ if no links



	1	2	3	4	5	6	7	8	9	10
1	0	6	0	0	0	0	0	0	0	0
2	6	0	1	3	0	0	0	0	0	0
3	0	1	0	3	0	0	0	0	0	0
4	0	3	3	0	6	0	4	0	0	0
5	0	0	0	6	0	1	0	3	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	4	0	0	0	0	0	0
8	0	0	0	0	3	0	0	0	0	0
9	0	0	0	0	0	0	0	0	4	0
10	0	0	0	0	0	0	0	0	4	0



Python and NetworkX



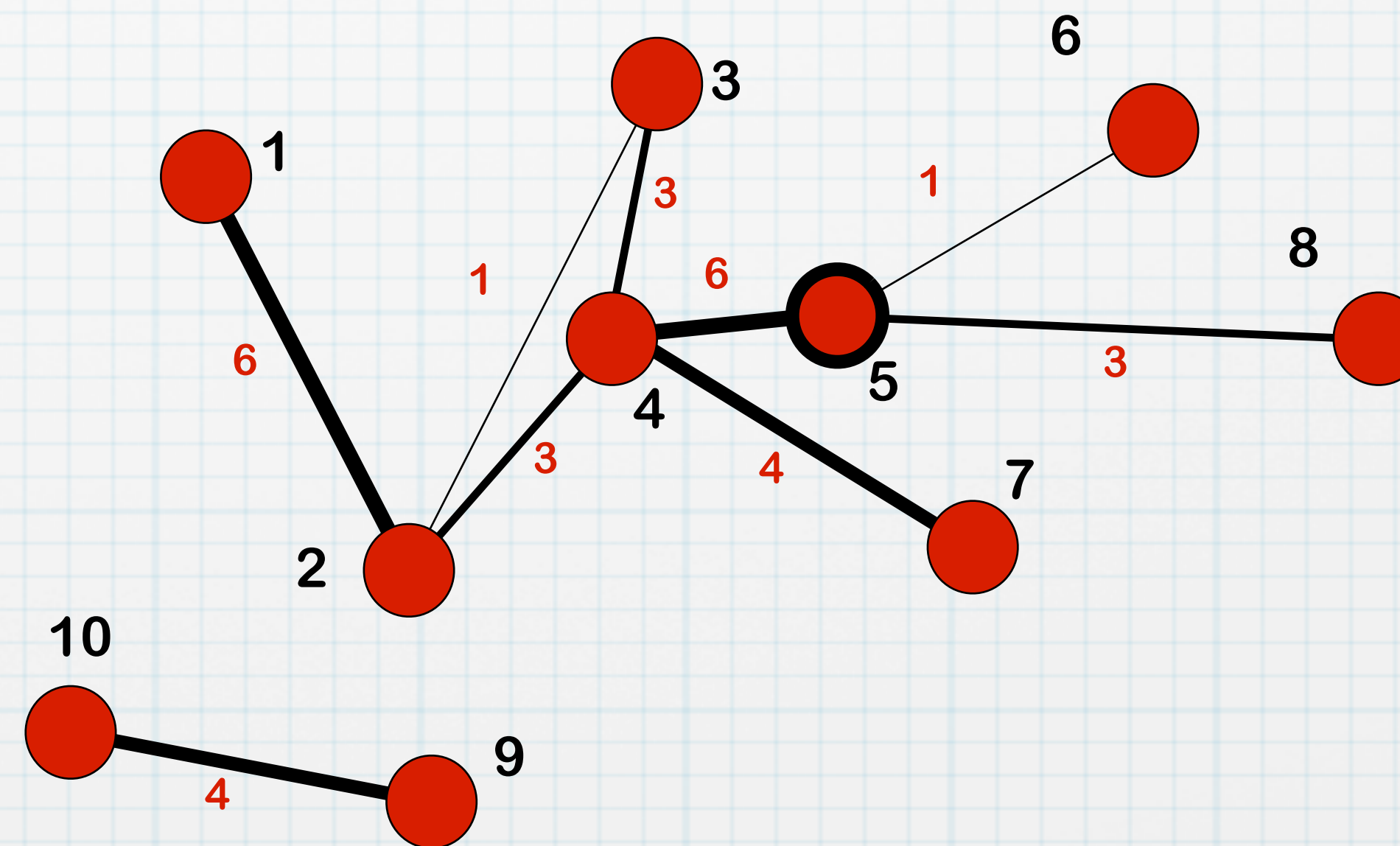
```
print(nx.adjacency_matrix(W))  
W.edge[2][3]  
W.edge[2]  
W.edge[2][3]['weight'] = 1  
W.edge[2][3]  
W.edge[2]
```

More local measures

In weighted networks, degree generalizes to strength (a.k.a. weighted degree):

$$S_i = W_{i1} + W_{i2} + \dots + W_{iN} = \sum_j W_{ij}$$

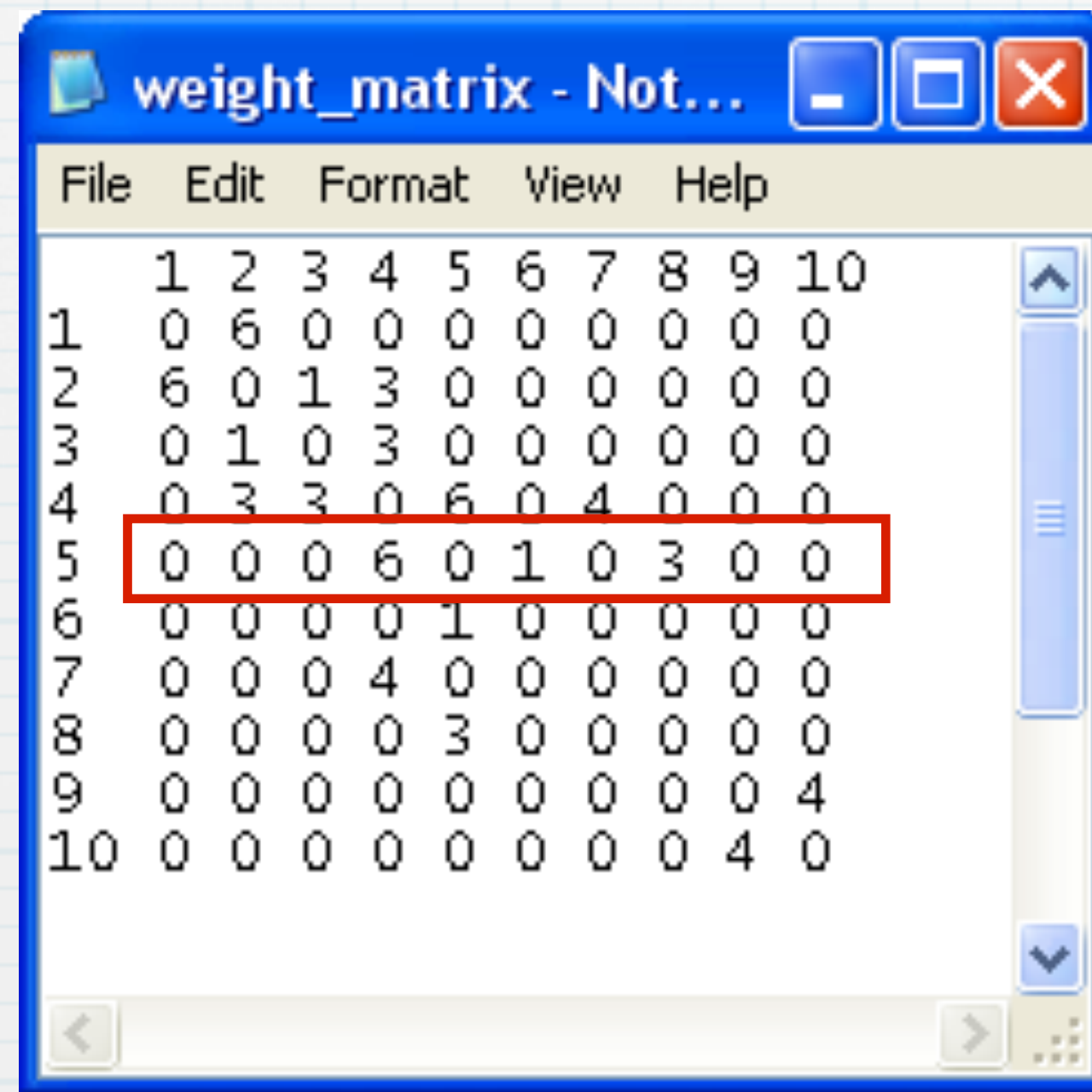
	1	2	3	4	5	6	7	8	9	10
1	0	6	0	0	0	0	0	0	0	0
2	6	0	1	3	0	0	0	0	0	0
3	0	1	0	3	0	0	0	0	0	0
4	0	3	3	0	6	0	4	0	0	0
5	0	0	0	6	0	1	0	3	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	4	0	0	0	0	0	0
8	0	0	0	0	3	0	0	0	0	0
9	0	0	0	0	0	0	0	0	4	0
10	0	0	0	0	0	0	0	0	4	0



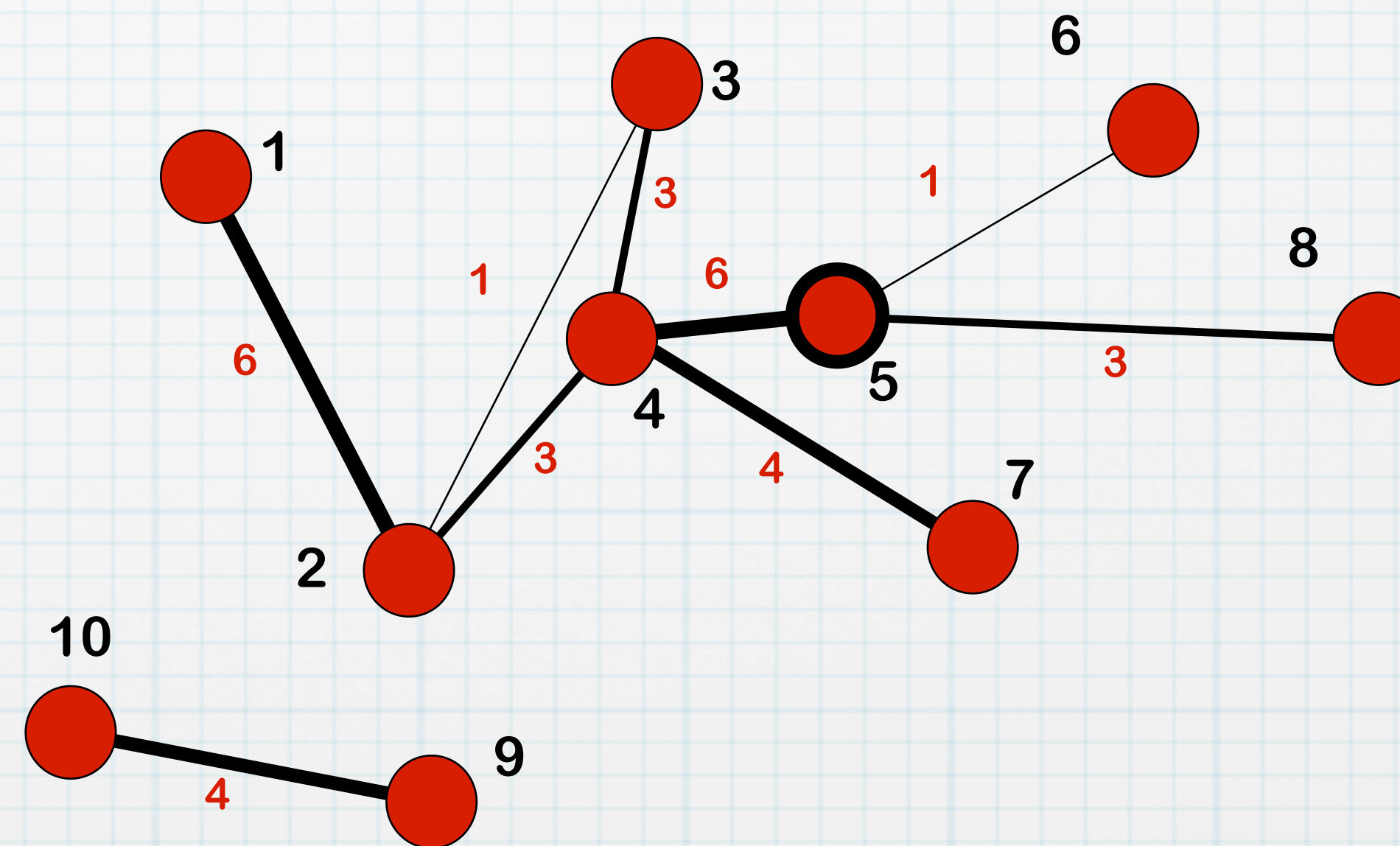
More local measures

In weighted networks, degree generalizes to strength (a.k.a. weighted degree):

$$S_i = W_{i1} + W_{i2} + \dots + W_{iN} = \sum_j W_{ij}$$



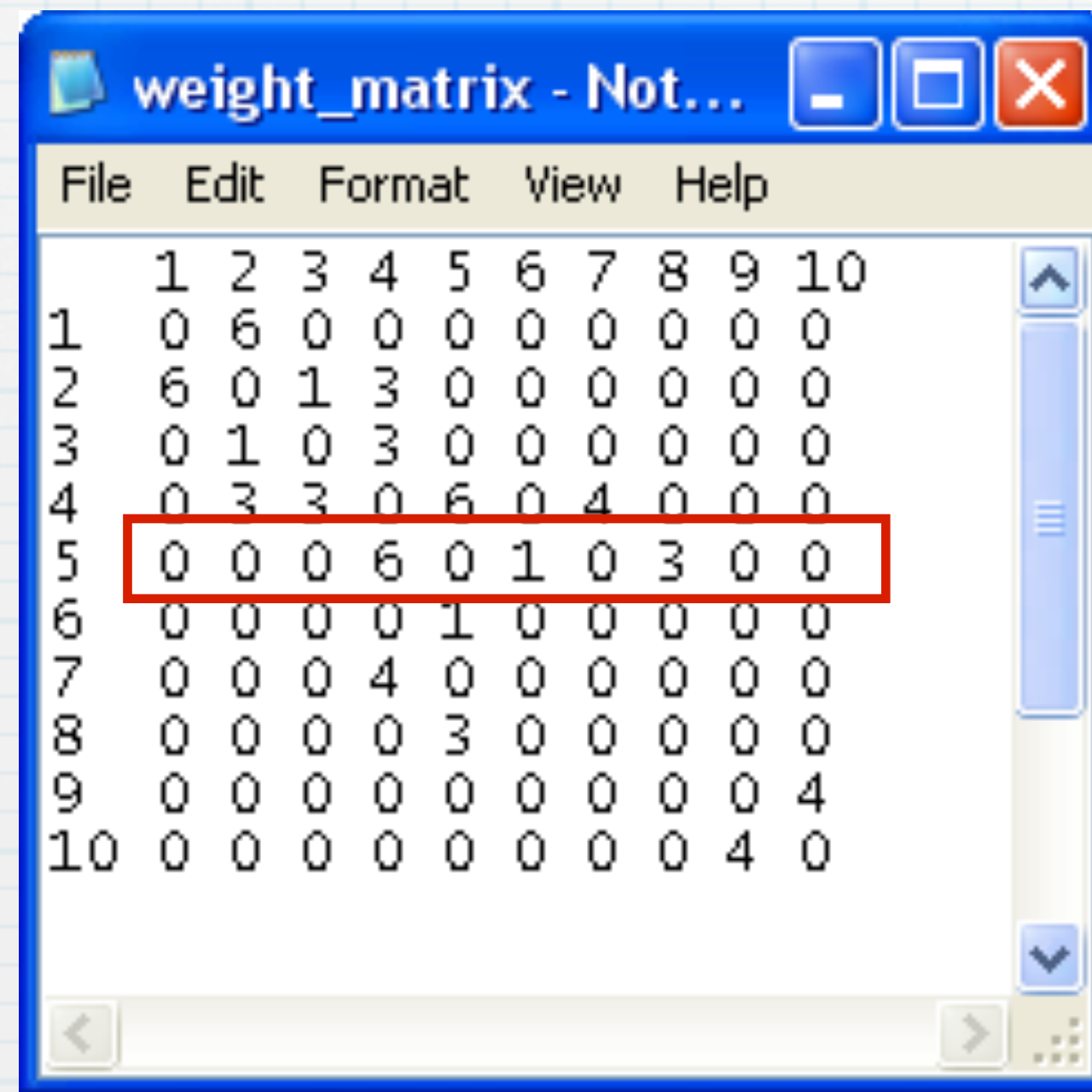
	1	2	3	4	5	6	7	8	9	10
1	0	6	0	0	0	0	0	0	0	0
2	6	0	1	3	0	0	0	0	0	0
3	0	1	0	3	0	0	0	0	0	0
4	0	3	3	0	6	0	4	0	0	0
5	0	0	0	6	0	1	0	3	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	4	0	0	0	0	0	0
8	0	0	0	0	3	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	4
10	0	0	0	0	0	0	0	0	4	0



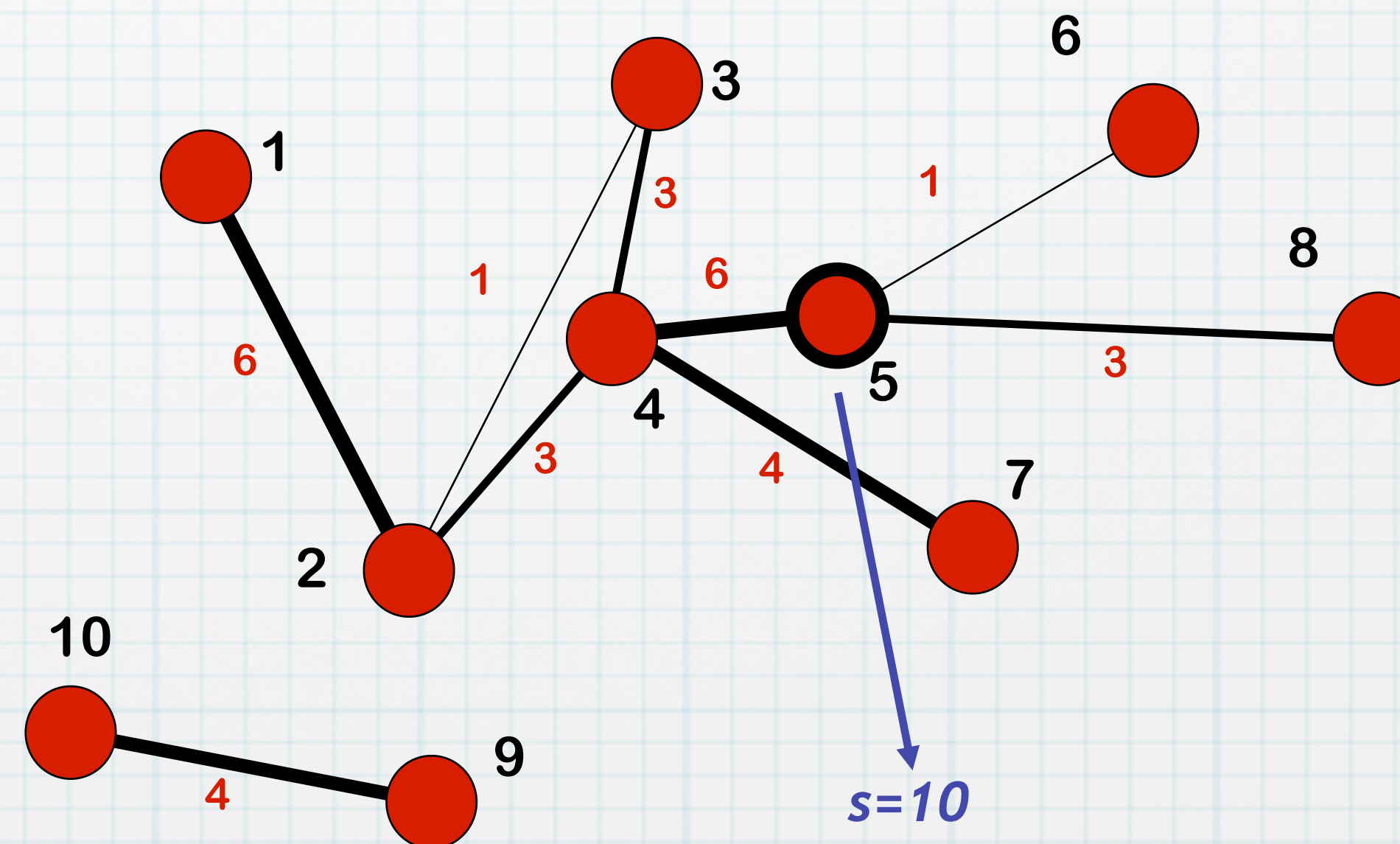
More local measures

In weighted networks, degree generalizes to strength (a.k.a. weighted degree):

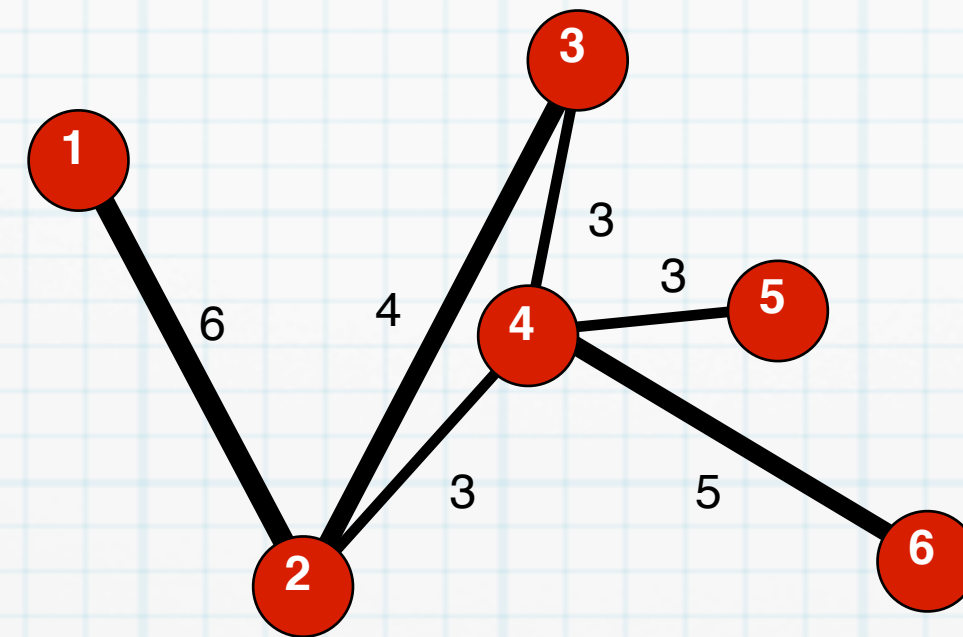
$$S_i = W_{i1} + W_{i2} + \dots + W_{iN} = \sum_j W_{ij}$$



	1	2	3	4	5	6	7	8	9	10
1	0	6	0	0	0	0	0	0	0	0
2	6	0	1	3	0	0	0	0	0	0
3	0	1	0	3	0	0	0	0	0	0
4	0	3	3	0	6	0	4	0	0	0
5	0	0	0	6	0	1	0	3	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	4	0	0	0	0	0	0
8	0	0	0	0	3	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	4
10	0	0	0	0	0	0	0	0	4	0



Python and NetworkX

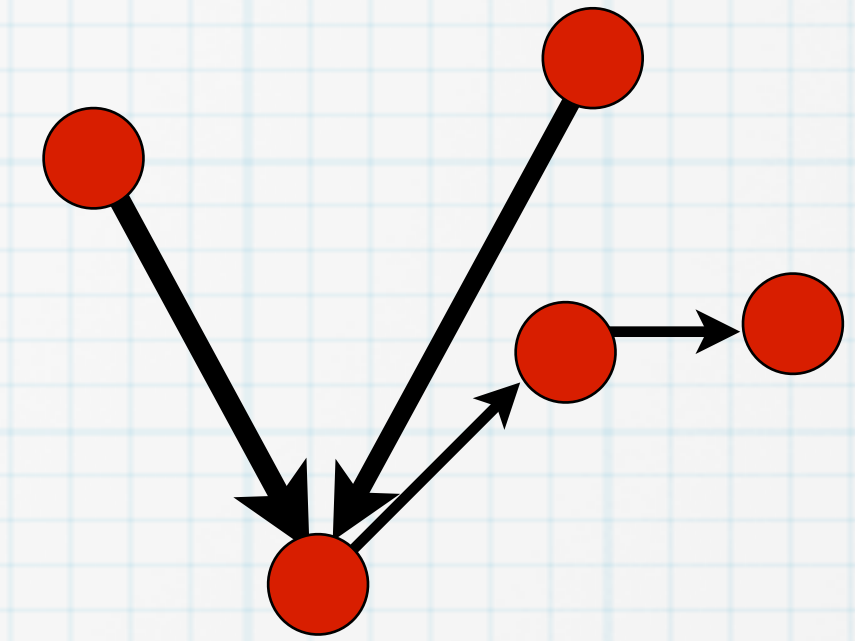


```
W.degree(4) # degree  
W.degree(4, weight='weight') # strength
```

More local measures

What about a directed, weighted network?

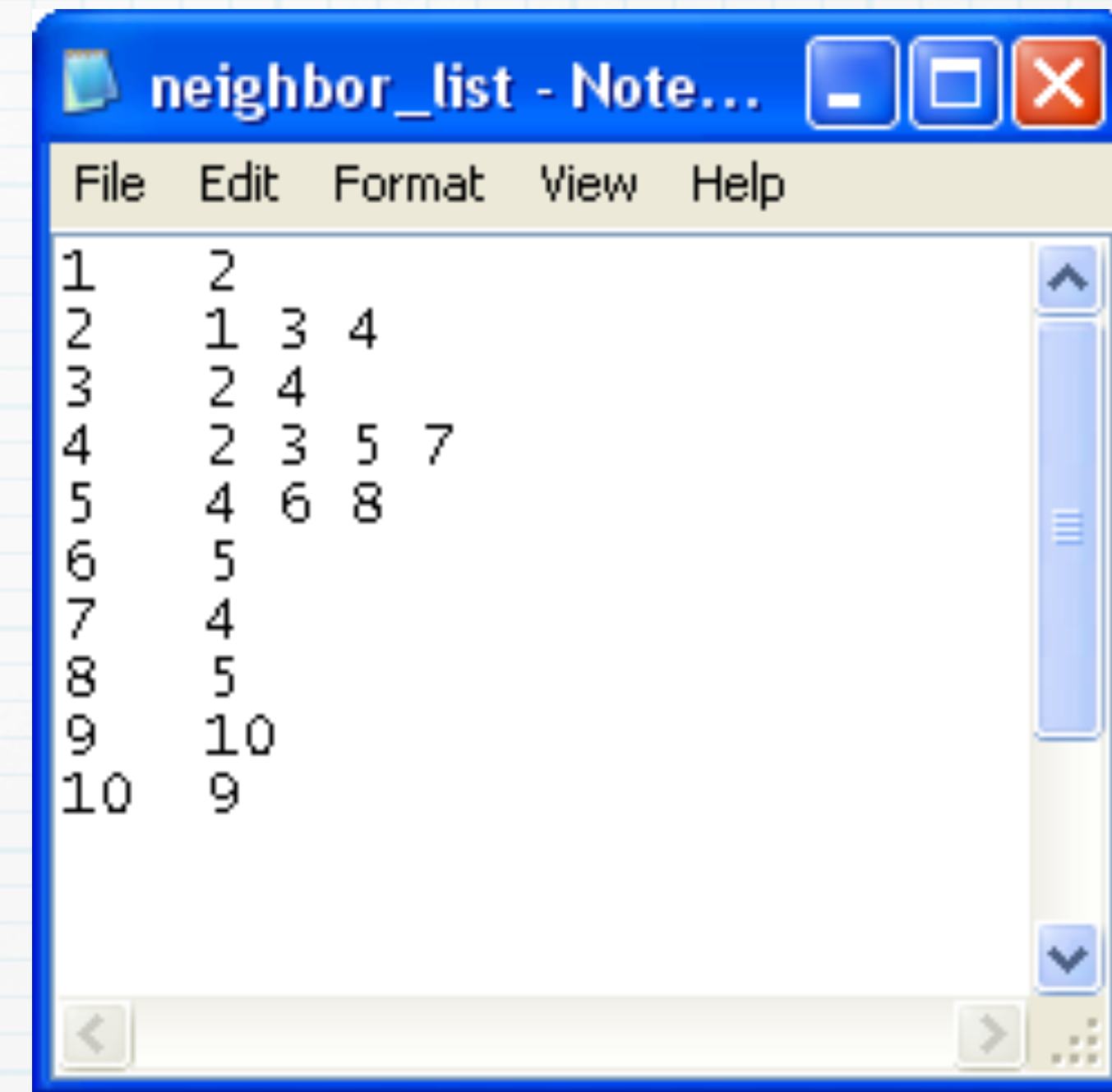
- In-strength (weighted in-degree)
- Out-strength (weighted out-degree)



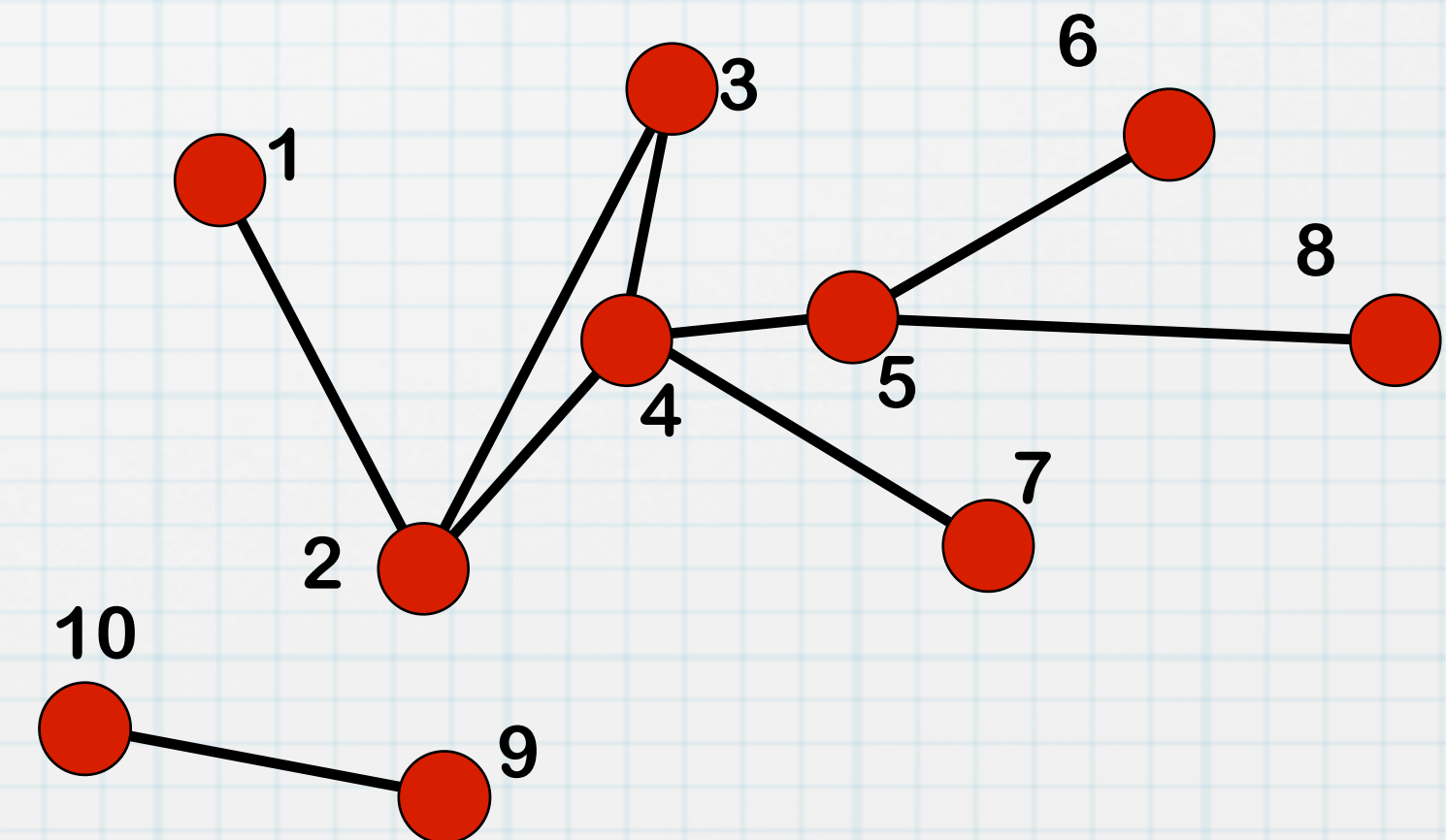
Exercise: create a small, weighted, directed network to represent "like" relationships among a group of friends. Print in- and out-strength of each node. (Hint: a `DiGraph` can have weight attributes, too.)

Mathematical representation

- * Neighbor list
 - * $i \quad n_1(i), n_2(i), n_3(i), \dots$
 - * useful for sparse graphs
 - * for every vertex: list of neighbors, i.e. k elements instead of N
- * a.k.a. adjacency list



1	2			
2	1	3	4	
3	2	4		
4	2	3	5	7
5	4	6	8	
6	5			
7	4			
8	5			
9	10			
10	9			

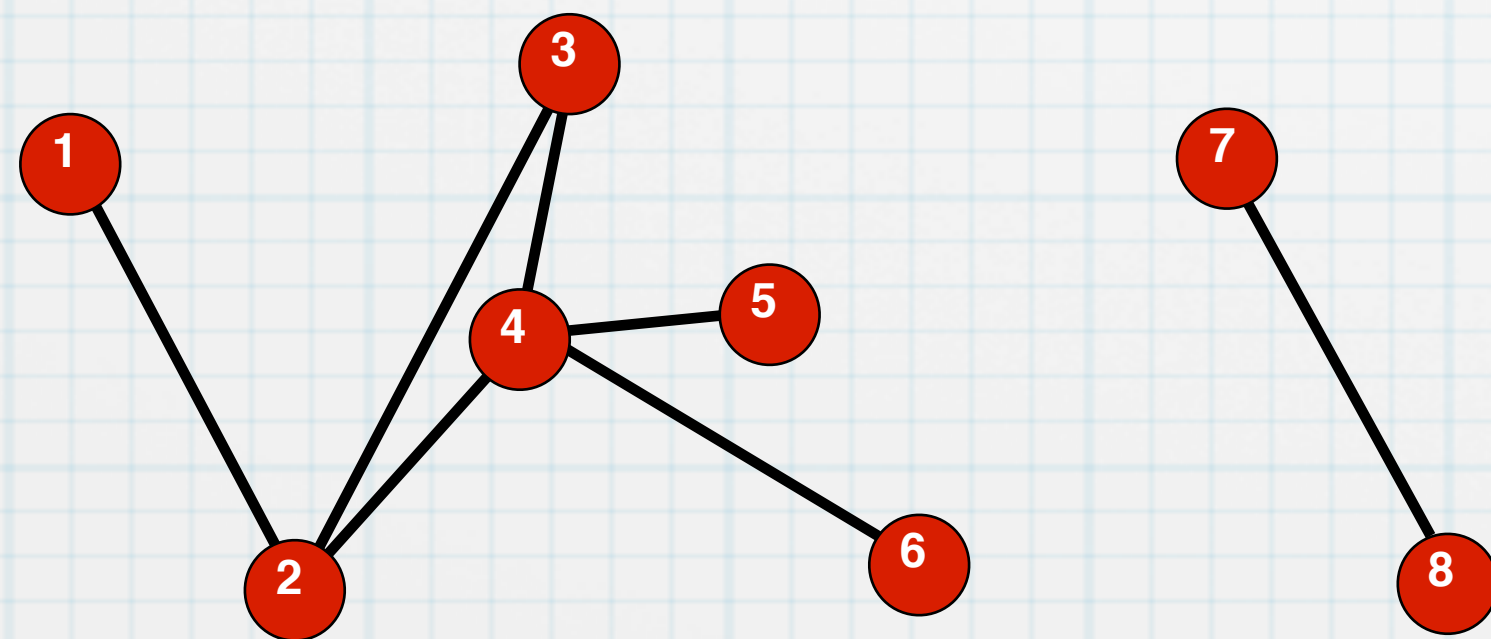


Python and NetworkX

```
G.neighbors(2)

for n,nbrs in G.adjacency():
    for nbr,eattr in nbrs.items():
        print('%d, %d' % (n,nbr))

nx.write_adjlist(G, "netfile.adjlist")
G2 = nx.read_adjlist("netfile.adjlist")
GM = nx.isomorphism.GraphMatcher(G,G2)
GM.is_isomorphic()
```



```
with open('netfile.adjlist') as f:
    for line in f:
        print(line)
```


Mathematical representation

* Edge list

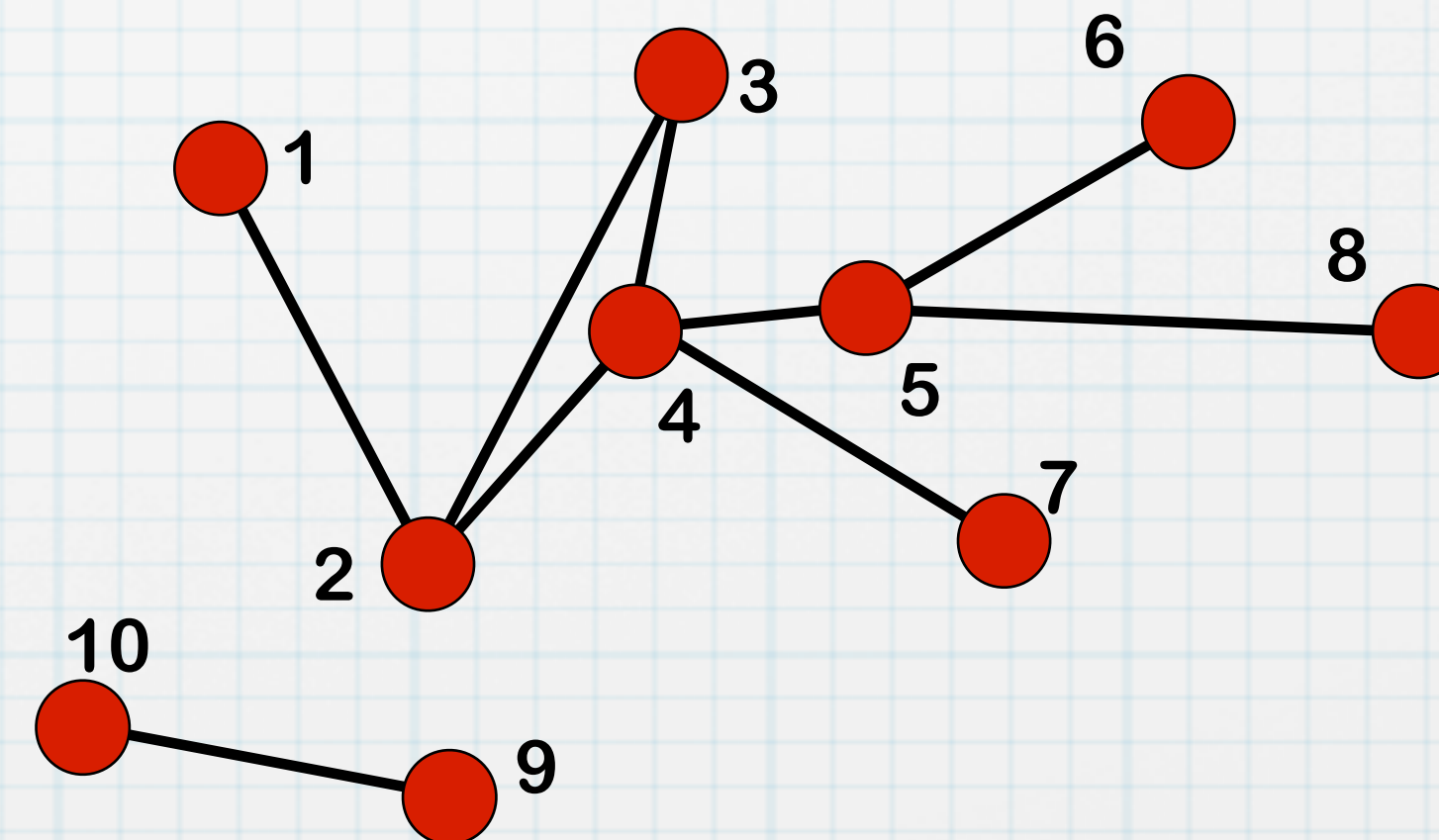
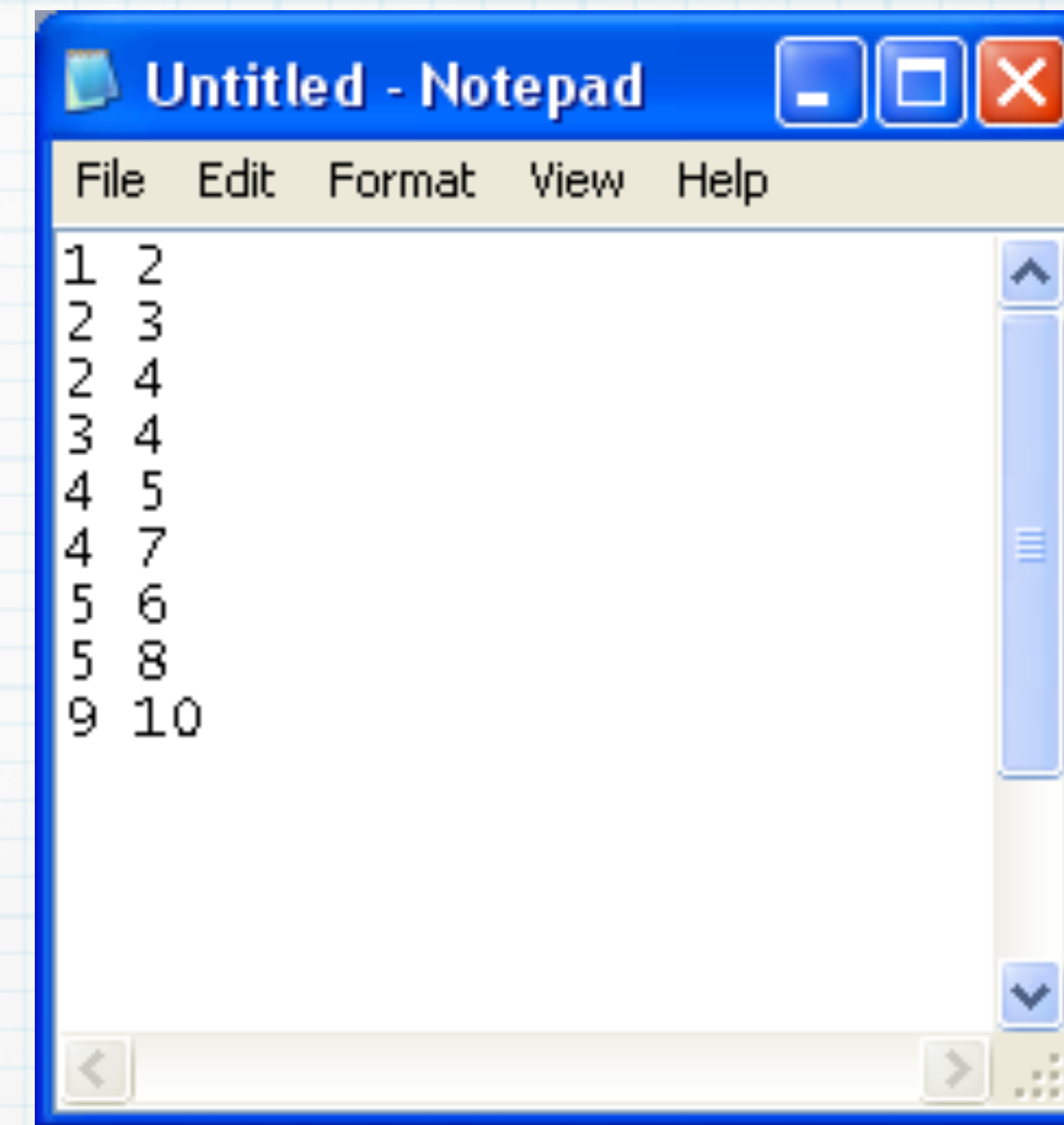
$i_1 j_1$

$i_2 j_2$

$i_3 j_3$

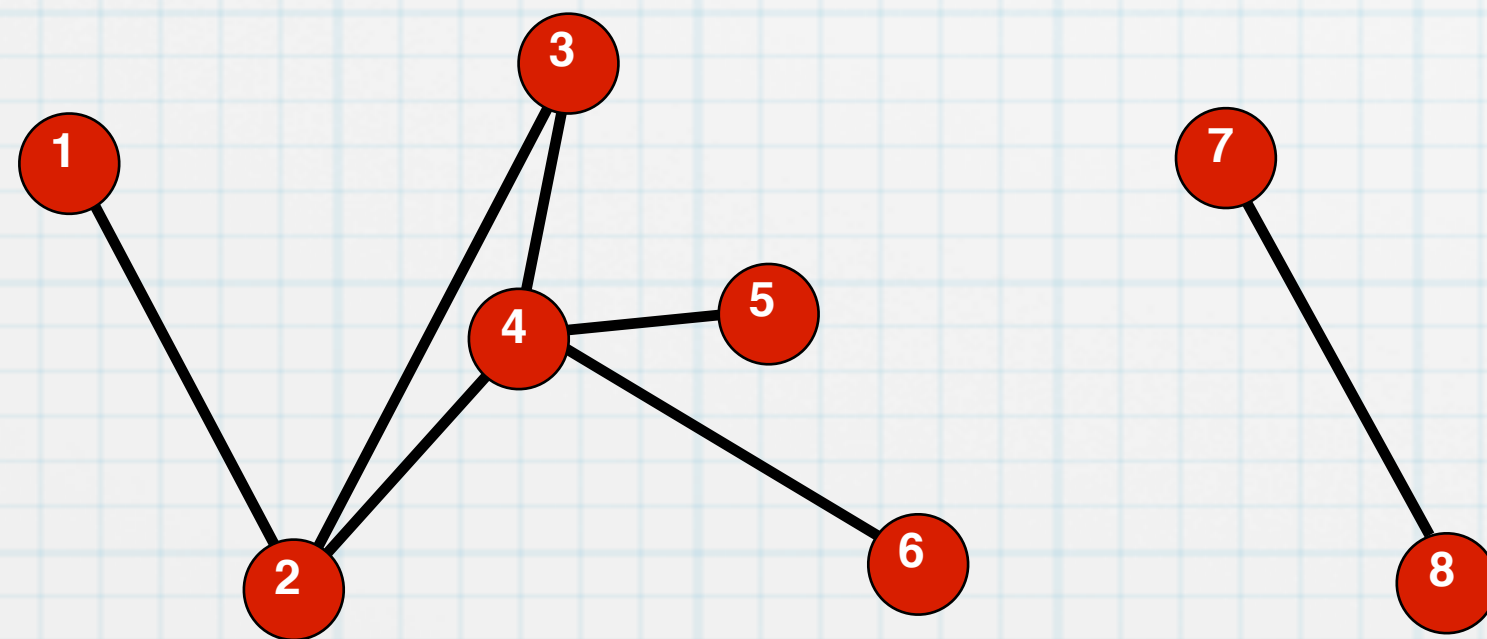
... ..

* list of E edges (each once)



Python and NetworkX

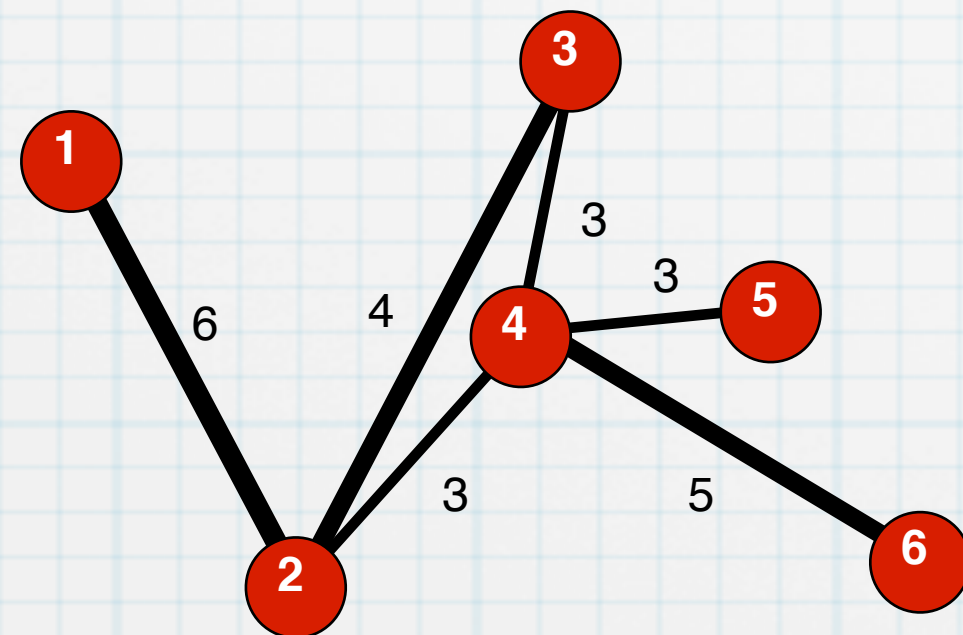
```
for i,j in G.edges:  
    print('%d %d' %(i,j))  
  
nx.write_edgelist(G, "netfile.edgelist")  
G3 = nx.read_edgelist("netfile.edgelist")  
GM = nx.isomorphism.GraphMatcher(G,G3)  
GM.is_isomorphic()
```



```
with open('netfile.edgelist') as f:  
    for line in f:  
        print(line)
```


Python and NetworkX

```
nx.write_weighted_edgelist(W, "netfile.edgelist")
W2 = nx.read_weighted_edgelist("netfile.edgelist")
GM = nx.isomorphism.GraphMatcher(W,W2)
GM.is_isomorphic()
```



```
with open('netfile.edgelist') as f:
    for line in f:
        print(line)
```

Drawing (with matplotlib)

```
In [96]: nx.draw(G)
```

