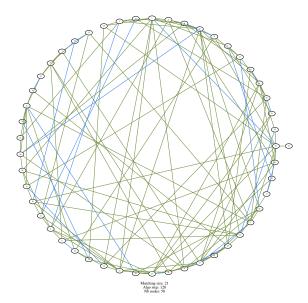
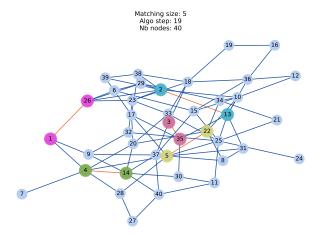
Algorithmic complexity and graphs: the matching problem

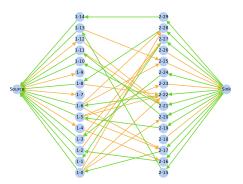
September 14, 2024

Introduction

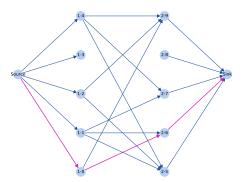




residual graph step 12



augmenting path step 1



The matching problem

The matching problem

Definition of the problem Experimental solutions Brute force algorithm Greedy algorithm

Problem 1: Max Flow

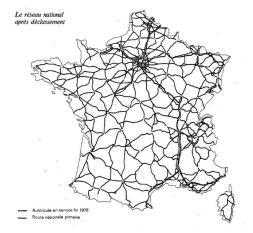


Figure: Problem 1: transporting merchandise through a network

Problem 2 : Maximum matching (Optimal assignment, problème d'affectation)

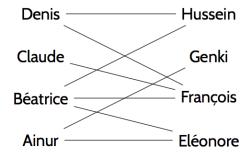


Figure: Problem 2 : Building the largest possible number of teams of 2 persons.

Problem 2

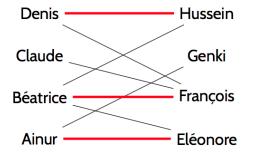


Figure: Problem 2: not optimal assignment

Problem 2

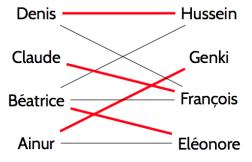


Figure: Problem 2: optimal assignment

Other examples

- Assigning students to internships
- Assigning machines to a task

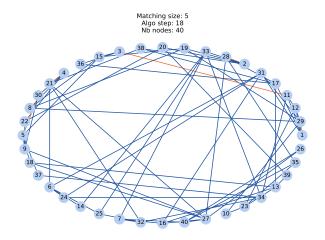
Summary

- ▶ Today we will work on **connnecting the two problems**.
- In some specific cases, the two problems equivalent.

- The matching problem
 - └ Definition of the problem

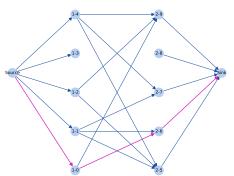
Networkx

We will use networkx.



Networkx

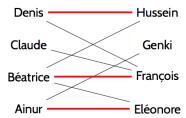
augmenting path step 1



Matching problem

Given a undirected graph G = (V, E), we want a matching M, which means:

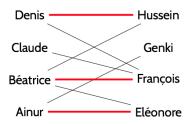
▶ A subset of edges $M \subset E$



Matching problem

Given a undirected graph G = (V, E), we want a matching, which means:

- ▶ A subset of edges $M \subset E$
- Such that no pairs of edges of M are incident
- Equivalently, each node in the graph is at most in one edge of M.

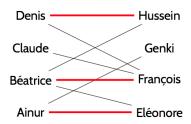


— Definition of the problem

Matching problem

Given undirected a graph G = (V, E), we want a matching, which means:

- ▶ A subset of edges $M \subset E$
- Equivalently, each node in the graph is at most in one edge of M.
- ▶ No pairs of edges of *M* are incident



Maximum matching

▶ The size of a matching is the number of edges it contains.

Maximum matching

- ▶ The size of a matching is the number of edges it contains.
- ▶ We want to find the matching of largest possible size in a given graph.

☐ Definition of the problem

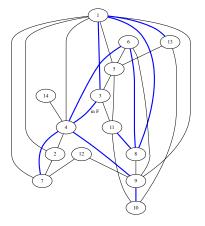


Figure: Is this a matching?

Definition of the problem

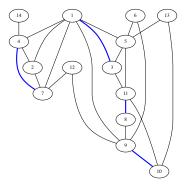


Figure: Is this a matching?

☐ Definition of the problem

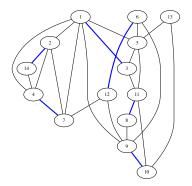


Figure: Is this an optimal matching?

Definition of the problem

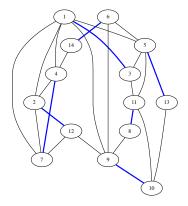


Figure: Is this an optimal matching?

Definition of the problem

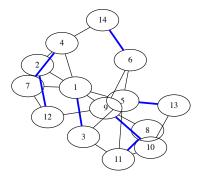


Figure: With neato

Exercice 1: What is the maximum size possible for a matching, in a general graph of size n? (in the sense that no graph with n nodes contains a larger matching)

Exercice 1: What is the maximum size possible for a matching, in a general graph of size n? (in the sense that no graph with n nodes contains a larger matching)

- ▶ If *n* is even : $\frac{n}{2}$
- ► Else *n* is odd : $\frac{n-1}{2}$

Hence,

$$\lfloor \frac{n}{2} \rfloor$$
 (1)

Exercice 1: Can you think of a graph with n nodes that contains a matching of size $\frac{n}{2}$? (assuming n is even)

Optimal

Exercice 1: Can you think of a graph with n nodes that contains a matching of size $\frac{n}{2}$? (assuming n is even)

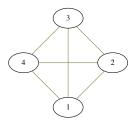


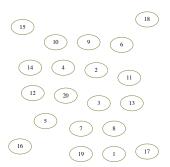
Figure: The complete graph works

Exercice 1: Can you think of a graph with n nodes that does **not** contains a matching of size $\frac{n}{2}$? (assuming n is even)

Definition of the problem

Optimal matching

Exercice 1: Can you think of a graph with n nodes that does **not** contains a matching of size $\frac{n}{2}$? (assuming n is even)



Exercice 1: Can you think of a **non trivial** graph that does **not** contains a matching of size $\frac{n}{2}$? (assuming n is even)

Definition of the problem

Optimal matching

Exercice 2: Can you think of a **non trivial** graph that does **not** contains a matching of size $\frac{n}{2}$? (assuming n is even)

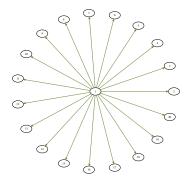


Figure: Star graph

Experiments

Possibilities to code a graph:

- list of sets of size 2 (for an undirected graph)
- a dictionary of successors (directed of undirected)

Coding a graph: as a list of edges



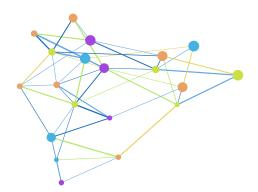
$$g1 = [\{1,2\},\{1,3\},\{2,3\},\{3,4\},\{1,4\}]$$

Coding a graph: as a dictionary of neighbors

$$g1 = \{ 1: \{2,3,4\}, 2: \{1,3\}, 3: \{1,2,4\}, 4: \{1,3\} \}$$

Experimental solutions

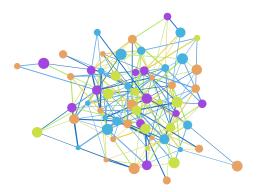
$Generating\ graphs\ with\ networks.$



Overview

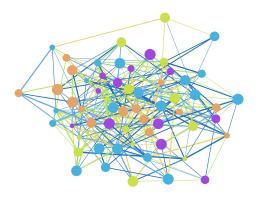
The matching problem

 $\mathrel{\sqsubseteq}_{\mathsf{Experimental}} \mathsf{solutions}$

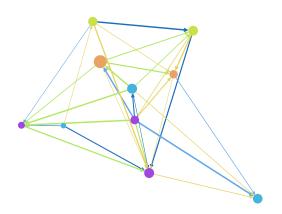


Overview

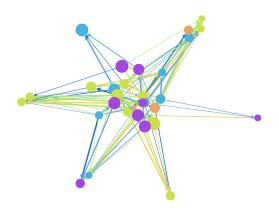
The matching problem
Experimental solutions



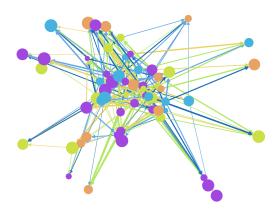
Directed graph



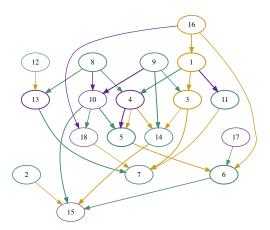
Directed graph II



Directed graph III



Example directed graph



Manual matching

Exercice 3: Please manually find an **optimal matching** in your **undirected** graph.

Big graph

We could not manually find an optimal matching in this graph :



Summary

- ▶ We have defined the matching problem.
- ▶ When the size of the problem is large, we can not manually find an optimal matching.

Exercice 4: Enumeration

Given a graph, what would a brute force approach on the matching problem be ?

Exercice 4: Exhaustive search

- Given a graph what would a brute force approach on the matching problem be ?
 - ▶ 1) Enumerate all possible subsets in the set of the edges.
 - 2) Check if each subset is a matching.
 - 3) Return the biggest one obtained.

Exercice 4: Exhaustive search

- Given a graph what would a brute force approach on the matching problem be ?
 - ▶ 1) Enumerate all possible subsets in the set of the edges.
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If the graph contains n nodes, and given a subset of edges, what if the number of computations needed to perform step 2 ?

Exercice 4: Exhaustive search

- Given a graph what would a brute force approach on the matching problem be ?
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 - 2) Check if each subset is a matching.
 - 3) Return the biggest one obtained.

If the graph contains n nodes, and given a subset of edges, what if the number of computations needed to perform step 2 ? You can give a rough approximation.

Exercice 4: Exhaustive search

- Given a graph what would a brute force approach on the matching problem be ?
 - ▶ 1) Enumerate all possible subsets in the set of the edges.
 - 2) Check if each subset is a matching.
 - 3) Return the biggest one obtained.

If the graph contains n nodes, and given a subset of edges, what if the number of computations needed to perform step 2 ? It is a **polynomial** number of computations.

Brute force search

Exercice 5 : Complexity of brute force

- ▶ 1) Enumerate all possible subsets in the set of the edges.
- 2) Check if each subset is a matching.
- ▶ 3) Return the biggest one obtained.

What is the complexity of step 1?

Brute force search

Exercice 5 : Complexity of brute force

- ▶ 1) Enumerate all possible subsets in the set of the edges.
- ▶ 2) Check if each subset is a matching.
- ▶ 3) Return the biggest one obtained.

What is the complexity of step 1?

The number of subsets is $2^{\frac{n(n-1)}{2}}$ (in the worst case), which is exponential. If p is the number of edges, we can also write it as 2^p .

Brute force search

Exercice 5: Complexity of brute force

Assume that checking a subset requires 1 microsecond. How long should we wait in order to check all possible matchings in a graph with 100 nodes ?

Summary II

- For the matching problem on a large graph, we can neither
 - manually find an optimal matching
 - perform the exhaustive search (brute force algorithm)

Algorithms

- ▶ Hence, we need different algorithms to solve the problem.
- Let us first introduce some theoretical notions.

Notion of maximal and maximum matching

We will say that a matching M of cardinality (number of elements) |M| is:

 Maximum if is has the maximum possible number of edges (it is thus optimal)

Notion of maximal and maximum matching

We will say that a matching M of cardinality |M| is:

- Maximum if it has the maximum possible number of edges (it is thus optimal)
- ▶ Maximal if the set of edges obtained by adding any edge to it is not a matching. This means that $M \cup \{e\}$ is not a matching for any e not in M.
- ▶ ∪ means union of sets.

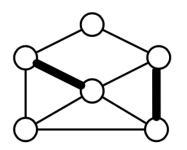
Is being a maximal matching the same thing as beeing a maximum matching ?

Maximum implies maximal

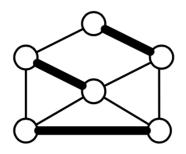
Let us show that a maximum matching is maximal.

Counter Example

However, a matching that is maximal is **not necessary Maximum** (example).



(a) A maximal matching not maximum



(b) A maximum matching

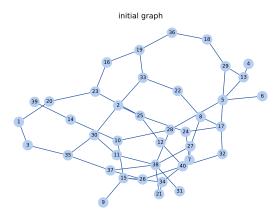
Can you propose a greedy algorithm to address the maximum matching problem ?

```
\begin{aligned} & \text{Result}: \text{ Matching M} \\ & M \leftarrow \emptyset; \\ & \text{for } e \in E \text{ do} \\ & & \text{ if } M \cup \{e\} \text{ is a matching then} \\ & & | M \leftarrow M \cup \{e\} \\ & \text{ end} \\ & \text{end} \\ & \text{return } M \\ & & \text{Algorithme 0}: \text{ Greedy algorithm to find a matching} \end{aligned}
```

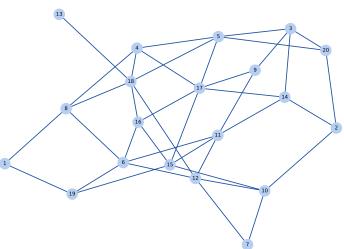
- ▶ What is the type of matching algorithm returned by this algorithm ?
- ▶ What is the complexity of this algorithm ? (as a function of the number of nodes *n* of the graph)

- The greedy algorithm returns a maximal matching (proof)
- Its complexity is smaller than $\mathcal{O}(np)$ (n nodes, p edges) (proof)
- ▶ smaller than **cubic** in the number of nodes : $\mathcal{O}(n^3)$

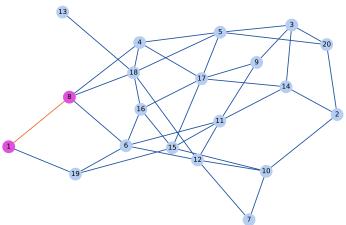
We will implement the greedy algorithm to find a maximal matching. Exercice 6: Implementing the greedy algorithm
Using main_matching_greedy.py, you can generate problem
instances and apply the greedy algorithm by fixing
matching_greedy/greedy_matching.py. The images are stored
in matching_greedy/images/.



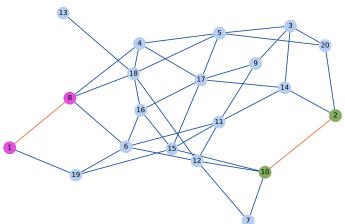
initial graph



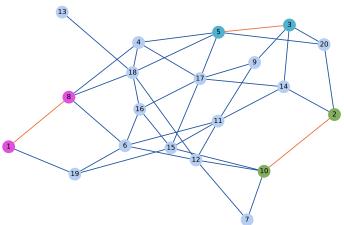
Matching size: 1 Algo step: 1 Nb nodes: 20



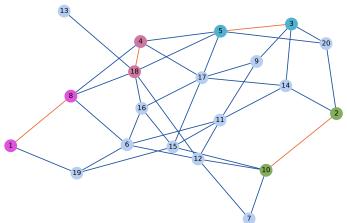
Matching size: 2 Algo step: 3 Nb nodes: 20



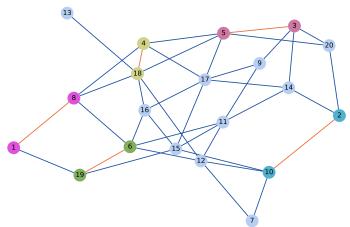
Matching size: 3 Algo step: 6 Nb nodes: 20



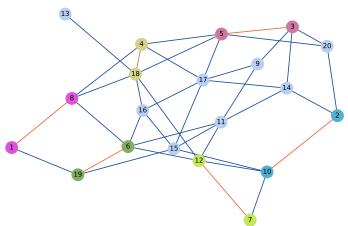
Matching size: 4 Algo step: 11 Nb nodes: 20



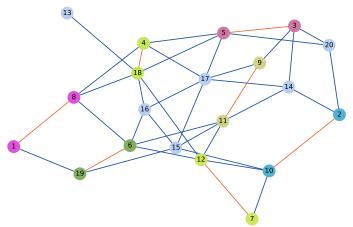
Matching size: 5 Algo step: 17 Nb nodes: 20



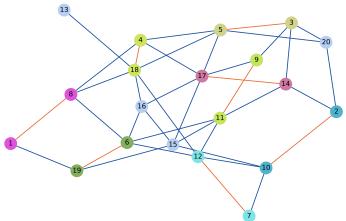
Matching size: 6 Algo step: 22 Nb nodes: 20



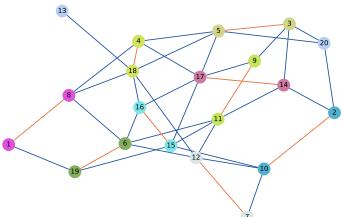
Matching size: 7 Algo step: 25 Nb nodes: 20



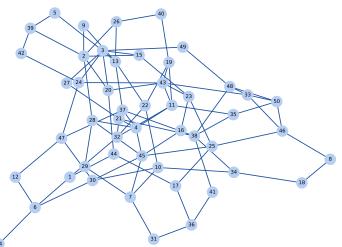
Matching size: 8 Algo step: 34 Nb nodes: 20



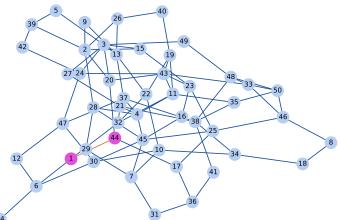
Matching size: 9 Algo step: 36 Nb nodes: 20



initial graph

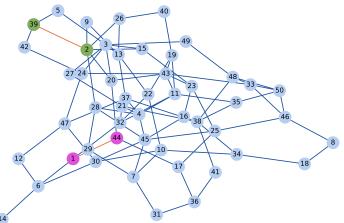


Matching size: 1 Algo step: 1 Nb nodes: 50

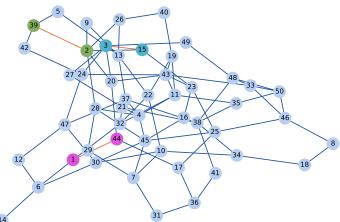


Greedy algorithm

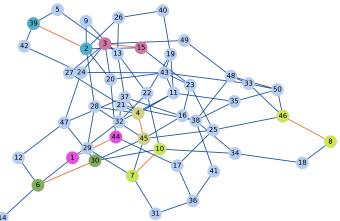
Matching size: 2 Algo step: 4 Nb nodes: 50



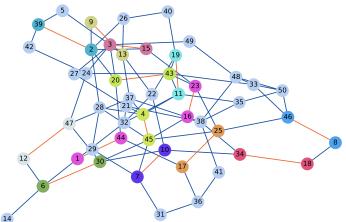
Matching size: 3 Algo step: 10 Nb nodes: 50



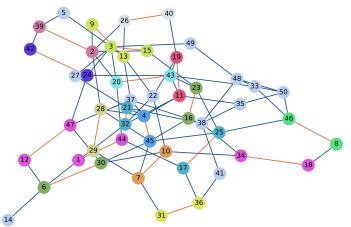
Matching size: 7 Algo step: 30 Nb nodes: 50



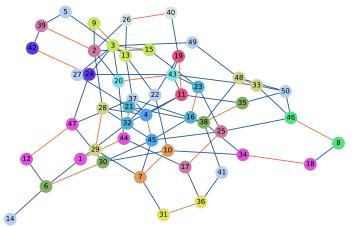
Matching size: 14 Algo step: 54 Nb nodes: 50



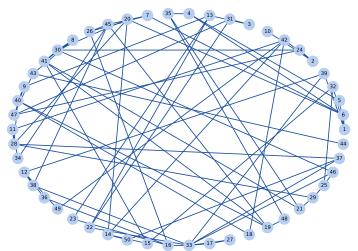
Matching size: 19 Algo step: 72 Nb nodes: 50



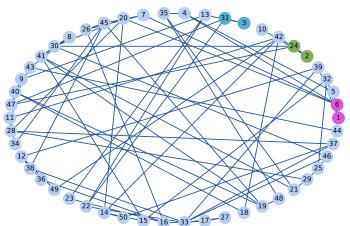
Matching size: 21 Algo step: 78 Nb nodes: 50



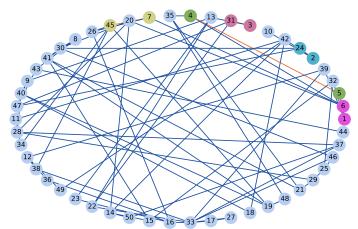
initial graph



Matching size: 3 Algo step: 8 Nb nodes: 50

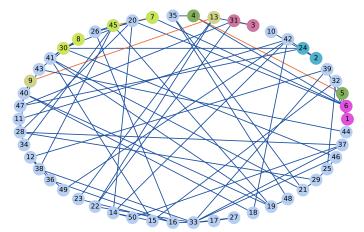


Matching size: 5 Algo step: 15 Nb nodes: 50



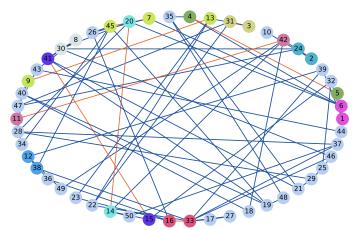
Greedy algorithm

Matching size: 7 Algo step: 20 Nb nodes: 50



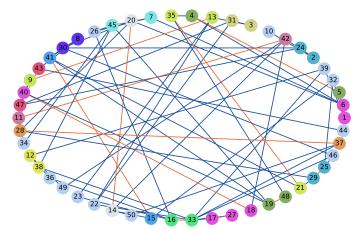
Greedy algorithm

Matching size: 12 Algo step: 38 Nb nodes: 50



Greedy algorithm

Matching size: 19 Algo step: 79 Nb nodes: 50



Example

Exercice 6: Can you think of an example where the greedy algorithm gives a **bad** matching, e.g. of the size **half** the size of an optimal matching?

Example

Exercice 6: Can you think of an example where the greedy algorithm gives a **bad** matching, e.g. of the size **half** the size of an optimal matching?



Greedy matching

However, is |M| is the cardinality of a matching returned by the greedy algorithm, and if $|M^*|$ is the cardinal of the real optimal matching, we can theoretically show that :

$$|M| \ge \frac{|M^*|}{2} \tag{2}$$

Speed comparison as a function of the data structure

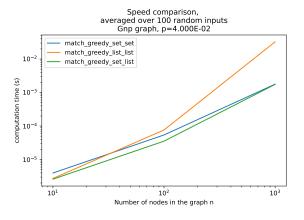


Figure: The functions will be available in code/solutions and shown during the class.

Matchings and vertex covers

Exercice 7: Show that the nodes of the edges selected in a maximal matching form a **vertex cover**. https://en.wikipedia.org/wiki/Vertex_cover

Matchings and vertex covers

Exercice 8: Show that any matching is smaller than any vertex cover.