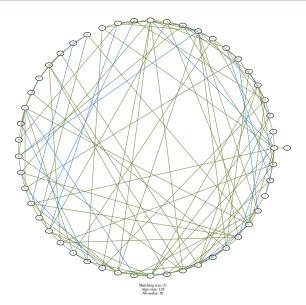
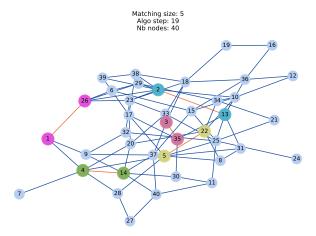
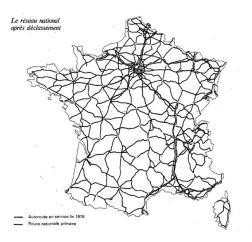
# Algorithmic complexity and graphs: the matching problem

October 1, 2022

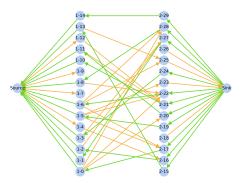
Introduction



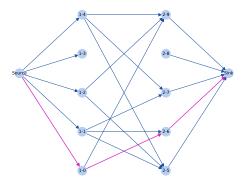




residual graph step 12



augmenting path step 1



### The mathing problem

#### The matching problem

Definition of the problem Experimental solutions Brute force algorithm Greedy algorithm

## Introductory example 1 : Max Flow



Figure: Problem 1: transporting merchandise through a network

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

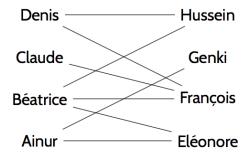


Figure: Problem 2 : Building the largest possible number of teams of 2 persons.

#### Introductory example 2

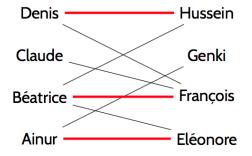


Figure: Problem 2: not optimal assignment

#### Introductory example 2

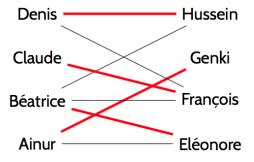


Figure: Problem 2: optimal assignment

## Other examples

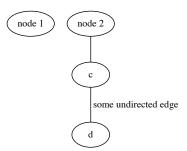
- Assigning students to internships
- Assigning machines to a task

#### Summary

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

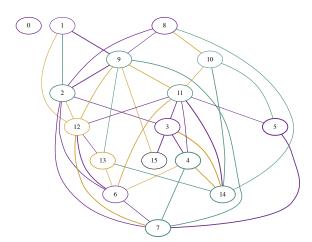
## Reminders on graphs

▶ A graph is defined by set of **vertices** (or **nodes** ) *V* and a set of **edges** *E*.



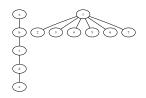
☐ Definition of the problem

#### Reminders on graphs Undirected graph

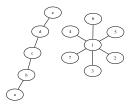


# Other available tool: graphviz

- ► A tool to visualize graphs
- ► Several **generator programs** : dot, neato



(a) Image generated with dot

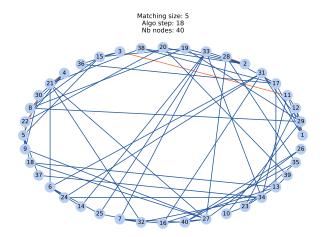


(b) Image generated with neato

- The matching problem
  - Definition of the problem

#### Networkx

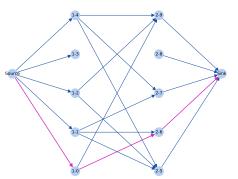
We will use networkx.



Definition of the problem

#### Networkx

#### augmenting path step 1



# Complete graph

Given a **directed** graph with n nodes, the maximum number of edges is:

$$n(n-1) \tag{1}$$

So if the graph is **undirected**, we can build :

$$\frac{n(n-1)}{2} \tag{2}$$

edges.

☐ Definition of the problem

#### Remark

 $\frac{n(n-1)}{2}$  is also the number of subsets of size 2 in a set of size n.

$$\binom{n}{2} = \frac{n!}{2!(n-2)!} \tag{3}$$

# Famous graph problem

Dominating set

# Famous graph problem

- Dominating set
- ► Maximum clique

# Famous graph problem

- ▶ Dominating set
- ► Maximum clique
- Coloring

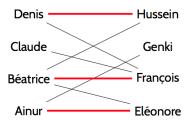
# Matching problem

Let us now focus on the  $matching\ problem$  (problème du couplage )

#### Back to our problem

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

▶ A subset of edges  $M \subset E$ 

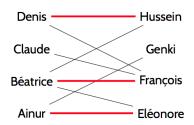


#### Definition of the problem

#### Back to our problem

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

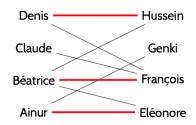
- ▶ A subset of edges  $M \subset E$
- ▶ Such that no pairs of edges of *M* are incident
- Equivalently, each node in the graph is at most in one edge of M.



#### Back to our 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 maximum size in a given graph.

☐ Definition of the problem

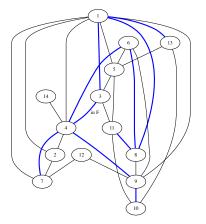


Figure: Is this a matching?

Definition of the problem

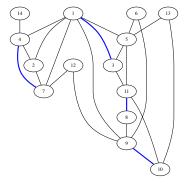


Figure: Is this a matching?

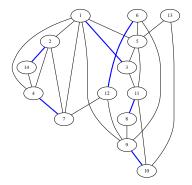


Figure: Is this an optimal matching?

Definition of the problem

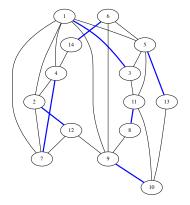


Figure: Is this an optimal matching?

Definition of the problem

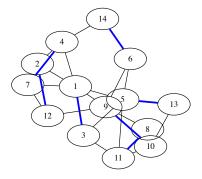


Figure: With neato

# Optimal matching

Exercice 1: Given a graph of size n, what is maximum size possible for a **matching** ?

# Optimal matching

Exercice 1: Given a graph of size n, what is maximum size possible for a matching ?

▶ If *n* is even :  $\frac{n}{2}$ 

▶ Else *n* is odd :  $\frac{n-1}{2}$ 

## Optimal matching

Exercice 1: Given a graph of size n, what is maximum size possible for a matching ?

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

Hence,

$$\lfloor \frac{n}{2} \rfloor$$
 (4)

## Optimal matching

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)

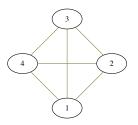


Figure: The complete graph works

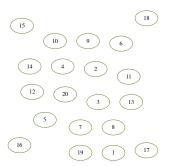
## Optimal matching

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)

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)



## Optimal matching

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)

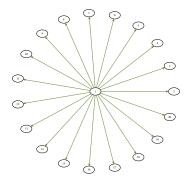


Figure: Star graph

## 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



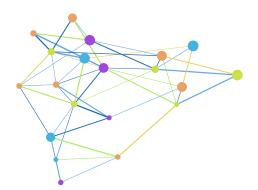
$$g1 = [\{1,2\},\{1,3\},\{2,3\},\{3,4\},\{1,4\}]$$

### Coding a graph: as a dictionary



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

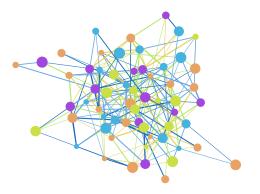
# $Generating\ graphs\ with\ networks.$



Overview

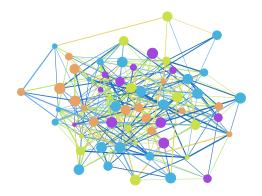
The matching problem

Experimental solutions

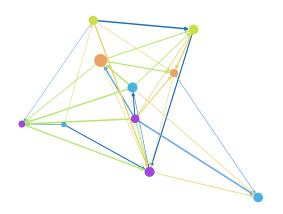


Overview

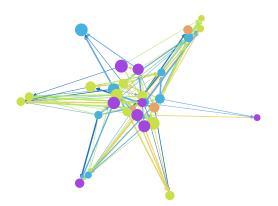
The matching problem
Experimental solutions



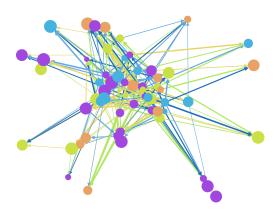
# Directed graph



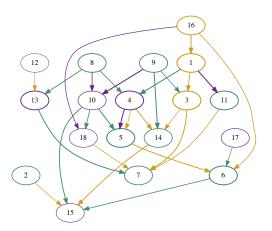
# Directed graph II



# Directed graph III



## Example directed graph

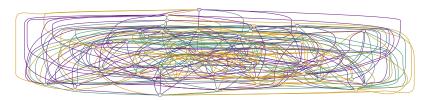


## Manual matching

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 ?
  - ▶ 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 ? 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 : so it is ok.

#### 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 (is 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 (is 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

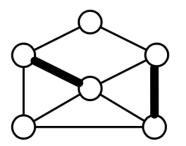
Let us show that a maximum matching is maximal.

### Counter Example

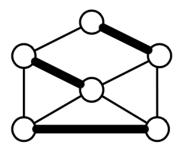
However, a matching that is maximal is not necessary Maximum.

#### Counter Example

However, a matching that is maximal is **not necessary Maximum**. Can you find an example ?



(a) A maximal matching not maximum



(b) A maximum matching

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

```
\label{eq:Result: Matching M} \begin{split} & 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{end} \\ \\ & \text{Algorithme 0 : Greedy algorithm to find a matching} \end{split}
```

- 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)

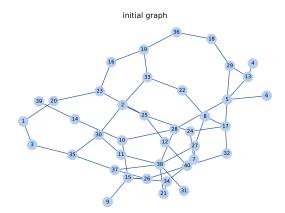
### Access times

https://wiki.python.org/moin/TimeComplexity

- ► 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 : cd matching \_greedy/ and use \_generate \_graph.py to build a graph with a least 30 nodes. The images are stored in images/, data stored in data/

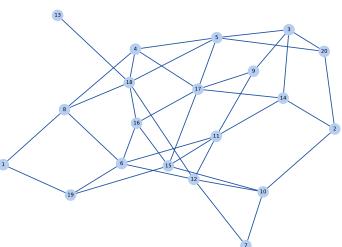


### Implementing the greedy algorithm

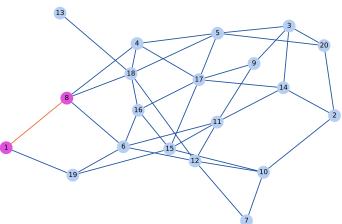
Exercice 6: Implement the greedy algorithm on this graph.

- Use the functions in matching functions.py and call them from apply matching algorithm.py
- More details in the file.

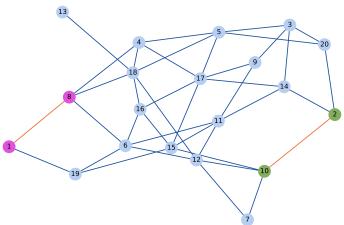
#### initial graph



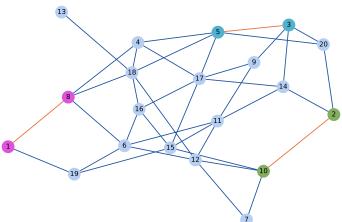
Matching size: 1 Algo step: 1 Nb nodes: 20



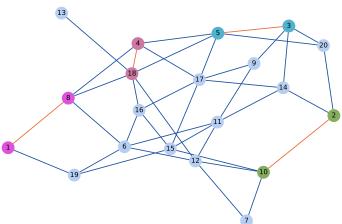
Matching size: 2 Algo step: 3 Nb nodes: 20



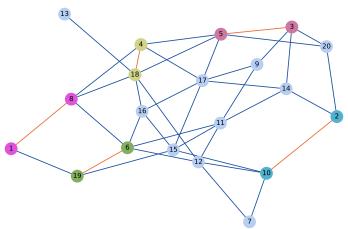
Matching size: 3 Algo step: 6 Nb nodes: 20



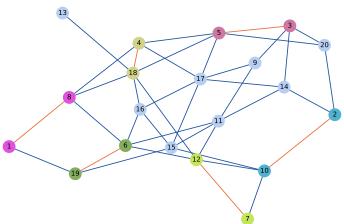
Matching size: 4 Algo step: 11 Nb nodes: 20



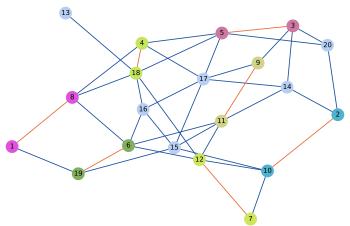
Matching size: 5 Algo step: 17 Nb nodes: 20



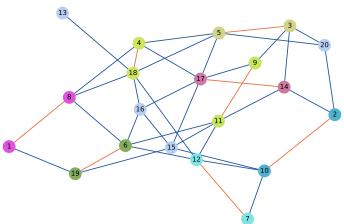
Matching size: 6 Algo step: 22 Nb nodes: 20



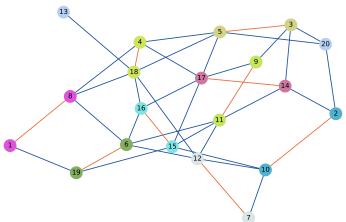
Matching size: 7 Algo step: 25 Nb nodes: 20



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



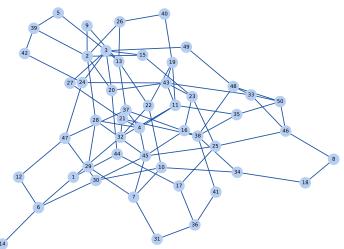
Matching size: 9 Algo step: 36 Nb nodes: 20



The matching problem

Greedy algorithm

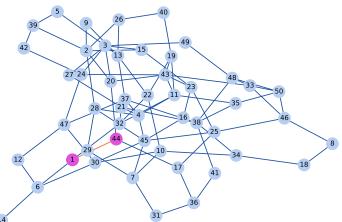
### initial graph



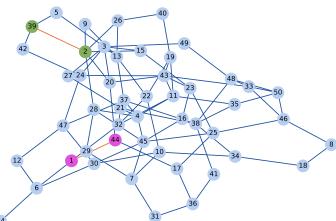
The matching problem

Greedy algorithm

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



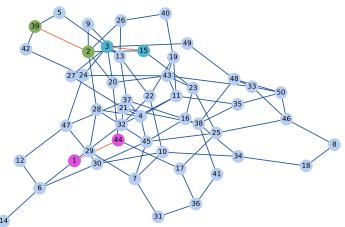
Matching size: 2 Algo step: 4 Nb nodes: 50



The matering proble

Greedy algorithm

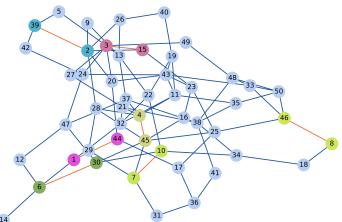
Matching size: 3 Algo step: 10 Nb nodes: 50



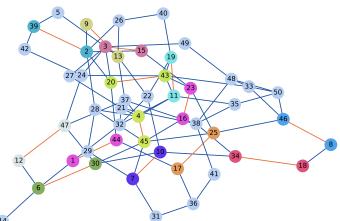
The matching problem

Greedy algorithm

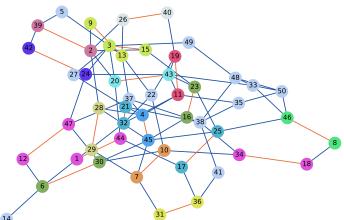
Matching size: 7 Algo step: 30 Nb nodes: 50



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

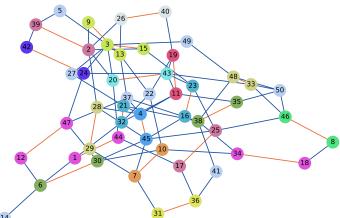




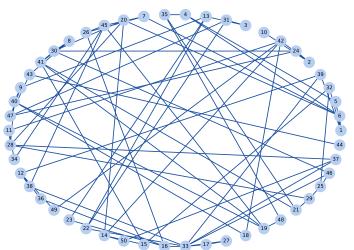


The matching problem





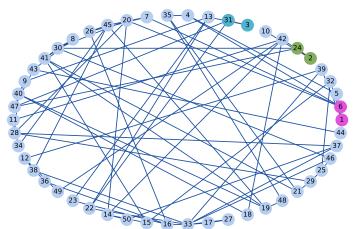
#### initial graph



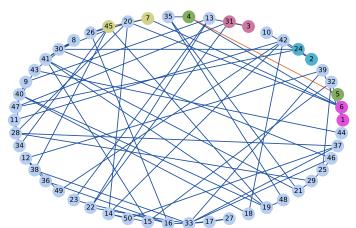
The matching problem

Greedy algorithm

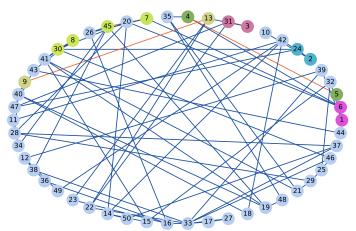
Matching size: 3 Algo step: 8 Nb nodes: 50



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



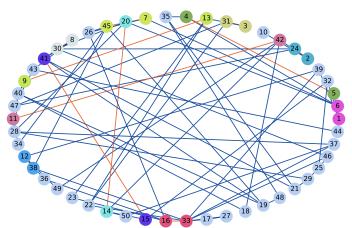
Matching size: 7 Algo step: 20 Nb nodes: 50



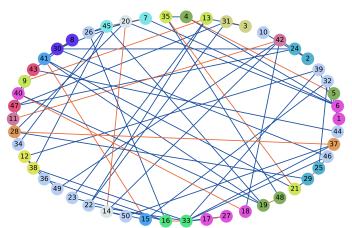
The matching problem

Greedy algorithm

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



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



### Example

Exercice 7: 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?

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Exercice 7: 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 :

$$|M| \ge \frac{|M^*|}{2} \tag{5}$$