

INF421, Lecture 8 Graphs

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ÉCOLE POLYTECHNIQUE

Course

- Objective: to teach you some data structures and associated algorithms
- **Evaluation**: TP noté en salle info le 16 septembre, Contrôle à la fin. Note: $\max(CC, \frac{3}{4}CC + \frac{1}{4}TP)$
- Organization: fri 26/8, 2/9, 9/9, 16/9, 23/9, 30/9, 7/10, 14/10, 21/10, amphi 1030-12 (Arago), TD 1330-1530, 1545-1745 (SI31,32,33,34)

Books:

- 1. Ph. Baptiste & L. Maranget, *Programmation et Algorithmique*, Ecole Polytechnique (Polycopié), 2009
- 2. G. Dowek, Les principes des langages de programmation, Editions de l'X, 2008
- 3. D. Knuth, The Art of Computer Programming, Addison-Wesley, 1997
- 4. K. Mehlhorn & P. Sanders, Algorithms and Data Structures, Springer, 2008
- Website: www.enseignement.polytechnique.fr/informatique/INF421
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Lecture summary

- Graph definitions
- Operations on graphs
- Combinatorial problems on graphs
- Easy and hard problems
- Modelling problems for a generic solution method



The minimal knowledge

- Operations on graphs: complement, line graph, contraction
- Decision/optimization problems: finding subgraphs with given properties
- Easy problems: solvable in polynomial time (P), e.g. minimum cost spanning tree, shortest paths, maximum matching
- Hard problems: efficient method for solving one would solve all of them (NP-hard), e.g. maximum clique, maximum stable set, vertex colouring
- Mathematical Programming: a generic model-and-solve approach



Graph definitions



Motivation

The ultimate data structure

Every time you see arrow connecting boxes, circles or black dots in a computer science course, you can think of graphs and digraphs!

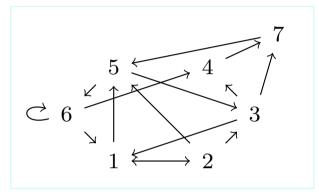


Graphs and digraphs

• Digraph G = (V, A): relation A on set V

ullet V: set of nodes

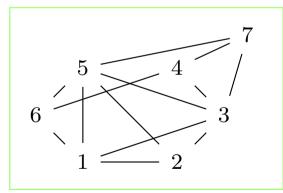
• A: set of arcs (u, v) with $u, v \in V$



• Graph G = (V, E): symmetric relation E on set V

V: set of vertices

• E: set of edges $\{u,v\}$ with $u,v\in V$



Simple (di)graphs: relation is irreflexive

(I.e., v not related to itself for all $v \in V$)



Remarks

- I shall mainly present results for undirected graphs
- Most results extend trivially to directed graphs (digraphs)
- Detailing such extensions is a good exercise
- Warning: not all such extensions are trivial

Example

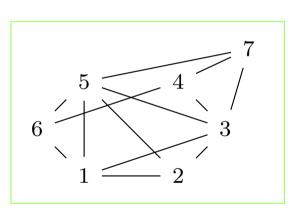
- If G is a graph, we indicate its set of vertices by V(G) and its set of edges by E(G)
- **Example of extension to digraphs:** If G is a digraph, we indicate its set of nodes by V(G) and its set of arcs by A(G)



Stars: sets of nodes/vertices or arcs/edges adjacent to a given node

$$\forall v \in V(G)$$
,

ullet if G is undirected,



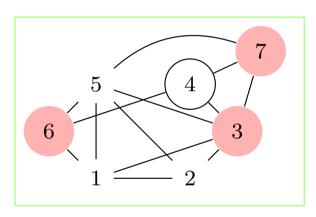


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$$\forall v \in V(G)$$
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•
$$N(v) = \{u \in V \mid \{u, v\} \in E(G)\}$$





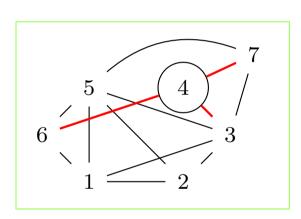
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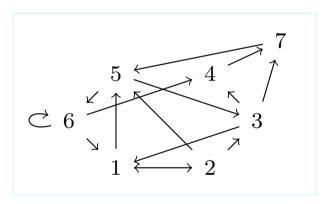
•
$$\delta(v) = \{\{u, v\} \mid u \in N(v)\}$$





$$\forall v \in V(G)$$
,

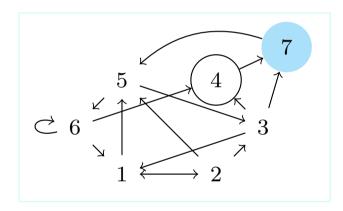
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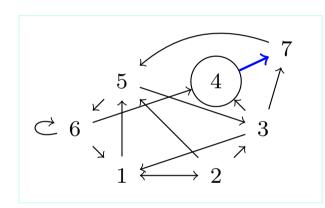
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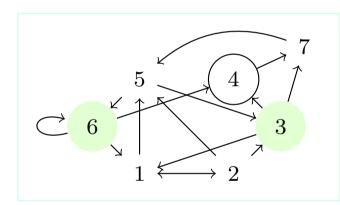
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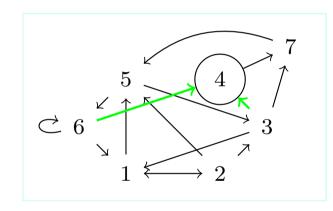
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Stars

Stars: sets of nodes/vertices or arcs/edges adjacent to a given node

$$\forall v \in V(G),$$

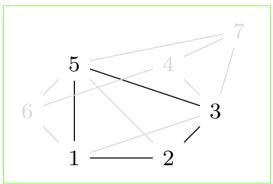
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 - $\delta^-(v) = \{(u, v) \mid u \in N^-(v)\}$
- ightharpoonup |N(v)| =degree, $|N^+(v)| =$ outdegree, $|N^-(v)| =$ indegree of v
- If v belongs to two graphs G,H, write $N_G(v)$ and $N_H(v)$

(similarly for other star notation)

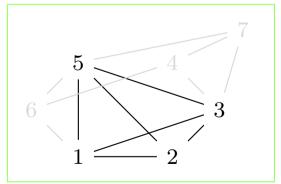


Subgraphs

● A graph H=(U,F) is a subgraph of G=(V,E) if $U\subseteq V, F\subseteq E$ and $\forall \{u,v\}\in F\ (u,v\in U)$



- ullet A subgraph H=(U,F) of G=(V,E) is spanning if U=V

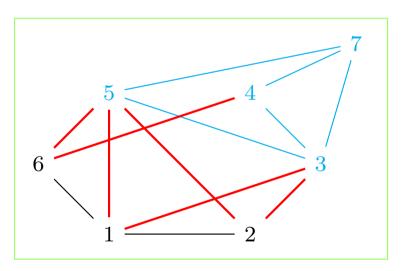


Induced subgraph notation: H = G[U]

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Cutsets

- Let H = (U, F) be a subgraph of G = (V, E) (i.e. $U \subsetneq V$)
- $\hbox{ The cutset } \delta(H) = \left(\bigcup_{u \in U} \delta(u)\right) \smallsetminus F$ is the edge set "separating" U and $V \smallsetminus U$
- E.g. let $U=\{1,2,6\}$ and H=G[U], then $\delta(H)$ is shown by the red edges below

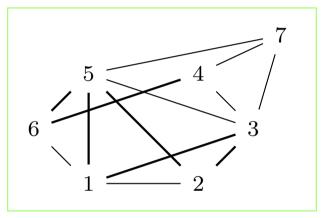


- Similar definitions hold for directed cutsets
- If G is undirected, $\delta(U) = \delta(V \setminus U)$ for all $U \subseteq V(G)$

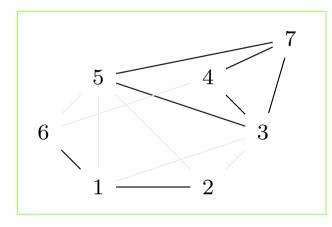


Connectedness

A graph is connected if there are no empty nontrivial cutsets



Connected



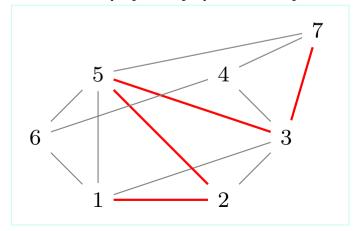
Not connected: $\delta(\{1,2,6\}) = \varnothing$

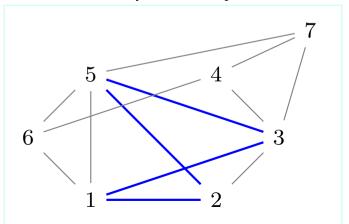
- Each maximal connected subgraph of a graph is a connected component
- Most graph algorithms assume the input graph to be connected: if not, just apply it to each connected component

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Paths and cycles

- ▶ Let G be a graph and $u, v \in V(G)$
- lacksquare A simple path P from u to v in G is a connected subgraph of G s.t.:
 - 1. each vertex w in P different from u, v has |N(w)| = 2
 - **2.** if $u \neq v$ then |N(u)| = |N(v)| = 1
 - 3. if u=v then either $E(P)=\varnothing$ or |N(u)|=|N(v)|=2
- ullet We indicate a path from u to v by the notation $u \to v$
- If P is a path $u \to v$, then u, v are called the endpoints of the path
- A simple cycle is a simple path with equal endpoints
- Definitions in Lecture 6 equivalent but more general
- Will simply say paths/cycles to mean simple paths/cycles

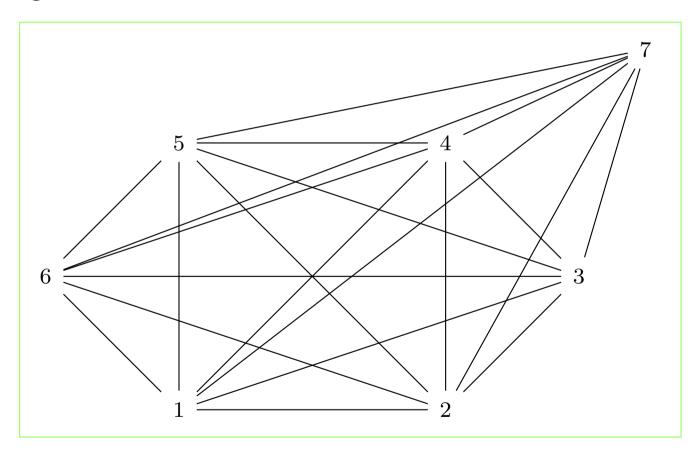






Complete graph

■ The complete graph K_n on n vertices has all possible edges

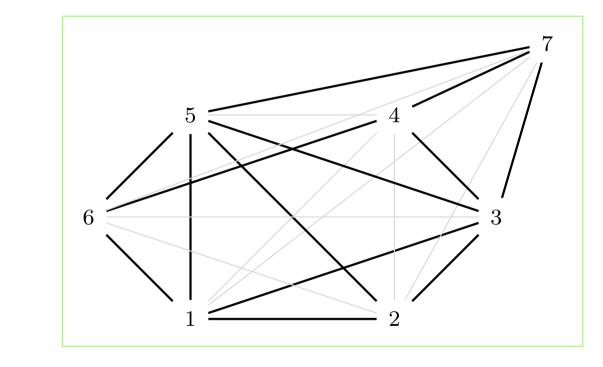


• K_n is also called n-clique; a complete graph on a vertex set U is denoted by K(U)



Complement graph

• Given G=(V,F) with |V|=n, the complement of G is $\bar{G}=(V,E(K_n)\smallsetminus F)$

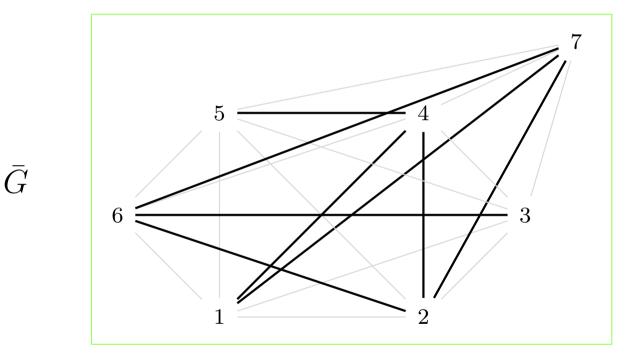


• The complement of K_n is the empty graph (V, \emptyset) on n vertices



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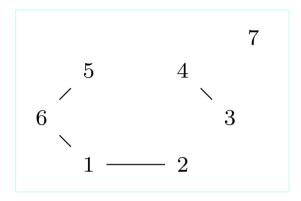


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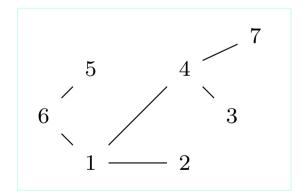


Forests and trees

A forest is a graph with no cycles



A tree is a connected forest

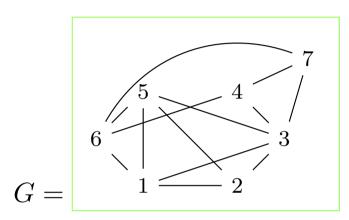


If a tree is a subgraph of another graph G, we also call it a spanning tree

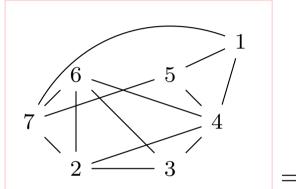


Graph isomorphism

Let |V| = n and S_n be the symmetric group of order n $\pi \in S_n$ acts on V, get new graph πG



$$\pi = (1,2,3,4,5,6,7)$$



$$=H$$

• $\exists \pi \in S_n \ (G = \pi H) \Rightarrow G, H$ isomorphic, π graph isomorphism

Take
$$G = (1, 7, 6, 5, 4, 3, 2)H$$

• If $(\pi G = G)$, then π is an automorphism of G

Automorphism group of G is $\operatorname{Aut}(G) = \langle (1,5), (4,7) \rangle \cong C_2 \times C_2$

$$N(1) = \{2, 3, 5, 6\}, N(2) = \{1, 3, 5\}$$

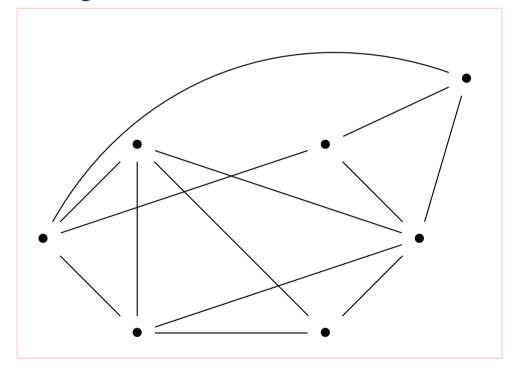
 $N(3) = \{1, 2, 4, 5, 7\}, N(4) = \{3, 6, 7\}$
 $N(5) = \{1, 2, 3, 6\}, N(6) = \{1, 4, 5, 7\}$
 $N(7) = \{3, 4, 6\}$

$$= \begin{cases} N(5) = \{2, 3, 1, 6\}, N(2) = \{5, 3, 1\} \\ N(3) = \{5, 2, 7, 1, 4\}, N(7) = \{3, 6, 4\} \\ N(1) = \{5, 2, 3, 6\}, N(6) = \{5, 7, 1, 4\} \\ N(4) = \{3, 7, 6\} \end{cases}$$



Graphs modulo symmetry

 Symmetries act on vertex labels — can represent equivalence classes of graphs modulo symmetry by simply ignoring labels



Not easy to treat mathematically, though...

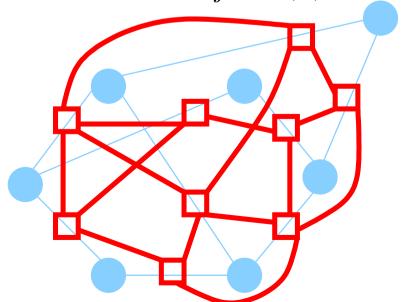


Line graphs

- Given a graph G = (V, E) where $E = \{e_1, \dots, e_m\}$
- ullet The line graph of G is

$$L(G) = (E, \{\{e_i, e_j\} \mid e_i \cap e_j \neq \emptyset\})$$

- Every vertex of L(G) is an edge of G
- Two vertices e_i, e_j of L(G) are adjacent if there is $v \in V$ such that $e_i, e_j \in \delta(v)$



Property: the degree $|N_{L(G)}(e)|$ of a vertex $e=\{u,v\}$ of L(G) is $|N_G(u)|+|N_G(v)|-2$.

L(G) can be constructed from G in polynomial time (how?)



Operations on graphs



Addition and removal

Add a vertex v:

update
$$V \leftarrow V \cup \{v\}$$

• Add an edge $e = \{u, v\}$:

add vertices
$$u, v$$
, update $E \leftarrow E \cup \{e\}$

• Remove an edge $e = \{u, v\}$:

update
$$E \leftarrow E \setminus \{e\}$$

Remove a vertex v:

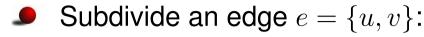
update
$$V \leftarrow V \setminus \{v\}$$
 and $E \leftarrow E \setminus \delta(v)$

Operations on sets of vertices/edges:

Apply operation to each set element



Subdivision and contraction



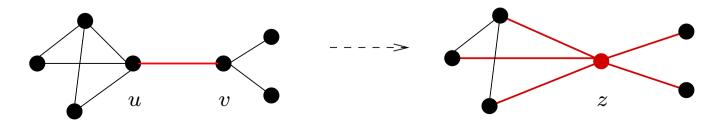
remove e, let $z \notin V$, add edges $\{u, z\}$ and $\{z, v\}$

$$u \bullet v \longrightarrow v \longrightarrow u \bullet v$$

• Contract an edge $e = \{u, v\}$:

contract(G,e):

- 1: Let $N(e) = (N(u) \cup N(v)) \setminus \{u, v\}$
- 2: Let z be a vertex $\notin V$;
- 3: Add vertex z;
- 4: for $v \in N(e)$ do
- 5: Add edge $\{v, z\}$;
- 6: end for
- 7: Remove edge e;





Subgraph contraction

- Let G = (V, E), $U \subseteq V$ and H = G[U]
- **●** The contraction G/U is a sort of "G modulo H"

```
contract(G,U):

1: Let z be a new vertex not in V;

2: Add vertex z;

3: for \{u,v\} \in \delta(H) (assume WLOG u \in U, v \in V \setminus U) do

4: add edge \{v,z\};

5: remove edge \{u,v\};

6: end for

7: remove G[U];

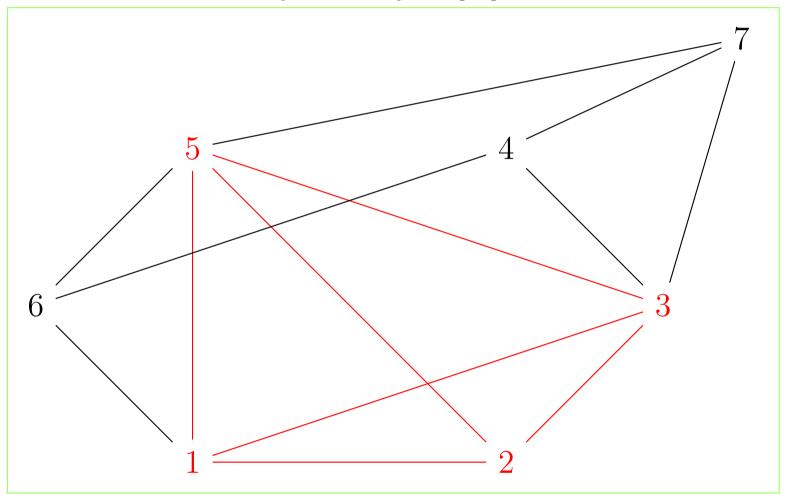
8: return G;
```

- ullet At the end of the contraction algorithm, the whole subgraph H is "replaced" by a single vertex z
- $m{\mathcal{G}}/U$ is formally defined to be $\mathtt{contract}(G,U)$ Thm.

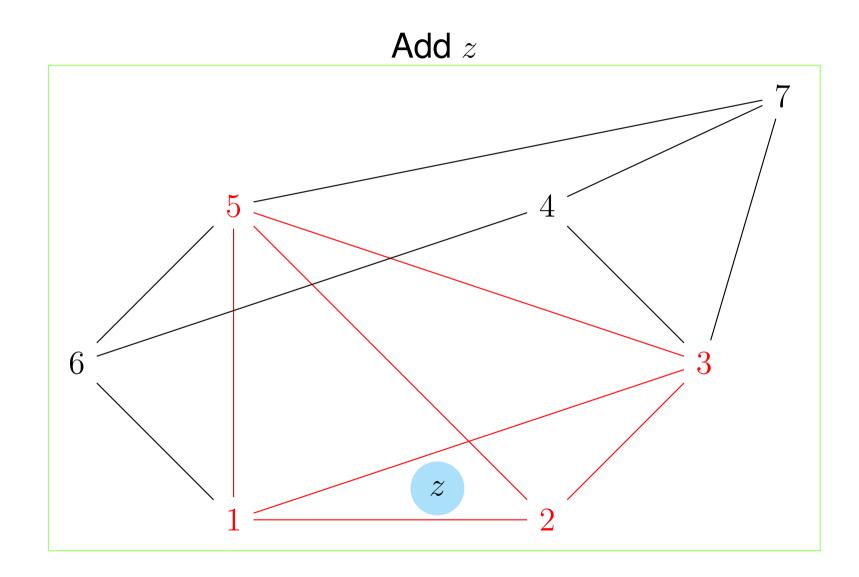
Subgraph contraction is equivalent to a sequence of edge contractions



 $U = \{1, 2, 3, 5\}, G[U]$ in red

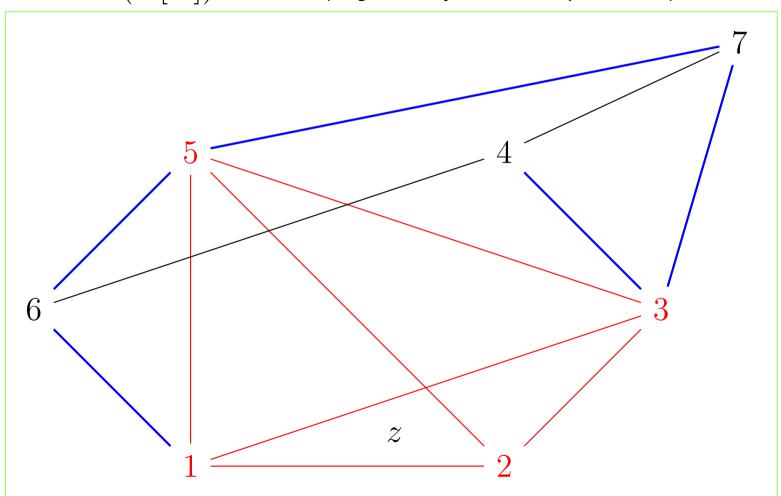






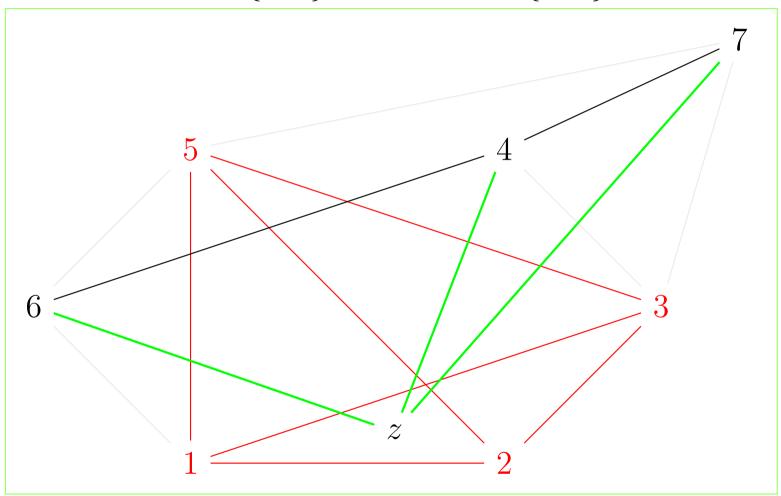


 $\delta(G[U])$ in blue (edges with just one endpoint in U)





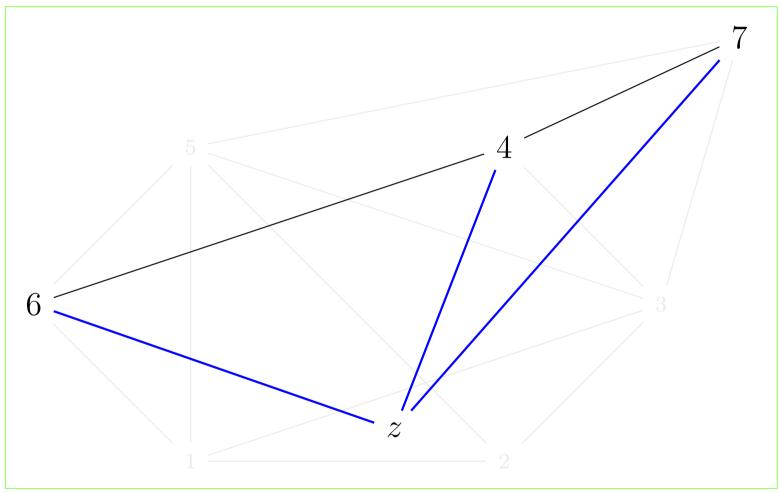
Add $\{v, z\}$ and remove $\{u, v\}$





Subgraph contraction algorithm

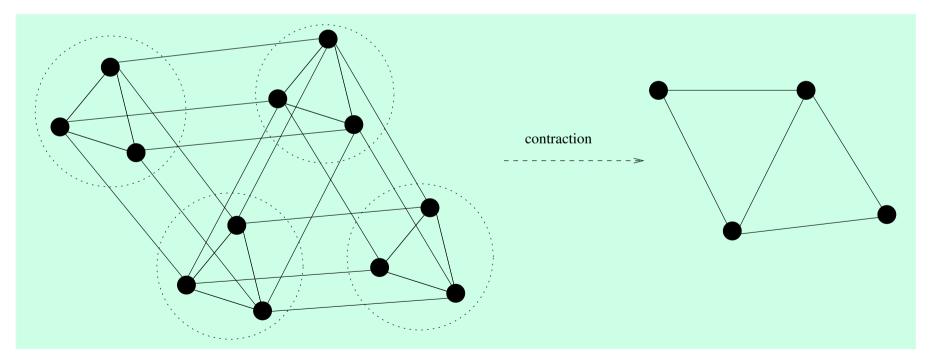
Remove G[U]





Minors

- The graph H is a minor of the graph G if it is isomorphic to the graph obtained by a sequence of contractions
- Useful to underline "essential structure" of a complicated graph



Contract some triangles



Combinatorial problems on graphs



The subgraph problem

- Let G be the class of all graphs
- ▶ For a set of valid propositions P(G) (for $G \in \mathbb{G}$), a typical decision problem in graph theory is the following:
 - Subgraph Problem (SP $_P$). Given a graph G, does it have a subgraph H with property P?
- Decision problem: YES/NO question parametrized over symbols representing the instance (i.e. the input)
- Formally, a decision problem is the set of all possible inputs
- Require decision problems to always provide a certificate (a proof that certifies the answer)
- **9** E.g. if $P(H) \equiv (H \text{ is a cycle})$ the certificate is the cycle
- ▶ NP is the class of decision problems whose certificates for YES instances can be verified in polynomial time w.r.t. the instance size



Graph optimization problems

- To most decision problems on graphs there is a corresponding optimization problem
- Let $\mu: \mathbb{G} \to \mathbb{R}$ be a "measure" function for graphs
- **Σ** E.g. μ could be the number of vertices, or of edges Subgraph Problem, *optimization version* (SP_{P,μ}). Given a graph G, does it have a subgraph H with property P and having minimum/maximum measure μ ?
- Given a property P and a graph measure μ , the set of instances of $SP_{P,\mu}$ is an optimization problem



Easy problems

- We let P be the class of decision or optimization problems which can be solved in polynomial time
- We call problems in P "easy"
 - MINIMUM SPANNING TREE (MST)

Seen in Lecture 6

SHORTEST PATH PROBLEM (SPP) from a vertex v to all other vertices

To be seen in Lecture 9

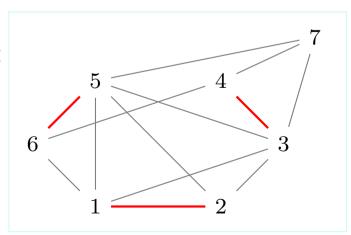
MAXIMUM MATCHING problem (MATCHING)

Discussed in INF550

Matching: subgraph given by set of mutually non-adjacent edges

A maximum matching M ,

$$\mu(M) = |E(M)|$$





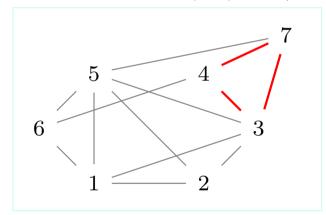
Hard(er) problems



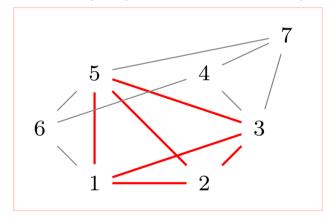
Maximum clique

CLIQUE PROBLEM (CLIQUE). Given a graph G, what is the largest n such that G has K_n as a subgraph?

• In CLIQUE, $\mu(H) = |V(H)|$ and $P(H) \equiv H = K(V(H))$



A clique in G



The largest clique in G

 Several applications to social networks and bioinformatics



Clique and NP-completeness

- The decision version of CLIQUE is:
 - k-СLIQUE PROBLEM (k-CLIQUE). Given a graph G and an integer k > 0, does G have K_k as a subgraph?
- Consider the following result (which we won't prove) Thm.

[Karp 1972] If CLIQUE \in P then P = NP

- Any decision problem for which such a result holds is called NP-complete
- It is not known whether NP-complete problems can be solved in polynomial time; the current guess is NO



Solving NP-complete problems

Essentially, proving NP-completeness of a problem amounts to say "it's really hard"

If it were easy, every problem in **NP** would be easy, which is unlikely: so it's likely to be hard

- Solution methods for NP-complete problems include:
 - exact but exponential worst-case complexity algorithms
 - heuristic algorithms
 whenever they find a YES answer they provide a certificate, but there is no guarantee that they are able to determine whether an answer is NO in a finite amount of time
- Some optimization problems are also such that "solvable in polynomial time" implies P = NP: they are called NP-hard
- With NP-hard optimization problems, can use approximation algorithms their solutions have a μ -value which is no more than f(|G|)-fold worse than the optimal one (e.g. $\mu \leq f(|G|)\mu^*$, where μ^* is the cost of an optimal solution)

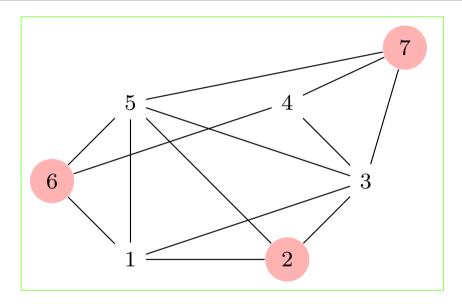


Stables

▶ A stable (or independent set) of a graph G = (V, E) is a subset $U \subseteq V$ such that $\forall u, v \in U \ (\{u, v\} \not\in E)$ Thm.

U is a stable in G if and only if $\bar{G}[U]$ is a clique

a stable in G



Decision problem: k-STABLE

Given G and $k \in \mathbb{N}$, is there a stable $U \subseteq V(G)$ of size k?

Optimization problem: STABLE

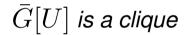
Given G, find the stable of G of maximum size

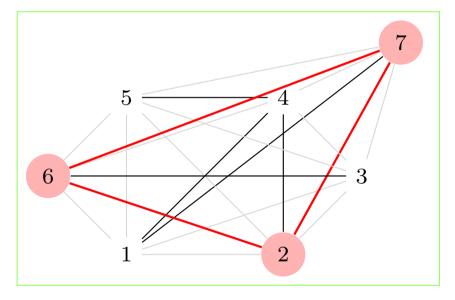


Stables

▶ A stable (or independent set) of a graph G = (V, E) is a subset $U \subseteq V$ such that $\forall u, v \in U \ (\{u, v\} \not\in E)$ Thm.

U is a stable in G if and only if $\bar{G}[U]$ is a clique





Decision problem: k-STABLE

Given G and $k \in \mathbb{N}$, is there a stable $U \subseteq V(G)$ of size k?

Optimization problem: STABLE

Given G, find the stable of G of maximum size



NP-completeness of k-Stable

Thm.

k-Stable is **NP**-complete

Proof

Consider an instance (G, k) of k-CLIQUE

The complement graph \bar{G} can be obtained in polynomial time (*)

It is easy to show that $\bar{\bar{G}} = G \pmod{**}$

By (**) and previous thm.,

(G,k) is a YES instance of k-Clique iff (\bar{G},k) is a YES instance of k-Stable

By (*), if k-Stable \in P then k-Clique \in P

By NP-completeness of k-Clique, k-Stable \in P implies P = NP

Hence k-Stable is **NP**-complete

- **P** How to show that a problem \mathcal{P} is **NP**-complete:
 - $oldsymbol{ iny}$ Take another **NP**-complete problem ${\mathcal Q}$ "similar" to ${\mathcal P}$
 - Transform (with a poly. alg.) an instance of Q to an instance of P
 - Show that transformation preserves the YES/NO property



- It suffices to give an algorithm for STABLE, the one for CLIQUE will follow trivially (why?)
- The following greedy method will find a maximal stable

```
1: U = \varnothing;

2: order V by increasing values of |N(v)|;

3: while V \neq \varnothing do

4: v = \min V;

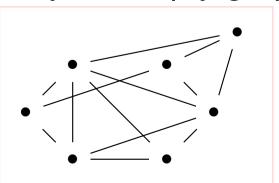
5: U \leftarrow U \cup \{v\};

6: V \leftarrow V \setminus (\{v\} \cup N(v))

7: end while
```

ullet Worst-case: O(n) (given by an empty graph)

degree sequence (3, 3, 3, 3, 4, 5, 5)





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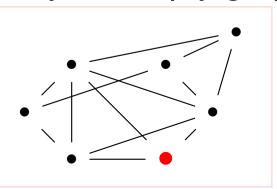
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ullet Worst-case: O(n) (given by an empty graph)

select $\min V$ put it in U





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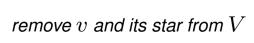
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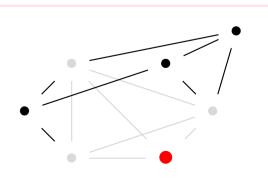
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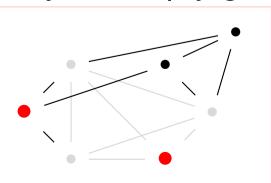
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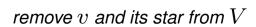
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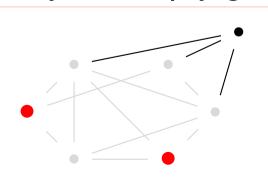
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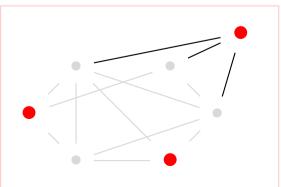
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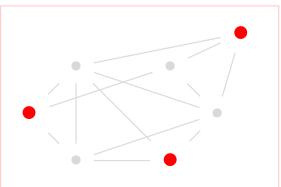
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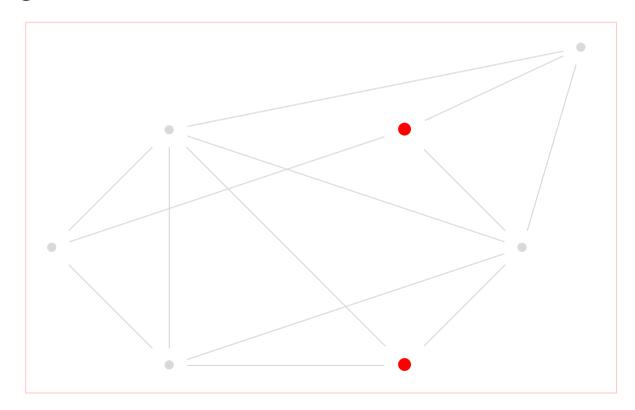
remove v and its star from V stop: maximal stable





Heuristic fails

- The above algorithm may fail to find a <u>maximum</u> stable
- When choosing second element of U, instead of choosing leftmost vertex, choose:



Then algorithm stops immediately with a stable of cardinality 2

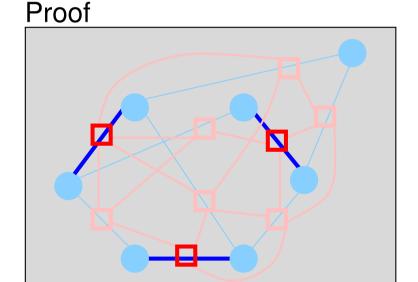


A polynomial case

- Not all instances of an NP-complete problem are hard
- Let P be a decision problem and C⊆P be an infinite set of instances for which there exists a polynomial algorithm
- **●** Then $C \in P$, and C is a polynomial case of P
- **▶** For example, let $\mathcal{L} = \{H \in \mathbb{G} \mid \exists G \in \mathbb{G} \ (H = L(G))\}$ be the class of graphs that are line graphs of another graph

Thm.

A maximum matching in G is a stable in L(G)



Since Matching \in P and finding L(G) can be done in polynomial time, Stable \in P



Vertex colouring

Decision problem

VERTEX k-COLOURING PROBLEM (k-VCP). Given a graph G=(V,E) and an integer k>0, find a function $c:V\to\{1,\ldots,k\}$ such that $\forall \{u,v\}\in E\; (c(u)\neq c(v))$

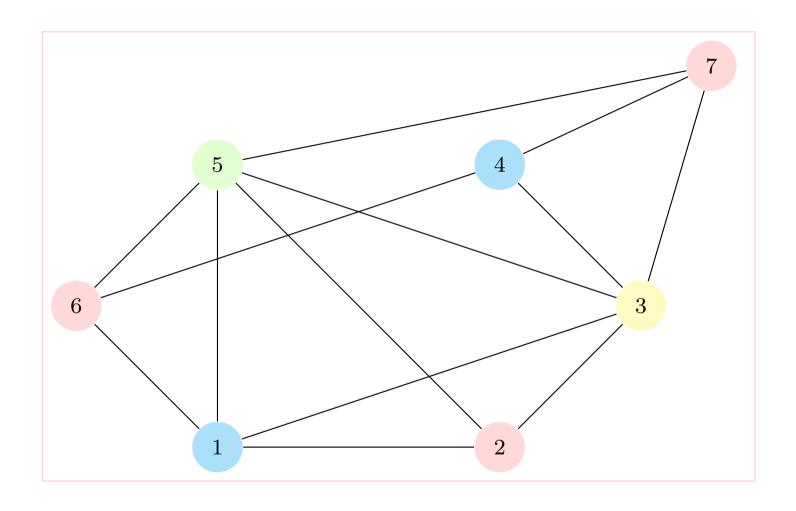
Optimization problem

VERTEX COLOURING PROBLEM (VCP). Given a graph G=(V,E), find the minimum $k\in\mathbb{N}$ such that there is a function $c:V\to\{1,\ldots,k\}$ with $\forall\{u,v\}\in E\ (c(u)\neq c(v))$

- Applications to scheduling and wireless networks
- In general, find how to allocate network resources to a few capacities such that there is no conflict



Vertex colouring example





Vertex colouring heuristic

Thm.

Each color set
$$C_k = \{v \in V \mid c(v) = k\}$$
 is a stable

Use stable set heuristic as a sub-step

```
1: k=1;

2: U=V;

3: while U \neq \emptyset do

4: C_k = \text{maximalStable}(G[U]);

5: U \leftarrow U \setminus C_k;

6: k \leftarrow k+1;

7: end while
```

• Worst-case: O(n) (given by an empty or complete graph)



Model-and-solve



Mathematical programming

- Take e.g. the STABLE problem
- Input (also called parameters):
 - set of vertices V
 - ullet set of edges E
- Output: $x: V \to \{0, 1\}$

$$\forall v \in V \quad x(v) = \left\{ \begin{array}{ll} 1 & \text{if } v \in \text{maximum stable} \\ 0 & \text{otherwise} \end{array} \right.$$

- We also write $x_v = x(v)$
- We'd like $x = (x_v \mid v \in V) \in \{0, 1\}^{|V|}$ to be the characteristic vector of the maximum stable S^*
- $x_1, \ldots, x_{|V|}$ are also called decision variables



Objective function

- If we take x=(0,0,0,0,0,0,0), $S^*=\varnothing$ and $|S^*|=0$ (minimum possible value)
- If we take $x=(1,1,1,1,1,1,1)=\mathbf{1},\,|S^*|=|V|=7$ has the maximum possible value
- Characteristic vector x should satisfy the objective function

$$\max_{x} \sum_{v \in V} x_v$$

ÉCOLE POLYTECHNIQUE

Constraints

- Consider the solution x = 1
- x certainly maximizes the objective
- ... but $S^* = V$ is not a stable!

x=1 is an infeasible solution

- The feasible set is the set of all vectors in $\{0,1\}^{|V|}$ which encode stable sets
- Defining property of a stable:

Two adjacent vertices cannot both belong to the stable

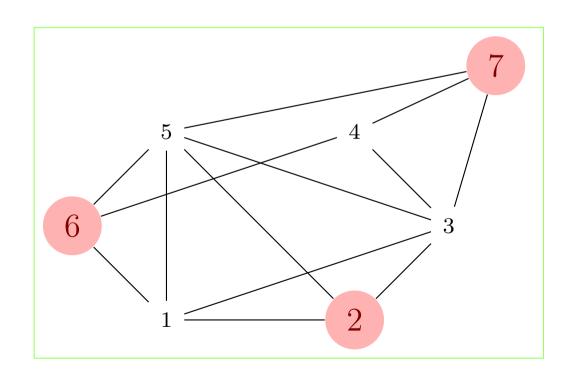
- In other words,
 choose at most one vertex adjacent to each edge
- Written formally,

$$\forall \{u, v\} \in E \quad x_u + x_v \le 1$$



Verify the constraints

- x = (0, 1, 0, 0, 0, 0, 1, 1) encodes $S^* = \{2, 6, 7\}$
- $x_u + x_v = 2$ only for $\{u, v\} \in F = \{\{2, 6\}, \{2, 7\}, \{6, 7\}\}$
- Notice $F \cap E = \emptyset$
- ▶ Hence, $x_u + x_v \le 1$ for all $\{u, v\} \in E$





So what?

OK, so the Mathematical Programming (MP) formulation

$$\max_{v \in V} \sum_{v \in V} x_v$$

$$\forall \{u, v\} \in E \quad x_u + x_v \leq 1$$

$$x \in \{0, 1\}^{|V|}$$

describes Stable correctly

As long as we can't solve it, why should we care?



The magical method

- But WE CAN!
- Use generic MP solvers
- These algorithms can solve ANY MP formulation expressed with linear forms, or prove that there is no solution
- Based on Branch-and-Bound (BB)
- The YES certificate is the characteristic vector of a feasible solution
- The NO certificate is the whole BB tree, which implicitly (and intelligently) enumerates the feasible set
- YES certificate lengths are polynomial, NO certificates may have exponential length



CLIQUE and MATCHING

Clique (use complement graph):

$$\max_{x} \sum_{v \in V} x_{v}$$

$$\forall \{u, v\} \notin E, u \neq v \quad x_{u} + x_{v} \leq 1$$

$$x \in \{0, 1\}^{|V|}$$

Matching:

$$\max_{x} \sum_{\{u,v\} \in E} x_{uv}$$

$$\forall u \in V \quad \sum_{v \in N(u)} x_{uv} \leq 1$$

$$x \in \{0,1\}^{|E|}$$

Warning: although MATCHING∈P, solving the MP formulation with BB is exponential-time



How to

- Come see me, I'll give you a personal demo
- Go to www.ampl.com and download the AMPL software, student version
- AMPL is for modelling, i.e. writing MP formulations
- Still from www.ampl.com, you can download a student version of the ILOG CPLEX BB implementation



And tomorrow?

If you're interested in modelling problems as MPs

- M1:
 - INF572 (Optimization: Modelling and Software)
 - MAP557 (Optimization: Theory and Applications)
- M2:
 - MPRO (Master Parisien en Recherche Operationnelle)

http://uma.ensta-paristech.fr/mpro/



End of Lecture 8