

Graph Algorithms - Lecture 10

December 13, 2024

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Exercises for the 11th seminar (december 16 - 20 week)

Minimum cost flows

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Suppose that in the flow network $R = (G, s, t, c)$, an additional **cost function** is given: $a : E \rightarrow \mathbb{R}; \forall ij \in E, a(ij) = a_{ij}$ is the cost of arc ij (interpreted as the cost of sending of an "unit" of flow on the arc ij). If x is a flow in R , then the **cost of x** is

$$a(x) = \sum_{i,j} a_{ij} x_{ij}.$$

Minimum cost flow problem

Given: R a flow network, $a : E \rightarrow \mathbb{R}$ a cost function, and $v \in \mathbb{R}_+$,
Find: a flow x^0 in R such that

$$a(x^0) = \min \{ a(x) : x \text{ flow in } R, v(x) = v \}.$$

Note that, if v is not greater than the maximum flow value in R , then the problem has always solutions ($a(x)$ is a linear function defined on the non-empty compact set in \mathbb{R}^{m+1} of all flows of value v).

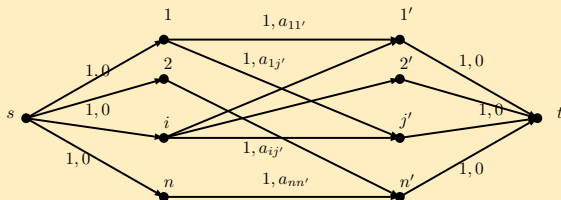
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Minimum cost flows - Examples

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1. Assignment Problem. There are n workers and n jobs. The cost of assigning worker i to job j is a_{ij} . Assign each worker to a job such that the total cost is minimized.

Let us consider the bipartite network flow depicted below, where each arc is labeled with its capacity followed by its cost. Hence, $c_{ij'} = 1$, $c_{si} = 1$, $a_{si} = 0$, $c_{j't} = 1$, and $a_{j't} = 0$, $\forall i, j \in \{1, 2, \dots, n\}$.

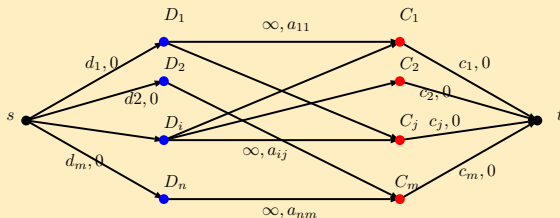


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Minimum cost flows

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The problem has a solution only if $\sum_{i=1}^n d_i \geq \sum_{j=1}^m c_j$. In this case, a minimum cost flow of value $v = \sum_{i=1}^m c_i$, in the network flow below, solves the problem.



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Definition

Let x be a flow in $R = (G, s, t, c)$ and $a : E \rightarrow \mathbb{R}$ a cost function.

- If P is an A -path in R w.r.t. x , then the **cost of the path** P is defined as

$$a(P) = \sum_{ij \in E(P), ij \text{ forward}} a_{ij} - \sum_{ij \in E(P), ji \text{ backward}} a_{ji}$$

- If C is a closed A -path in R w.r.t. x , then $a(C)$ is computed using the above equation, after establishing a direction of traversal of C (it is possible that C is an A -path w.r.t. x in both directions).

Remarks

- If P is an augmenting path w.r.t. x , then $x^1 = x \otimes r(P)$ is a flow of value $v(x^1) = v(x) + r(P)$ and cost $a(x^1) = a(x) + r(P) \cdot a(P)$.

Minimum cost flows

Remarks

- If C is a closed A -path in R w.r.t. x , then $x^1 = x \otimes r(C)$ is a flow of value $v(x^1) = v(x)$ and cost $a(x^1) = a(x) + r(C) \cdot a(C)$. It follows that if $a(C) < 0$ then x^1 is a flow of the same value as x but of cost $a(x^1) < a(x)$.

Theorem 1

A flow x of value v is a minimum cost flow if and only if there is no negative cost closed A -path w.r.t. x in R .

Proof. " \Rightarrow " It follows from the above remark.

" \Leftarrow " Let x be a flow (of value v) such that there is no negative cost closed A -path w.r.t. x in R . Let x^* be a minimum cost flow of value v such that

$$\Delta(x, x^*) = \min \{ \Delta(x, x') : x' \text{ minimum cost flow of value } v \},$$

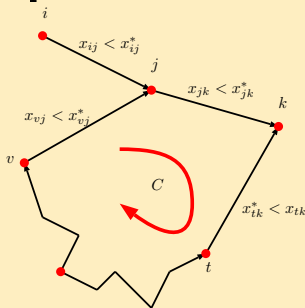
where $\Delta(x, x') = |\{ij \in E : x_{ij} \neq x'_{ij}\}|$.

Minimum cost flows

If $\Delta(x, x^*) = 0$, then $x = x^*$, hence x is a minimum cost flow.

Otherwise, $\Delta(x, x^*) > 0$ and there is ij such that $x_{ij} \neq x_{ij}^*$. Suppose $0 \leq x_{ij} < x_{ij}^* \leq c_{ij}$ (a similar argument works if $x_{ij} > x_{ij}^*$). By the flow conservation law, there exists an arc $jk \in E$ such that $0 \leq x_{jk} < x_{jk}^* \leq c_{jk}$, or there is $kj \in E$ such that $0 \leq x_{kj}^* < x_{kj} \leq c_{kj}$.

Since the number of vertices is finite, by repeating this argument, we construct, C , a closed A -path w.r.t. x in R :



If we traverse C in the converse direction, we obtain a closed A -path, C' , w.r.t. x^* . Since $a(C) \geq 0$ (by the hypothesis), and $a(C') = -a(C)$ it follows that $a(C) = 0$. (x^* is of minimum cost; hence, by the first part of the proof, $a(C') \geq 0$).

If we consider $x' = x^* \otimes \delta(C')$, where

$$\delta(C') = \min \left\{ \min_{kj \text{ forward in } C'} (x_{kj} - x_{kj}^*), \min_{kj \text{ backward in } C'} (x_{jk}^* - x_{jk}) \right\},$$

then x' satisfies $v(x') = v(x^*) = v$, $a(x') = a(x^*) + \delta(C') \cdot a(C') = a(x^*)$.

Hence x' is a minimum cost flow of value v , but $\Delta(x, x') < \Delta(x, x^*)$, contradicting the choice of x^* . Thus $\Delta(x, x^*) = 0$, and the theorem is proved. \square

Theorem 2

If x is a minimum cost flow of value v and P_0 is an augmenting path w.r.t. x such that

$$a(P_0) = \min \{a(P) : P \text{ augmenting path w.r.t. } x\},$$

then $x^1 = x \otimes r(P_0)$ is a minimum cost flow of value $v(x^1) = v + r(P_0)$.

Proof. Omitted.

An augmenting path of minimum cost can be found using shortest paths algorithms. If x is a flow in R and $a : E \rightarrow \mathbb{R}$ is the cost function, then taking $a_{ij} = \infty$ if $ij \notin E$ (when $x_{ij} = 0$), we define

Minimum cost flows

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$$\bar{a}_{ij} = \begin{cases} a_{ij}, & \text{if } x_{ij} < c_{ij} \text{ and } x_{ji} = 0, \\ \min \{ a_{ij}, -a_{ji} \}, & \text{if } x_{ij} < c_{ij} \text{ and } x_{ji} > 0, \\ -a_{ji}, & \text{if } x_{ij} = c_{ij} \text{ and } x_{ji} > 0, \\ +\infty, & \text{if } x_{ij} = c_{ij} \text{ and } x_{ji} = 0. \end{cases}$$

A shortest \bar{a} st -path corresponds to a minimum cost augmenting path w.r.t. x in R , and a negative a cycle corresponds to a negative cost closed A -path w.r.t. x in R .

Then, we have the following algorithm to solve the minimum cost flow problem:

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Minimum cost flows - A generic algorithm (Klein, Busacker, Gowan etc.)

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let  $x$  be a flow of value  $v' \leq v$ ;  
//  $x$  could be null or  $x = (v/v(y)) \cdot y$ ,  
// where  $y$  is a flow of maximum value.  
while ( $\exists C$  an  $\bar{a}$  negative cycle) do  
     $x \leftarrow x \otimes r(C)$ ;  
end while  
while ( $v(x) < v$ ) do  
    find an shortest  $\bar{a}$   $st$ -path  $P$ ;  
     $x \leftarrow x \otimes \min\{r(P), v - v(x)\}$ ;  
end while
```

The time complexity of the second **while** is $\mathcal{O}(n^3 v)$ (if we start with the null flow and if we have integer capacities only). The first **while** can be implemented to run in $\mathcal{O}(nm^2 \log n)$ iterations.

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Polynomial-time reductions for graph problems - Reminder

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Reminder

- Let $P_i : I_i \rightarrow \{yes, no\}$ ($i \in \{1, 2\}$) be two **decision problems**.
 P_1 **polynomial reduces to** P_2 , and we denote this by $P_1 \leq_P P_2$, if there is a **polynomial-time computable function** $\Phi : I_1 \rightarrow I_2$, such that $P_1(i) = P_2(\Phi(i))$, $\forall i \in I_1$.
- The function Φ will be given using an algorithm that constructs, for every instance $i_1 \in I_1$, an instance $i_2 \in I_2$ in polynomial time (in the size of i_1), such that $P_1(i_1) = yes$ if and only if $P_2(i_2) = yes$.
- The construction behind a polynomial-time reduction shows how the first problem can be efficiently solved using an oracle for the second one.

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Polynomial-time reductions for graph problems - Reminder

- The relation \leq_P is a transitive relation on the set of decision problems (since, the class of polynomial functions is closed to function composition).

Example

$\text{SAT} \leq_P \text{3SAT}$

SAT

Instance: $U = \{u_1, u_2, \dots, u_n\}$ a finite set of boolean variables;

$C = C_1 \wedge C_2 \dots \wedge C_m$ a CNF formula over U :

$C_i = v_{i_1} \vee v_{i_2} \vee \dots \vee v_{i_{k_i}}, i = \overline{1, m}$, where

$\forall i_j, \exists \alpha \in \{1, 2, \dots, n\}$ s. t. $v_{i_j} = u_\alpha$ or $v_{i_j} = \overline{u_\alpha}$.

Question: Is there an assignment $t : U \rightarrow \{\text{true}, \text{false}\}$ s. t. $t(C) = \text{true}$?

Polynomial-time reductions for graph problems - Reminder

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3SAT is the restriction of SAT to the set of instances in which each clause C_i has exactly 3 literals ($k_i = 3$), where a literal, v_i , as described above, is a variable or its negation.

The SAT problem is famous since it was the first discovered NP-complete problem (Cook, 1971).

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$$\text{NP} \neq \text{NP} \cap \text{coNP} = \text{P}$$

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Polynomial-time reductions - Maximum stable problem

SM

Instance: $G = (V, E)$ a graph and $k \in \mathbb{N}$.

Question: Is there a stable set S in G such that $|S| \geq k$?

Theorem 3

(Karp, 1972). $3\text{SAT} \leq_P \text{SM}$.

Proof. Let $U = \{u_1, u_2, \dots, u_n\}$, ($n \in \mathbb{N}^*$), $C = C_1 \wedge C_2 \dots \wedge C_m$ ($m \in \mathbb{N}^*$) with $C_i = v_{i1} \vee v_{i2} \vee v_{i3}$, $i = \overline{1, m}$, (where $\forall i_j, \exists \alpha \in \{1, 2, \dots, n\}$ s. t. $v_{ij} = u_\alpha$ or $v_{ij} = \overline{u_\alpha}$) representing the data of an instance of the 3SAT problem.

We will build in $\mathcal{O}(n + m)$ time complexity, a graph G and $k \in \mathbb{N}$, such that there is a satisfying assignment t for C if and only if there is a stable set S in G such that $|S| \geq k$.

Proof cont'd.

Construction of the graph G :

- $\forall i \in \{1, 2, \dots, n\}$, let the disjoint graphs $T_i = (\{u_i, \bar{u}_i\}, \{u_i \bar{u}_i\})$.
- $\forall j \in \{1, 2, \dots, m\}$, let the disjoint graphs $Z_j = (\{a_{j1}, a_{j2}, a_{j3}\}, \{a_{j1} a_{j2}, a_{j2} a_{j3}, a_{j3} a_{j1}\})$.
- $\forall j \in \{1, 2, \dots, m\}$, let the edge-set $E_j = \{a_{j1} v_{j1}, a_{j2} v_{j2}, a_{j3} v_{j3}\}$, where $v_{j1} \vee v_{j2} \vee v_{j3}$ is the clause C_j .

$$V(G) = \left(\bigcup_{i=1}^n V(T_i) \right) \cup \left(\bigcup_{j=1}^m V(Z_j) \right)$$

$$E(G) = \left(\bigcup_{i=1}^n E(T_i) \right) \cup \left(\bigcup_{j=1}^m E(Z_j) \cup E_j \right).$$

Polynomial-time reductions - Maximum stable problem

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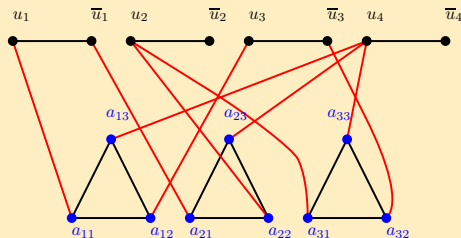
Clearly, the construction of G can be done in polynomial time w.r.t. the size $n + m$ of the 3SAT instance. Let $k = n + m$.

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Example

$U = \{u_1, u_2, u_3, u_4\}$; $C = (u_1 \vee u_3 \vee u_4) \wedge (\bar{u}_1 \vee u_2 \vee u_4) \wedge (u_2 \vee \bar{u}_3 \vee u_4)$;
 $k = 4 + 3 = 7$.

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Polynomial-time reductions - Maximum stable problem

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Suppose that the answer of the problem SM for the above (G, k) instance is yes.

So, $\exists S \in \mathcal{S}_G$ (the family of all stable sets in G) such that $|S| \geq k$. Since any stable set can have at most one vertex from each $V(T_i)$ and each $V(Z_j)$, it follows that $|S| = k$ and $|S \cap V(T_i)| = 1$, $|S \cap V(Z_j)| = 1$, $\forall i = \overline{1, n}$, $\forall j = \overline{1, m}$.

Let $t : U \rightarrow \{true, false\}$ given by

$$t(u_i) = \begin{cases} true, & \text{if } S \cap V(T_i) = \{\overline{u_i}\} \\ false, & \text{if } S \cap V(T_i) = \{u_i\} \end{cases}$$

Then, $t(C_j) = true$, $\forall j = \overline{1, m}$, hence $t(C) = true$, and the answer of 3SAT is yes.

Indeed, $\forall j = \overline{1, m}$, if $C_j = v_{j1} \vee v_{j2} \vee v_{j3}$ and $S \cap V(Z_j) = \{a_{jk}\}$ ($k \in \{1, 2, 3\}$), then (since $a_{jk}v_{jk} \in E$) it follows that $v_{jk} \notin S$.

Polynomial-time reductions - Maximum stable problem

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- If $v_{jk} = u_\alpha$, then $u_\alpha \notin S$, so $\bar{u}_\alpha \in S$, and - by the definition of t - we have $t(u_\alpha) = \text{true}$, that is, $t(v_{jk}) = \text{true}$, which implies $t(C_j) = \text{true}$.
- If $v_{jk} = \bar{u}_\alpha$, then $\bar{u}_\alpha \notin S$, so $u_\alpha \in S$, and - by the definition of t - we have $t(\bar{u}_\alpha) = \text{true}$, that is, $t(v_{jk}) = \text{true}$, which implies $t(C_j) = \text{true}$.

Conversely, if the answer of the 3SAT is yes, then $\exists t : U \rightarrow \{\text{true}, \text{false}\}$ such that $t(C_j) = \text{true}$, $\forall j = \overline{1, m}$.

Let \bar{S}_1 be the stable set $\bar{S}_1 = \bigcup_{i=1}^n V'_i$ with n vertices, where

$$V'_i = \begin{cases} \{\bar{u}_i\}, & \text{if } t(u_i) = \text{true} \\ \{u_i\}, & \text{if } t(u_i) = \text{false} \end{cases}$$

Polynomial-time reductions - Maximum stable problem

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Then, since $t(C_j) = \text{true}$, $\forall j = \overline{1, m}$, it follows that there is $k_j \in \{1, 2, 3\}$ such that $t(v_{jk_j}) = \text{true}$. Let $\overline{S}_2 = \bigcup_{j=1}^m \{a_{jk_j}\}$. Obviously, \overline{S}_2 is a stable set in G having m vertices.

Let $\overline{S} = \overline{S}_1 \cup \overline{S}_2$. Obviously, $|\overline{S}| = n + m = k$ (hence $|\overline{S}| \geq k$). If we prove that \overline{S} is a stable set, then **the answer of SM for the instance (G, k) is yes.**

Suppose that $\exists v, w \in \overline{S}$ such that $e = vw \in E(G)$. Then one extremity of e is in \overline{S}_1 and the other in \overline{S}_2 . If $v \in \overline{S}_1$, then we have two cases:

- $v = u_\alpha$, $w = a_{jk_j}$, $\alpha \in \{1, 2, \dots, n\}$, $j \in \{1, 2, \dots, m\}$, $k_j \in \{1, 2, 3\}$ and $v_{jk_j} = u_\alpha$.

Since $t(v_{jk_j}) = \text{true}$, it follows that $t(u_\alpha) = \text{true}$, therefore $v = u_\alpha \notin \overline{S}_1$, contradiction.

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Polynomial-time reductions - Maximum stable problem

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- $v = \overline{u}_\alpha$, $w = a_{jk_j}$, $\alpha \in \{1, 2, \dots, n\}$, $j \in \{1, 2, \dots, m\}$, $k_j \in \{1, 2, 3\}$
and $v_{jk_j} = \overline{u}_\alpha$.

Since $t(v_{jk_j}) = \text{true}$, it follows that $t(\overline{u}_\alpha) = \text{true}$, therefore $t(u_\alpha) = \text{false}$. Hence, $v = \overline{u}_\alpha \notin \overline{S}_1$, contradiction. \square

Note that a similar proof can be given to prove $\text{SAT} \leq_P \text{SM}$, the only difference is that the graphs Z_i are complete graphs with k_i vertices.

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Polynomial-time reductions - Vertex colorings

COL

Instance: $G = (V, E)$ graph and $k \in \mathbb{N}^*$.

Question: Is there a k -coloring of (the vertices) of G ?

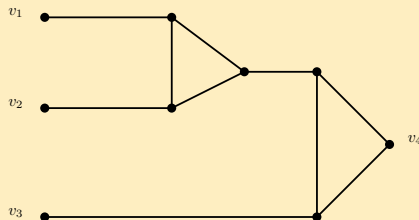
Theorem 4

$3\text{SAT} \leq_P \text{COL}$.

This theorem shows that the vertex coloring problem is NP-hard. The proof given below shows more: the problem, obtained from COL by restricting only to instances with $k = 3$, is NP-hard, too.

Lemma 1

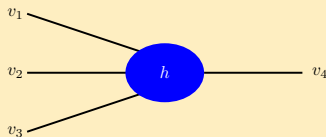
Let H be the graph



- a) If c is a 3-coloring of H s. t. $c(v_1) = c(v_2) = c(v_3) = a \in \{1, 2, 3\}$, then necessarily $c(v_4) = a$ (forcing).
- b) If $c : \{v_1, v_2, v_3\} \rightarrow \{1, 2, 3\}$ satisfies $c(\{v_1, v_2, v_3\}) \neq \{a\}$, ($a \in \{1, 2, 3\}$), then c can be extended to a 3-coloring c of H with $c(v_4) \neq a$.

Proof. By examining the list of 3-colorings of H .

We will use the following simplified representation of the graph H :

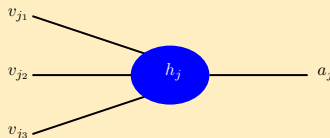


Proof of Theorem 4. Let us consider the data of an instance of 3SAT: $U = \{u_1, \dots, u_n\}$, ($n \in \mathbb{N}^*$), a set of boolean variables, and $C = C_1 \wedge \dots \wedge C_m$, ($m \in \mathbb{N}^*$) a CNF-formula with $C_j = v_{j_1} \vee v_{j_2} \vee v_{j_3}$, $\forall j = \overline{1, m}$, where $\forall i = \overline{1, 3}$, $\exists \alpha$ s. t. $v_{j_i} = u_\alpha$ or $v_{j_i} = \overline{u}_\alpha$.

We will build a graph G with the property that G is 3-colorable if and only if the answer of 3SAT for this instance is yes, that is, there is an assignment $t : U \rightarrow \{true, false\}$ such that $t(C) = true$.

The construction needs polynomial time, and consists in the following steps:

- Consider the disjoint graphs (V_i, E_i) , $\forall i = \overline{1, n}$, where $V_i = \{u_i, \overline{u}_i\}$ and $E_i = \{u_i \overline{u}_i\}$.
- For $C_j = v_{j_1} \vee v_{j_2} \vee v_{j_3}$, $\forall j = \overline{1, m}$, consider the graphs:



where, v_{j_k} ($k = \overline{1, 3}$) are the vertices corresponding to the literals v_{j_k} , the graphs h_j are disjoint, and a_j are distinct vertices.

- Consider a new vertex a , and all edges aa_j , $\forall j = \overline{1, m}$.
- Consider a new vertex b , all edges bu_i , $b\overline{u}_i$, $\forall i = \overline{1, n}$ and ba .

The graph G has a linear number of vertices w.r.t. $n + m$.

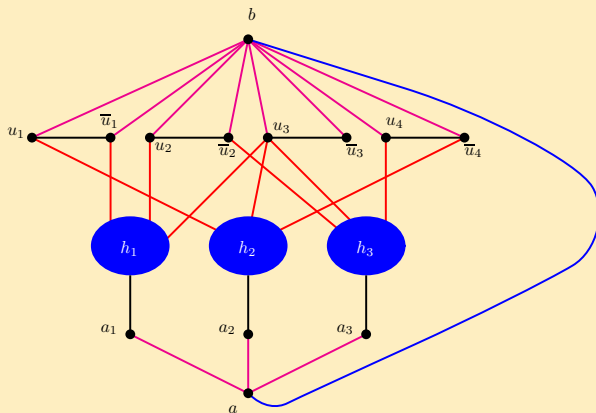
Polynomial-time reductions - Vertex colorings

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Example

$U = \{u_1, u_2, u_3, u_4\}$, $C = (\overline{u_1} \vee u_2 \vee u_3) \wedge (u_1 \vee u_3 \vee \overline{u_4}) \wedge (\overline{u_2} \vee u_3 \vee u_4)$.

The graph G is



Suppose that the answer of 3SAT for the considered instance is yes

Hence $\exists t : U \rightarrow \{true, false\}$ s. t. $t(C) = true$, that is, $t(C_j) = true$, $\forall j = \overline{1, m}$. We will show that G is 3-colorable.

We color, firstly, vertices u_i and $\overline{u_i}$, $\forall i \overline{1, n}$.

$$\begin{cases} c(u_i) = 1 \text{ and } c(\overline{u_i}) = 2, \text{ if } t(u_i) = true \\ c(u_i) = 2 \text{ and } c(\overline{u_i}) = 1, \text{ if } t(u_i) = false \end{cases}$$

Note that, if v is a literal, then $c(v) = 2$ if and only if $t(v) = false$.

Since t is a satisfying assignment, $t(C_j) = true$, $\forall j = \overline{1, m}$. It follows that $c(\{v_{j_1}, v_{j_2}, v_{j_3}\}) \neq \{2\}$, $\forall j = \overline{1, m}$.

By Lemma 1 b), we can extend c to a 3-coloring, in each graph h_j , such that $c(a_j) \neq 2$, that is $c(a_j) \in \{1, 3\}$.

By taking $c(a) = 2$ and $c(b) = 3$, we obtain a 3-coloring of G .

Conversely, suppose that G is 3-colorable.

We can assume that $c(b) = 3$ and $c(a) = 2$ (otherwise, rename the colors).

It follows that $\{c(u_i), c(\overline{u_i})\} = \{1, 2\}$, $\forall i = \overline{1, n}$ and $c(a_j) \in \{1, 3\}$, $\forall j = \overline{1, m}$.

By Lemma 1 a), it follows that $c(\{v_{j_1}, v_{j_2}, v_{j_3}\}) \neq 2$, $\forall j = \overline{1, m}$. This means that, $\forall j = \overline{1, m}$, there is a $v_{j_k} \in C_j$ such that $c(v_{j_k}) = 1$.

Hence, defining $t : U \rightarrow \{true, false\}$ by

$$t(u_i) = \begin{cases} true, & \text{if } c(u_i) = 1 \\ false, & \text{if } c(u_i) = 2 \end{cases},$$

we obtain an assignment with the property that $t(C_j) = true$, $\forall j = \overline{1, m}$.

Therefore the answer of 3SAT for the given instance is yes.

Polynomial-time reductions - Hamiltonian problems

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Remember: Let G be a (di)graph. A cycle C of G is a Hamiltonian cycle if $V(C) = V(G)$.

An open path P of G is a **Hamiltonian path** if $V(P) = V(G)$. A **Hamiltonian (di)graph** is a (di)graph which has a Hamiltonian cycle. A **traceable (di)graph** is a (di)graph which has a Hamiltonian path.

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Theorem 5

(Nash-Williams, 1969) The following five problems are polynomial equivalent:

CH: Given a graph G . Is G Hamiltonian?

TR: Given a graph G . Is G traceable?

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Polynomial-time reductions - Hamiltonian problems

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DCH: Given a digraph G . Is G Hamiltonian?

DTR: Given a digraph G . Is G traceable?

BCH: Given a bipartite graph G . Is G Hamiltonian?

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Remark

P_1 and P_2 are **polynomial equivalent** if $P_1 \leq_P P_2$ and $P_2 \leq_P P_1$.

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Proof of Theorem 5.

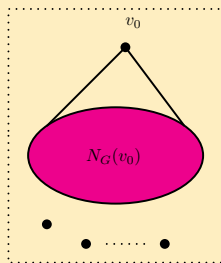
$CH \leq_P TR$ Let G be a graph and $v_0 \in V(G)$. We construct in polynomial time a graph H such that G is Hamiltonian if and only if H is traceable.

Let $V(H) = V(G) \cup \{x, y, z\}$ and $E(H) = E(G) \cup \{xv_0, yz\} \cup \{wy : w \in V(G) \cap N_G(v_0)\}$.

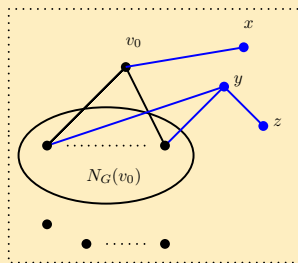
Polynomial-time reductions - Hamiltonian problems

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Proof of Theorem 5 cont'd.



G



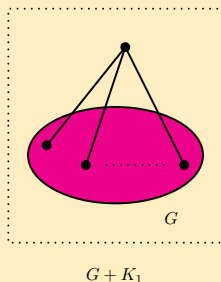
H

Then, H is traceable if and only if it has a Hamiltonian path, P , with extremities x and z (which have degree 1 in H). P exists in H if and only if in G there is a Hamiltonian path with an extremity in v_0 and the other a neighbor of v_0 , that is, if and only if G is Hamiltonian.

Polynomial-time reductions - Hamiltonian problems

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$TR \leq_P CH$ Let G be a graph. Consider $H = G + K_1$. Then, H is Hamiltonian if and only if G has a Hamiltonian path.



The equivalence of the problems DCH and DTR can be proved in a similar way.

$CH \leq_P DCH$ Let G be a graph. Let D be the digraph obtained from G by replacing each edge with a symmetric pair of arcs.

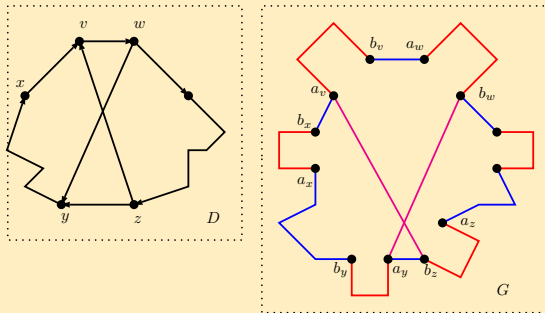
Polynomial-time reductions - Hamiltonian problems

Clearly any cycle in G gives rise to a cycle in D and conversely, any cycle in D gives rise to a cycle in G .

$\text{DCH} \leq_P \text{CH}$ Let D be a digraph. Each vertex $v \in V(D)$ is replaced by an undirected graph $P_3(v)$ with extremities a_v and b_v :

$$P_3(v) = (\{a_v, b_v, c_v, d_v\}, \{a_v c_v, c_v d_v, d_v b_v\}).$$

Each arc $vw \in E(D)$ is replaced by the (undirected) edge $b_v a_w$. Let G be the graph obtained (in polynomial time) in this way:



Polynomial-time reductions - Hamiltonian problems

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Each cycle C of D corresponds to a cycle in G and conversely, each cycle in G corresponds to a cycle of D . It follows that D is Hamiltonian if and only if G is Hamiltonian.

Note that if C is a cycle of G , then this is generated by a cycle C' of D , and $\text{length}(C) = 3 \cdot \text{length}(C') + \text{length}(C') = 4 \cdot \text{length}(C')$. It follows that any cycle of G is even, therefore G is a bipartite graph. Therefore the above proof ($\text{DCH} \leq_P \text{CH}$) is in fact $\text{DCH} \leq_P \text{BCH}$. Since $\text{BCH} \leq_P \text{CH}$ is obvious, the Theorem 5 is completely proved. \square

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Memento

- In order to prove that a certain decision problem is NP-complete:
 - I. First you have to show that the problem is in the class NP which means: there exists a polynomial time certificate for a any solution-candidate (for 3COL: given a function $c : V(G) \rightarrow \mathbb{N}$, we can verify if c is a coloring and uses at most three colors in $\mathcal{O}(n^2)$).
 - II. Second, we must polynomially reduce a known NP-complete problem to your decision problem (for 3COL: e. g. 3SAT can be polynomially reduced to 3COL).
- If step I. misses, then the decision problem in sight is NP-hard.

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Exercises for the 11th seminar

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Exercise 1. Show that the following problem is NP-complete

INT

Instance: $n, m \in \mathbb{N}^*, A \in \mathbb{Z}^{m \times n}, b \in \mathbb{Z}^m$.

Question: Is there an assignment $x \in \mathbb{Z}^n$ s. t. $Ax \leq b$?

(The inequality \leq between two vectors is componentwise.) *Hint:* You can try $\text{SM} \leq_P \text{INT}$.

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Exercise 2. Consider the following decision problem

ACYCLIC

Instance: $G = (V, E)$ a digraph and $p \in \mathbb{N}$.

Question: Is there $A \subseteq V$ such that $|A| \leq p$ and $G - A$ doesn't contain cycles?

Prove that $\text{SM} \leq_P \text{ACYCLIC}$.

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Exercise 3. A k -uniform hypergraph is a pair $H = (V, E)$, where $V \neq \emptyset$ is a finite set, $k \in \mathbb{N}^* \setminus \{1\}$, and $E \subseteq \mathcal{P}_k(V) = \{A \subseteq V : |A| = k\}$. It is easy to see that a 2-uniform hypergraph is a simple graph.

We say that a k -uniform hypergraph $H = (V, E)$ is *simple* if there is a function $c : V \rightarrow \{1, 2, \dots, k\}$ such that $\forall u, v \in V, u \neq v$, if $u, v \in e$ for a certain $e \in E$, then $c(u) \neq c(v)$. Now we consider the following decision problem

k -SIMPLE

Instance: H a k -uniform hypergraph.

Question: Is H simple?

- (a) Prove that problem 3-SIMPLE is NP-complete.
- (b) Show that problem 2-SIMPLE belongs to P.

Exercise 4. Consider the following decision problem:

AGM

Instance: G a graph, $k \in \mathbb{N}$.

Question: Has G a spanning tree T such that $\Delta(T) \geq k$?

Prove that $\text{AGM} \in \text{P}$.

Exercise 5. Let $D = (V, E)$ a loopless digraph. A stable set of D , $S \subseteq V$, is a quasi-kernel if every vertex $v \in V \setminus S$ can be accessed from inside S along a path having length at most 2.

- (a) Prove that a quasi-kernel can be constructed in $\mathcal{O}(n + m)$, where $n = |V|$ and $m = |E|$.
- (b) Show that 3-SAT is polynomially reducible to the problem of finding out if in a given digraph it exists a quasi-kernel containing a given vertex.

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Exercise 6. Consider the following decision problem: **LPL**

Instance: G a graph, $k \in \mathbb{N}$.

Question: Has G a path P such that $\text{length}(P) \geq k$?

Prove that LPL is NP-complete.

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Exercise 7. A **kernel** in a digraph $G = (V, E)$ is a stable set $S \subseteq V$ such that $\forall u \in V \setminus S$ exists $v \in S$ with $vu \in E$. We consider the following decision problem:

KERNEL

Instance: G a digraph.

Question: Has G a kernel?

Prove that the following construction leads to a polynomial reduction of SAT to KERNEL (i.e. $\text{SAT} \leq_P \text{KERNEL}$):

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Exercise 7 (cont'd). For every conjunction of clauses, F , instance of SAT, we define a digraph G (an instance for NUCLEU):

- for each clause C of F we add a 3-cycle to G

$$v_C^1, v_C^1 v_C^2, v_C^2, v_C^2 v_C^3, v_C^3, v_C^3 v_C^1;$$

- for every variable x which occurs in formula F , we add a 2-cycle to G

$$v_x, v_x v_{\bar{x}}, v_{\bar{x}}, v_{\bar{x}} v_x, v_x;$$

- for every clause C and every literal u which occurs in C we add to G three arcs

$$v_u v_C^1, v_u v_C^2, v_u v_C^3.$$

Exercises for the 11th seminar

Exercise 8. Consider the following decision problem

2SAT

Instance: \mathcal{C} a family of clauses of exactly two literals each.

Question: There exists an assignment of truth to the variables which satisfies all the clauses form \mathcal{C} ?

We define G (the implication) digraph: $V(G) = \text{set of all literals used in } \mathcal{C}$ and $E(G) = \{\overline{v_j} w_j, \overline{w_j} v_j : C_j = v_j \vee w_j, j = \overline{1, m}\}$ (each clause introduces in G two arcs). Show that \mathcal{C} is satisfiable if and only if x_i and $\overline{x_i}$ belong to different strongly connected components of G , $\forall i = \overline{1, n}$. Show that this property can be tested in $\mathcal{O}(n + m)$ time complexity.

Exercise 9. Show that the following problem is NP-complete.

MAX-2SAT

Instance: \mathcal{C} a family of clauses of at most two literals each and $k \in \mathbb{N}$.

Question: There exists an assignment of truth to the variables which satisfies at least k clauses form \mathcal{C} ?

Exercise 10. Consider the following decision problem

NAE-3SAT

Instance: \mathcal{C} a family of clauses of exactly three literals each.

Question: Is there a truth assignment to the variables such that each clause has at least one true literal and at least one false literal?

Prove that the following construction leads to a polynomial reduction of 3SAT to NAE 3SAT (i.e. $3SAT \leq_P \text{NAE } 3SAT$):

- we keep the boolean variables from 3SAT, $U = \{u_1, u_2, \dots, u_n\}$ and add a new variable x ;
- for each clause $C_j = v_{j_1} \vee v_{j_2} \vee v_{j_3}$ we add a new variable y_j and we replace C_j by two clauses:

$$C_j^1 = v_{j_1} \vee v_{j_2} \vee y_j, C_j^2 = v_{j_3} \vee x \vee \overline{y_j}.$$

Exercise 11. Consider the following decision problem

MAX-CUT

Instance: $G = (V, E)$ a graph, $c : E \rightarrow \mathbb{R}$ a weight on its edges, $k \in \mathbb{R}$.

Question: Is there a cut of G of weight at least k ?

Prove that the following construction leads to a polynomial reduction of NAE 3SAT to MAXCUT (i.e. $\text{NAE 3SAT} \leq_P \text{MAXCUT}$):

- consider an instance for NAE 3SAT problem with clauses $\mathcal{C} = \{C_1, \dots, C_m\}$ over the set of boolean variables $U = \{u_1, u_2, \dots, u_n\}$; we can suppose that every clause contains three different literals,
- $V(G) = \{u_i, \bar{u}_i : i = \overline{1, n}\}$ and we add to $E(G)$ the edges $u_i \bar{u}_i$ of weight $10m$,
- for each clause $C_j = v_{j_1} \vee v_{j_2} \vee v_{j_3}$ we add to $E(G)$ three edges: $v_{j_1} v_{j_2}$, $v_{j_2} v_{j_3}$, and $v_{j_1} v_{j_3}$ each of weight 1,
- $k = 10nm + 2m$.