

Finding optimal schedule using matroid

How to find an optimal schedule?

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 - 1.1 Find a set A of tasks that are early

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 - 1.2 Sort the tasks of A in increasing deadlines

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 - 1.1 Find a set A of tasks that are early
 - 1.2 Sort the tasks of A in increasing deadlines
 - 1.3 Add late tasks in any order

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2. Minimize penalties of late tasks \equiv maximize penalties of early tasks

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Finding an optimal schedule \equiv finding max-weighted indep. subset of M

Such M for task scheduling is a matroid

$M = (S, \mathcal{I})$ is a matroid

if $B \in \mathcal{I}$ and $A \subseteq B$, then $A \in \mathcal{I}$

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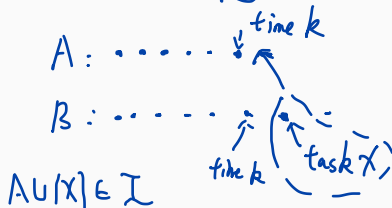
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- \mathcal{I} has the hereditary property:
if $A \subseteq B$ and $B \in \mathcal{I}$ then $A \in \mathcal{I}$
- Exchange property:
Say $A, B \in \mathcal{I}$ and $|B| > |A|$.

need to show $\exists x \in B - A$ s.t. $A \cup \{x\} \in \mathcal{I}$



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Let k be the time when the last task in A is finished

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Then $A \cup \{x\} \subseteq \mathcal{I}$

Greedy algorithm for finding optimal scheduling

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Running time: let $n = |S|$

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// $O(n \log n)$

$O(n \log n + n \cdot f(n))$

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Assume checking if $A \cup \{x\} \in \mathcal{I}$ takes $O(f(n))$.

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Claim: $f(n) = O(n)$ for task scheduling problem (Homework)

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Total running time: $O(n^2)$

Greedy algorithms

Horn formulas (Textbook Section 5.3)

Consider the following puzzle

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Question: what pets do they have?

Boolean formulas

Basics of boolean formulas

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- **Variables:** possibilities

Knowledge about variables is represented by a special type of boolean formulas

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- **Boolean variable:** $x = 1$ (true) or $x = 0$ (false)

Boolean formulas

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- **Boolean variable:** $x = 1$ (true) or $x = 0$ (false)

- **Literal:** x (positive literal), \bar{x} (negative literal)

$\neg x$

if $x=1$
then $\bar{x}=0$
if $x=0$
 $\bar{x}=1$

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Examples: $x \wedge \bar{y}$, $(x \wedge y) \implies z$

Horn formulas

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 - $(x \checkmark y) \implies z$

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- $(x \vee y) \implies z$ ✗
- $\implies z$

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- $(x \wedge \bar{y}) \implies z$ ✗

- $(x \vee y) \implies z$ ✗

- $\implies z$ ✓

$\vdash \implies z$

$(x \wedge y \wedge z) \implies (\bar{w})$ ✗

Horn formulas

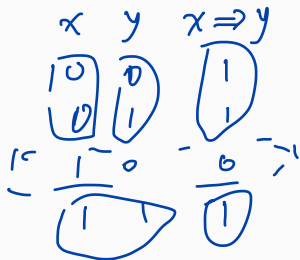
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 - $(x \wedge \bar{y}) \implies z$ ✗
 - $(x \vee y) \implies z$ ✗
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- **Pure negative clauses** $\bar{x}_1 \vee \bar{x}_2 \vee \cdots \vee \bar{x}_n$

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 - $\implies z$ ✓
- **Pure negative clauses** $\bar{x}_1 \vee \bar{x}_2 \vee \dots \vee \bar{x}_n$
OR of any number of negative literals



Horn formula example

Consider the puzzle:

- If Alice has a dog, then Bob has a cat
- If Charlie and Bob both have pets of the same species, then Alice has a cat
- Charlie and Alice don't share a pet of the same species

a Alice has dog
b Bob has a cat.

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Define variables:

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- b : Bob has a dog
- c : Charlie has a dog

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Define variables:

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- b : Bob has a dog
- c : Charlie has a dog
- x : Alice has a cat

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- x : Alice has a cat
- y : Bob has a cat
- z : Charlie has a cat

$a=1$: Alice has a dog

$a=0$: Alice has no dog

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Modelled by a set of Horn clauses:

$$a \implies y$$

$$((b \wedge c) \vee (y \wedge z)) \implies x \quad ?$$

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Horn formula example

$$p \Rightarrow q \Leftrightarrow \bar{p} \vee q$$

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Modelled by a set of Horn clauses:

$$a \Rightarrow y$$

$$(b \wedge c) \Rightarrow x$$

$$(y \wedge z) \Rightarrow x$$

$$a \Rightarrow \bar{c} \Leftrightarrow \bar{a} \vee \bar{c}$$

$$x \Rightarrow \bar{z} \Leftrightarrow \bar{x} \vee \bar{z}$$

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$$\bar{a} \vee \bar{c}$$

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Define variables:

- a : Alice has a dog 0/1
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- c : Charlie has a dog 0/1
- x : Alice has a cat 0/1
- y : Bob has a cat 0/1
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Modelled by a set of Horn clauses:

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- c : Charlie has a dog 1 0
- x : Alice has a cat 1 0
- y : Bob has a cat 1 0
- z : Charlie has a cat 0 1

Modelled by a set of Horn clauses:

$$\checkmark a \implies y$$

$$\checkmark (b \wedge c) \implies x$$

$$\checkmark (y \wedge z) \implies x$$

$$\checkmark \bar{a} \vee \bar{c}$$

$$\checkmark \bar{x} \vee \bar{z}$$

Question: satisfying assignment?

Greedy approach for Horn formulas

Problem (Horn Satisfiability)

Given a set of Horn clauses, determine whether or not there is a consistent explanation,

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Given a set of Horn clauses, determine whether or not there is a consistent explanation, i.e., an assignment of 0/1 to variables that satisfy all clauses

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Example: $(x \wedge y) \implies z, \bar{x} \vee \bar{w}$

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$x = 0, y = 0, z = 0, w = 0$

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Greedy heuristic: start with all 0. Only set a variable to 1 if you need to, i.e., when an implication says you need to

$$\begin{array}{ll} x \Rightarrow y & \Rightarrow x \\ 1 & \nwarrow y \text{ must be } 1 \\ & : 1 \Rightarrow x \rightarrow x \text{ must be } 1 \end{array}$$

Greedy approach for Horn formulas

Problem (Horn Satisfiability)

Given a set of Horn clauses, determine whether or not there is a consistent explanation, i.e., an assignment of 0/1 to variables that satisfy all clauses

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Recall: $p \implies q \iff \bar{p} \vee q$

Pseudocode

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def GREEDY_HORN(set of Horn clauses):
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0	0

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x y

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$\implies x$ ✗

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1 0 $x \implies y$ ✗ set y to 1

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Running time: Let n be the size of the Horn formula, i.e., the number of occurrences of literals. $\ell_2 \Rightarrow \ell_1, \dots, \ell_{n-1} \Rightarrow \ell_{n-2}, \ell_n \Rightarrow \ell_{n-1}, \Rightarrow \ell_n$

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Total running time: $O(n^2)$.

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Total running time: $O(n^2)$. Can be improved to $O(n)$ (exercise)