## Lecture 22: Intractability II NP-Hardness

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## Overview

Navigating the world of P and NP
 2SAT

- Beyond decision problems: NP-hardness
  - NP-hard reductions

• Acknowledgements

Similar looking problems, wildly different complexity:

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- In general, we need to be careful when distinguishing or making reductions between problems.

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- Run BFS or DFS from each literal y, and call it *bad* if for some  $i \in [n]$ , the BFS from y visits both  $x_i, \overline{x_i}$
- If for some *i* ∈ [*n*], both *x<sub>i</sub>* and *x<sub>i</sub>* are bad, then return NO. Otherwise, return YES.

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• The above is our definition of *NP-hardness*:

Problem *B* is *NP-hard* if there is NP-complete problem *A* such that  $A <_{T} B$ .

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- TSP-OPT:
  - Input: complete graph G(V, E, d) where  $d: E \to \mathbb{R}_{\geq 0}$
  - Output: hamiltonian cycle in G of minimum total distance

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  - add vertices u<sub>e</sub>, v<sub>e</sub>,
  - and edges:  $\{x, u_e\}, \{x, v_e\}, \{u, u_e\}, \{v, v_e\}, \{u_e, v_e\},$

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- Edge gadget *H<sub>e</sub>*:
- Let H(U, F) be graph given by:

• 
$$U = V \sqcup \{x\} \sqcup \{u_e, v_e\}_{\{u,v\}=:e \in E}$$

•  $F = \{\{x, w\}\}_{w \in U \setminus \{x\}} \sqcup \{\{u_e, v_e\}\}_{e \in E} \sqcup \{\{u, u_e\}, \{v, v_e\}\}_{\{u, v\}=:e \in E}$ Note that *H* does not have any edges from *G* 

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In all above cases, add four of five *edge gadget*  $H_e$  edges Analyzing the cut given by S:

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In all above cases, add four of five *edge gadget*  $H_e$  edges Analyzing the cut given by S:

- For every  $w \in I$ , the edge  $\{x, w\}$  is cut by S
- For every edge  $\{u, v\} =: e \in E$ , exactly 4 edges of  $H_e$  are cut.

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- Letting e(I) be number of edges between elements of I in G:

$$|\delta(S)| = |I| + \sum_{e \in E} |\delta_{H_e}(S)| \le |I| + 3e(I) + 4(|E| - e(I)) = |I| + 4|E| - e(I)$$

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• As  $|\delta(S)| \ge k + 4|E|$ , we have

$$|I| \geq k + e(I)$$

So for each u, v ∈ I with {u, v} ∈ E, we can afford to remove one of the endpoints from S, decreasing |I| by one. After e(I) removals, get our independent set.

# Acknowledgement

Based on

- [Erickson 2019, Chapter 12]
- Debmalya's Lecture 22

https://courses.cs.duke.edu/fall19/compsci638/fall19\_ notes/lecture22.pdf

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