

# Design and Analysis of Algorithms (III)

NP Problem

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# Homework

- Assignment 1 is announced! (deadline Mar. 26)

## NP-Complete Problems

# Set Cover

## Set Cover

- **Input:** A set of elements  $B$ , sets  $S_1, \dots, S_m \subseteq B$
- **Output:** A selection of the  $S_i$  whose union is  $B$ .
- **Cost:** Number of sets picked.

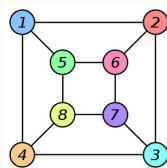
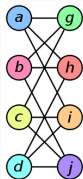
# Graph Isomorphism

## Graph Isomorphism

An isomorphism of graphs  $G$  and  $H$  is a bijection between the vertex sets of  $G$  and  $H$

$$f : V(G) \rightarrow V(H)$$

such that any two vertices  $u$  and  $v$  of  $G$  are adjacent in  $G$  if and only if  $f(u)$  and  $f(v)$  are adjacent in  $H$ .



# Hard Problems, Easy Problems

Hard problems (NP-complete)	Easy problems (in P)
3SAT	2SAT, Horn SAT
Traveling salesman problem	Minimum spanning tree
Longest path	Shortest path
3D matching	Bipartite matching
Knapsack	Unary knapsack
Independent set	Independent set on trees
Integer linear programming	Linear programming
Rudrata path	Euler path
Balanced cut	Minimum cut

# NP

Recall a **search problem** is defined by:

- ① An efficient checking algorithm  $C$ , taking as input the given **instance**  $I$ , a **solution**  $S$ , and outputs **true** iff  $S$  is a solution  $I$ .
- ② The running time of  $C(I, S)$  is bounded by a polynomial in  $|I|$ .

We denote the class of all search problems by **NP**.

# P

An algorithm that takes as input an *instance*  $I$  and has a running time polynomial in  $|I|$ .

- $I$  has a solution, the algorithm returns such a solution;
- $I$  has no solution, the algorithm correctly reports so.

The class of all *search problems* that can be solved in *polynomial time* is denoted **P**.



# Why P and NP

**P:** polynomial time

**NP:** nondeterministic polynomial time

# Complementation

A class of problems  $\mathcal{C}$  is **closed under complementation** if for any problem in  $\mathcal{C}$ , its complement is also in  $\mathcal{C}$ .

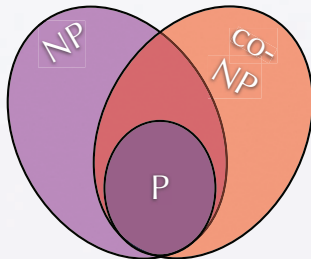
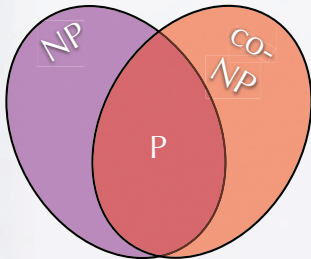
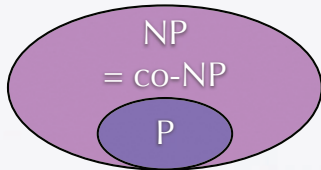
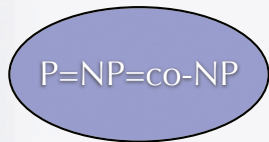
**P**: is closed under complementation.

**NP** ?

## Example (Complementation of TSP)

Given  $n$  cities with their intercity distances, is it the case that there does not exist any tour length  $k$  or less?

# Conjectures



$$P \neq NP$$

### Theorem Proving

- **Input:** A mathematical statement  $\varphi$  and  $n$ .
- **Problem:** Find a proof of  $\varphi$  of length  $\leq n$  if there is one.

A formal proof of a mathematical assertion is written out in excruciating detail, it can be checked mechanically, by an **efficient algorithm** and is therefore in **NP**.

So if **P = NP**, there would be an efficient method to prove any theorem, thus eliminating the need for mathematicians!

# Solve One and All Solved

Even if we believe  $P \neq NP$ , can we find an evidence that these particular problems have no efficient algorithm?

Such evidence is provided by **reductions**, which translate one search problem into another.

We will show that the hard problems in previous lecture exactly the same problem, the **hardest search problems** in **NP**.

If one of them has a **polynomial time algorithm**, then every problem in **NP** has a polynomial time algorithm.

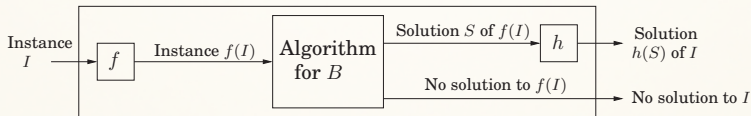
# Reduction Between Search Problems

A **reduction** from  $A$  to  $B$  is a **polynomial** time algorithm  $f$  that transforms any instance  $I$  of  $A$  into an instance  $f(I)$  of  $B$

Together with another **polynomial** time algorithm  $h$  that maps any solution  $S$  of  $f(I)$  back into a solution  $h(S)$  of  $I$ .

If  $f(I)$  has **no solution**, then neither does  $I$ .

These two translation procedures  $f$  and  $h$  imply that any algorithm for  $B$  can be **converted** into an algorithm for  $A$ .



# The Two Ways to Use Reductions

Assume there is a **reduction** from a problem  $A$  to a problem  $B$ .

$$A \rightarrow B$$

- If we can solve  $B$  **efficiently**, then we can also solve  $A$  **efficiently**.
- If we know  $A$  is **hard**, then  $B$  must be **hard** too.

If  $A \rightarrow B$  and  $B \rightarrow C$ , then  $A \rightarrow C$ .

# NP-Completeness

## Definition

A **NP** problem is **NP-complete** if all other **NP** problems reduce to it.



# Reductions to NP-Complete

**NP-complete** problems are hard: all other search problems reduce to them.

For a problem to be **NP-complete**, it can solve every **NP** problem in the world.

If even one **NP-complete** problem is in **P**, then **P = NP**.

If a problem **A** is NP-complete, a new NP problem **B** is proved to be NP-complete, by reducing **A** to **B**.

# Co-NP-Completeness

## Definition

A co-NP problem is **co-NP-complete** if all other co-NP problems reduce to it.

A problem is **NP-complete** if and only if its complement is **co-NP-complete**.

If a problem and its complement are **NP-complete** then **co-NP = NP**.

# Tautology

A **CNF formula**  $f$  is unsatisfiable if and only if its negation is a **tautology**. The negation of a CNF formula can be converted into a **DNF** formula. The resulting DNF formula is a **tautology** if and only if the negation of the CNF formula is a tautology.

The problem **Tautology**: Given a formula  $f$  in DNF, is it a tautology?

- **Tautology** is in **P** if and only if **co-NP = P**, and
- **Tautology** is in **NP** if and only if **co-NP = NP**.

# Factoring

The difficulty of **Factoring** is of a different nature than that of the other hard search problems we have just seen.

Nobody believes that **Factoring** is **NP-complete**.

One evidence is that a number can always be factored into primes.

Another difference: **Factoring** succumbs to the power of **quantum computation**, while **SAT**, **TSP** and the other NPC problems do not seem to.

# NPI

## Primality

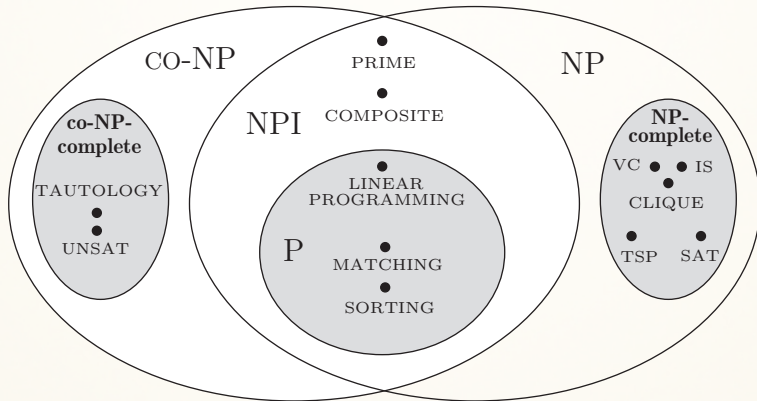
Given an integer  $k \geq 2$ , is  $k$  a prime number?

## Composite

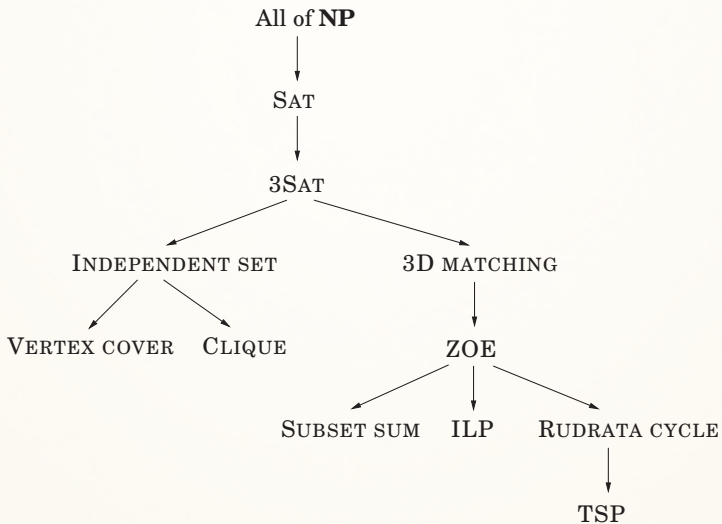
Given an integer  $k \geq 4$ , are there two integers  $p, q \geq 2$  such that  $k = pq$ ?

Such class is called **NPI**, standing for **NP-Intermediate**.

# NPI (A Problematic Category)



## The Reductions

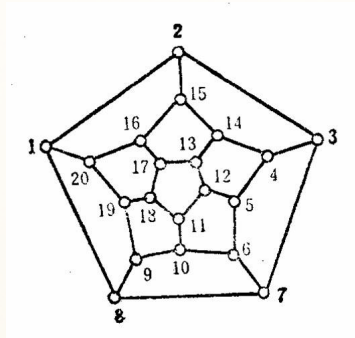




1. Rudrata path  $\rightarrow$  Rudrata cycle

# Rudrata Cycle

**Rudrata Cycle:** Given a graph, find a cycle that visits each vertex exactly once.



# Rudrata $(s, t)$ -path $\rightarrow$ Rudrata cycle

A **Rudrata  $(s, t)$ -path** problem specifies two vertices  $s$  and  $t$  and wants a path starting at  $s$  and ending at  $t$  that goes through each vertex exactly once.

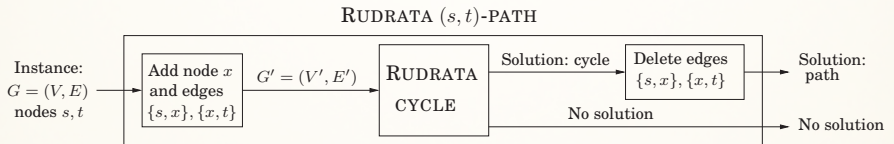
**Q:** Is it possible that **Rudrata cycle** is easier than **Rudrata  $(s, t)$ -path**?

The reduction maps an **instance  $G$**  of **Rudrata  $(s, t)$ -Path** into an **instance  $G'$**  of **Rudrata cycle** as follows:  $G'$  is  $G$  with an additional vertex  $x$  and two new edges  $\{s, x\}$  and  $\{x, t\}$ .

# Rudrata $(s, t)$ -path $\rightarrow$ Rudrata cycle



# Rudrata $(s, t)$ -path $\rightarrow$ Rudrata cycle



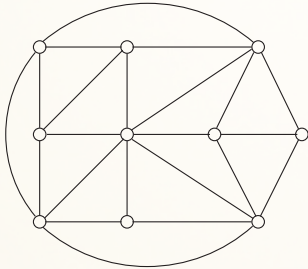
2. 3SAT  $\rightarrow$  Independent set

# 3 SAT

The instances of **3SAT**, is set of clauses, each with three or fewer literals.

$$(x \vee y \vee z)(x \vee \bar{y})(y \vee \bar{z})(z \vee \bar{x})(\bar{x} \vee \bar{y} \vee \bar{z})$$

# Independent Set



**Independent set:** Given a graph  $G$  and an integer  $g$ , find  $g$  vertices, no two of which have an edge between them.



# True Assignment

To form a satisfying truth assignment we must pick one literal from each clause and give it the value `true`.

The choices must be consistent, if we choose  $\bar{x}$  in one clause, we cannot choose  $x$  in another.

**Solution:** put an edge between any two vertices that correspond to opposite literals.

# Clause

Represent a clause, say  $(x \vee \bar{y} \vee z)$ , by a triangle, with vertices labeled  $x, \bar{y}, z$ .

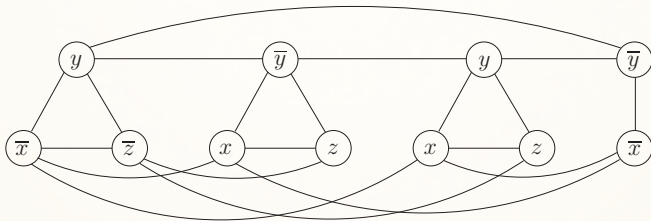
Because a triangle has its three vertices maximally connected, and thus forces to pick only one of them for the **independent set**.

# 3SAT $\rightarrow$ Independent set

Given an instance  $I$  of 3SAT, create an instance  $(G, g)$  of Independent set as follows,

- A triangle for each clause, with vertices labeled by the clause's literals.
- Additional edges between any two vertices that represent opposite literals.
- The goal  $g$  is set to the number of clauses.

## 3SAT $\rightarrow$ Independent set



$$(\bar{x} \vee y \vee \bar{z})(x \vee \bar{y} \vee z)(x \vee y \vee z)(\bar{x} \vee \bar{y})$$

### 3. SAT $\rightarrow$ 3SAT

# SAT $\rightarrow$ 3SAT

This is an interesting and common kind of reduction, from a problem to a special case of itself.

Given an **instance**  $I$  of **SAT**, use exactly the same **instance** for **3SAT**, except that any clause with more than three literals,

$$(a_1 \vee a_2 \vee \dots \vee a_k)$$

is replaced by a set of clauses,

$$(a_1 \vee a_2 \vee y_1)(\overline{y_1} \vee a_3 \vee y_2)(\overline{y_2} \vee a_4 \vee y_3) \dots (\overline{y_{k-3}} \vee a_{k-1} \vee a_k)$$

where the  $y_i$ 's are new variables.

The reduction is in polynomial and  $I'$  is equivalent to  $I$  in terms of satisfiability.

# SAT $\rightarrow$ 3SAT

$$\left\{ \begin{array}{c} (a_1 \vee a_2 \vee \dots \vee a_k) \\ \text{is satisfied} \end{array} \right\} \iff \left\{ \begin{array}{c} \text{there is a setting of the } y_i\text{'s for which} \\ (a_1 \vee a_2 \vee y_1) (\overline{y}_1 \vee a_3 \vee y_2) \dots (\overline{y}_{k-3} \vee a_{k-1} \vee a_k) \\ \text{are all satisfied} \end{array} \right\}$$

Suppose that the clauses on the right are all satisfied. Then **at least** one of the literals  $a_1, \dots, a_k$  must be **true**. Otherwise  $y_1$  would have to be **true**, which would in turn force  $y_2$  to be **true**, and so on.

Conversely, if  $(a_1 \vee a_2 \vee \dots \vee a_k)$  is **satisfied**, then some  $a_i$  must be **true**. Set  $y_1, \dots, y_{i-2}$  to **true** and the rest to **false**.

# SAT $\rightarrow$ 3SAT

**3SAT** remains hard even under the further **restriction** that no variable appears in more than three clauses.

Suppose that in the **3SAT instance**, variable  $x$  appears in  $k > 3$  clauses. Then replace its first appearance by  $x_1$ , its second by  $x_2$ , and so on, replacing each of its  $k$  appearances by a different new variable.

Finally, add the clauses

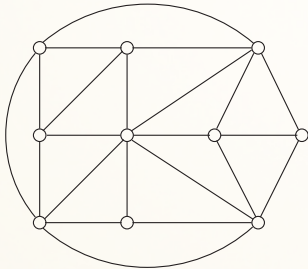
$$(\overline{x_1} \vee x_2)(\overline{x_2} \vee x_3) \dots (\overline{x_k} \vee x_1)$$

In the new formula no variable appears more than three times (and in fact, no literal appears more than **twice**).



4. Independent set  $\rightarrow$  Vertex cover

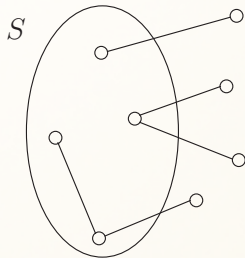
# Vertex Cover



**Vertex cover:** Given a graph  $G$  and an integer  $b$ , find  $b$  vertices cover (touch) every edge.

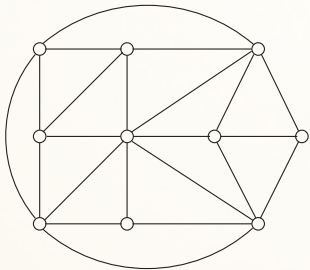
# Independent set $\rightarrow$ Vertex cover

A set of nodes  $S$  is a vertex cover of graph  $G = (V, E)$  iff the remaining nodes,  $V - S$ , are an **independent set** of  $G$ .



5. Independent set  $\rightarrow$  Clique

# Clique



**Clique:** Given a graph  $G$  and an integer  $g$ , find  $g$  vertices such that all possible edges between them are present.

# Independent set $\rightarrow$ Clique

The complement of a graph  $G = (V, E)$  is  $\overline{G} = (V, \overline{E})$ , where  $\overline{E}$  contains precisely those unordered pairs of vertices that are not in  $E$ .  
A set of nodes  $S$  is an independent set of  $G$  iff  $S$  is a **clique** of  $\overline{G}$ .

Therefore, we can reduce **Independent set** to **Clique** by mapping an instance  $(G, g)$  of **Independent set** to the corresponding instance  $(\overline{G}, g)$  of **Clique**.

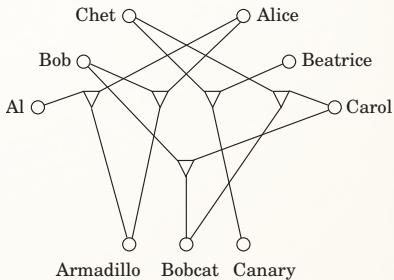
## 6. 3SAT $\rightarrow$ 3D matching

# Three-Dimensional Matching

**3D matching:** There are  $n$  boys,  $n$  girls, and  $n$  pets. The compatibilities are specified by a set of **triples**, each containing a boy, a girl, and a pet.

A triple  $(b, g, p)$  means that boy  $b$ , girl  $g$ , and pet  $p$  get along well together.

To find  $n$  disjoint triples and thereby create  $n$  harmonious households.

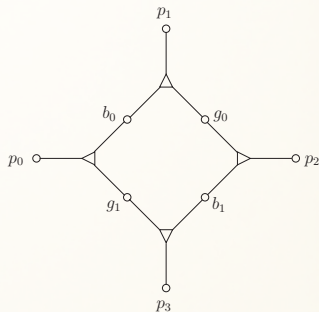




# 3SAT $\rightarrow$ 3D matching

Consider a set of four triples, each represented by a triangular node joining a boy, girl, and pet.

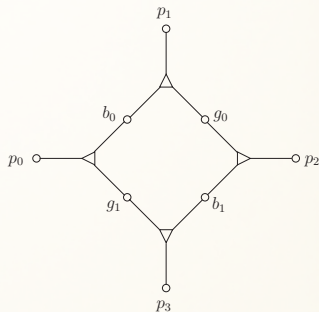
Any matching must contain either the two triples  $(b_0, g_1, p_0)$ ,  $(b_1, g_0, p_2)$  or  $(b_0, g_0, p_1)$ ,  $(b_1, g_1, p_3)$ .



# 3SAT $\rightarrow$ 3D matching

Therefore, this “**gadget**” has two possible states: it behaves like a Boolean variable.

Transform an **instance** of **3SAT** to one of **3D matching**, by creating a gadget for each variable  $x$ .



# 3SAT $\rightarrow$ 3D matching

For each clause  $c$  introduce a new boy  $b_c$  and a new girl  $g_c$ .

E.g.,  $c = (x \vee \bar{y} \vee z)$ ,  $b_c, g_c$  will be involved in three triples, one for each literal in the clause.

And the pets in these triples must reflect the three ways whereby the clause can be satisfied:

- ①  $x = \text{true}$ ,
- ②  $y = \text{false}$ ,
- ③  $z = \text{true}$ .

# 3SAT $\rightarrow$ 3D matching

For  $x = \text{true}$ , we have the triple  $(b_c, g_c, p_{x1})$ , where  $p_{x1}$  is the pet  $p_1$  in the gadget for  $x$ .

- If  $x = \text{true}$ , then  $b_{x0}$  is matched with  $g_{x1}$  and  $b_{x1}$  with  $g_{x0}$ , and so pets  $p_{x0}$  and  $p_{x2}$  are taken.
- If  $x = \text{false}$ , then  $p_{x1}$  and  $p_{x3}$  are taken, and so  $g_c$  and  $b_c$  cannot be accommodated.

We do the same thing for the other two literals, which yield triples involving  $b_c$  and  $g_c$  with either  $p_{y0}$  or  $p_{y2}$  and with either  $p_{z1}$  or  $p_{z3}$ .

# 3SAT $\rightarrow$ 3D matching

We have to make sure that for every occurrence of a literal in a clause  $c$  there is a different pet to match with  $b_c$  and  $g_c$ .

This is easy: an earlier reduction guarantees that no literal appears more than twice, and so each variable gadget has enough pets, two for negated occurrences and two for positive.

# 3SAT $\rightarrow$ 3D matching

The last problem remains: in the matching defined so far, some pets may be left unmatched.

If there are  $n$  variables and  $m$  clauses, then  $2n - m$  pets will be left unmatched.

Add  $2n - m$  new boy-girl couples that are “generic animal-lovers”, and match them by triples with all the pets!

## 7. 3D matching $\rightarrow$ ZOE

# Zero-One Equations

ZOE (Zero-one equations)

Given an  $m \times n$  matrix  $A$  with  $0 - 1$  entries, and find a  $0 - 1$  vector  $\mathbf{x} = (x_1, \dots, x_n)$  such that the  $m$  equations  $A\mathbf{x} = \mathbf{1}$ ; are satisfied.



# 3D matching $\rightarrow$ ZOE

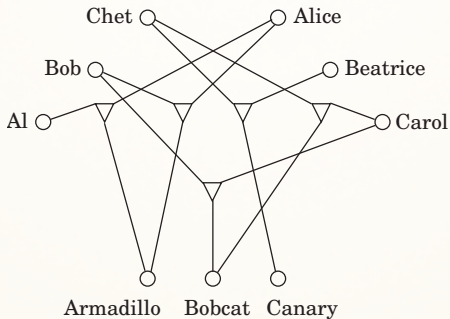
Assume in **3D matching**, there are  $m$  boys,  $m$  girls,  $m$  pets, and  $n$  boy-girl-pet triples.

We have  $0 - 1$  variables,  $x_1, \dots, x_n$ , one per triple, where  $x_i = 1$  means that the  $i$ -th triple is chosen for the matching, and  $x_i = 0$  means that it is not chosen.

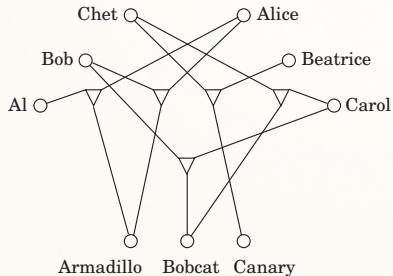
For each boy, girl, or pet, suppose that the triples containing him (or her, or it) are those numbered  $j_1, j_2, \dots, j_k$ ; the appropriate equation is then

$$x_{j_1} + x_{j_2} + \dots + x_{j_k} = 1$$

# 3D matching $\rightarrow$ ZOE



# 3D matching $\rightarrow$ ZOE



$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{pmatrix}$$

## 8. ZOE $\rightarrow$ Subset Sum

# Subset Sum

**Subset sum:** Find a subset of a given set of integers that adds up to exactly  $W$ .

# ZOE $\rightarrow$ Subset Sum

This is a reduction between two special cases of **ILP**:

- One with many equations but only **0 – 1** coefficients;
- The other with a single equation but arbitrary integer coefficients.

The reduction is based on a simple and time-honored idea: **0 – 1** vectors can encode numbers!

If the columns is regarded as binary integers, a subset of the integers corresponds to the columns of **A** that add up to the binary integer **11...1**.

This is an **instance** of **Subset sum**. The reduction seems complete!

# An Example

$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

# ZOE $\rightarrow$ Subset Sum

Except for one detail: **carry**.

E.g., 5-bit binary integers can add up to  $11111 = 31$ , for example,  
 $5 + 6 + 20 = 31$  or, in binary,

$$00101 + 00110 + 10100 = 11111$$

even when the sum of the **corresponding vectors** is not  $(1, 1, 1, 1, 1)$ .

Solution: The column vectors not as integers in **base 2**, but as integers in **base  $n + 1$** , one more than the number of columns.

At most  $n$  integers are added, and all their digits are 0 and 1. There is no carry anymore.



## 9. ZOE $\rightarrow$ ILP

# Special Cases

**3SAT** is a special case of **SAT**, or, **SAT** is a generalization of **3SAT**.

By special case we mean that the *instances* of **3SAT** are a subset of the *instances* of **SAT**.

There is a **reduction** from **3SAT** to **SAT**, where the input has no transformation, and the solution to the target instance also kept unchanged.

A useful and common way of establishing that a problem is **NP-complete**: it is a generalization of a known **NP-complete problem**.

E.g., the **Set cover** problem is **NP-complete** because it is a generalization of **Vertex cover**.

# ZOE $\rightarrow$ ILP

In **ILP** we are looking for an integer vector  $\mathbf{x}$  that satisfies  $A\mathbf{x} \leq b$ , for given matrix  $A$  and vector  $b$ .

To write an **instance** of **ZOE** in this precise form, we need to rewrite each equation of the **ZOE instance** as two inequalities, and to add for each variable  $x_i$  the inequalities  $x_i \leq 1$  and  $-x_i \leq 0$ .

## 10. ZOE $\rightarrow$ Rudrata Cycle

# ZOE $\rightarrow$ Rudrata Cycle

In **Rudrata cycle**, seek a cycle in a graph that visits every vertex exactly once.

In **ZOE**, given an  $m \times n$  matrix  $A$  with  $0 - 1$  entries, and find a  $0 - 1$  vector  $\mathbf{x} = (x_1, \dots, x_n)$  such that the  $m$  equations  $A\mathbf{x} = \mathbf{1}$ ; are satisfied.

# ZOE $\rightarrow$ Rudrata Cycle

We will prove it **NP-complete** in two stages:

- ① Firstly, reduce **ZOE** to a generalization of **Rudrata cycle**, called **Rudrata cycle with Paired edges**.
- ② Secondly, get rid of the extra features of that problem and reduce it to the plain **Rudrata cycle**.

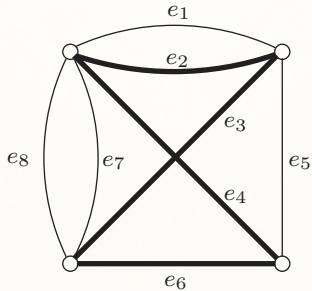
# Rudrata Cycle with Paired Edges

Given a **graph**  $G = (V, E)$  and a set  $C \subseteq E \times E$  of pairs of edges.  
Find a cycle that,

- ① visits all vertices once,
- ② for every pair of edges  $(e, e')$  in  $C$ , traverses either edge  $e$  or edge  $e'$  exactly one of them.

Notice that two or more parallel edges between two nodes are allowed.

# An Example



$$C = \{(e_1, e_3), (e_5, e_6), (e_4, e_5), (e_3, e_7), (e_3, e_8)\}$$



# ZOE $\rightarrow$ Rudrata Cycle with Paired Edges

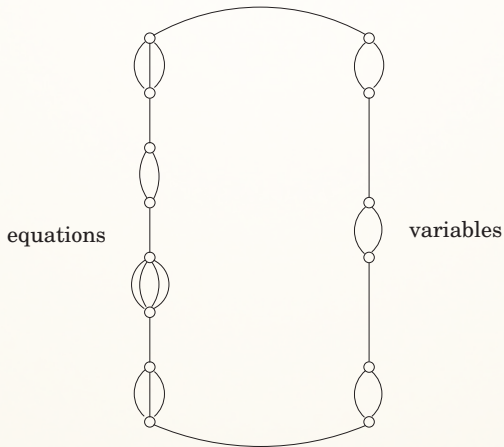
Given an instance of ZOE,  $Ax = 1$ , where  $A$  is an  $m \times n$  matrix with  $0 - 1$  entries, the graph is as follows

- A cycle that connects  $m + n$  collections of parallel edges.
- Each variable  $x_i$  has two parallel edges, for  $x_i = 1$  and  $x_i = 0$ ).
- Each equation  $x_{j_1} + \dots + x_{j_k} = 1$  involving  $k$  variables has  $k$  parallel edges, one for every variable appearing in the equation.

Any Rudrata cycle traverses the  $m + n$  collections of parallel edges one by one, choosing one edge from each collection.

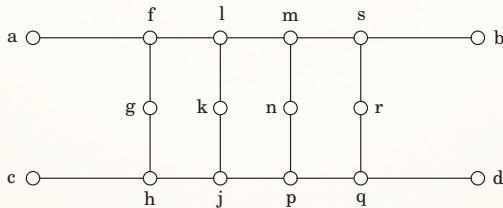
The cycle “chooses” for each variable a value  $0$  or  $1$  and, for each equation, a variable appearing in it.

# ZOE $\rightarrow$ Rudrata Cycle with Paired Edges



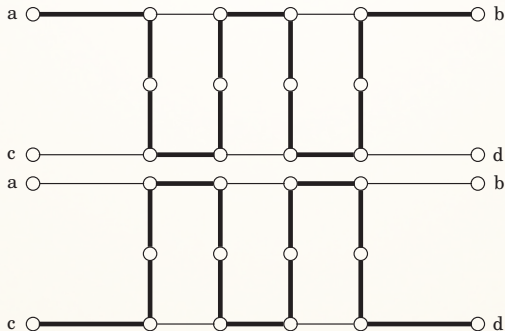
# Get Rid of Edge Pairs

Consider the **graph**, and suppose it is a part of a larger graph **G** in such a way that only the four endpoints **a, b, c, d** touch the **rest** of the graph.



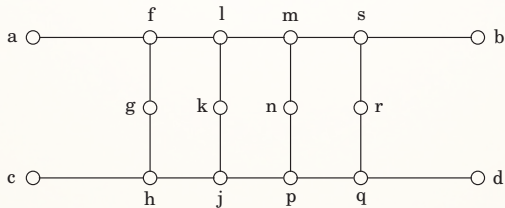
# Get Rid of Edge Pairs

We claim that this graph has the following important property: in any **Rudrata cycle** of **G** the subgraph shown must be traversed in one of the two ways.



# Get Rid of Edge Pairs

This gadget behaves just like two edges  $\{a, b\}$  and  $\{c, d\}$  that are paired up in the **Rudrata cycle with paired edges**.



$$C = \{(\{a, b\}, \{c, d\})\}$$

# Rudrata Cycle with Paired Edges $\rightarrow$ Rudrata Cycle

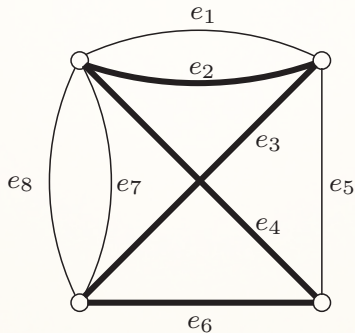
Go through the pairs in  $C$  one by one. To get rid of each pair  $(\{a, b\}, \{c, d\})$  by replacing the two edges with the gadget.

For any other pair in  $C$  that involves  $\{a, b\}$ , replace the edge  $\{a, b\}$  with the new edge  $\{a, f\}$ , where  $f$  is from the gadget.

Similarly,  $\{c, h\}$  replaces  $\{c, d\}$ .

The Rudrata cycles in the resulting graph will be in one-to-one correspondence with the Rudrata cycles in the original graph that conform to the constraints in  $C$ .

# An Example



$$C = \{(e_1, e_3), (e_5, e_6), (e_4, e_5), (e_3, e_7), (e_3, e_8)\}$$

11. Rudrata cycle  $\rightarrow$  TSP



# Rudrata cycle $\rightarrow$ TSP

Given a graph  $G = (V, E)$ , construct the **instance** of the **TSP**:

- The set of **nodes** is the same as  $V$ .
- The **distance** between cities  $u$  and  $v$  is  $1$  if  $\{u, v\}$  is an **edge** of  $G$  and  $1 + \alpha$  otherwise, for some  $\alpha > 1$  to be **determined**.
- The **budget** of the **TSP instance** is  $|V|$ .

If  $G$  has a **Rudrata cycle**, then the same cycle is also a tour within the **budget** of the **TSP instance**.

If  $G$  has no **Rudrata cycle**, then there is no solution: the cheapest possible **TSP** tour has cost at least  $n + \alpha$ .

# Rudrata cycle $\rightarrow$ TSP

If  $\alpha = 1$ , then all distances are either 1 or 2, and so this instance of the TSP satisfies the triangle inequality: if  $i, j, k$  are cities, then

$$d_{ij} + d_{jk} \geq d_{ik}$$

This is a special case of the TSP which is in a certain sense easier, since it can be efficiently approximated.

# Rudrata cycle $\rightarrow$ TSP

If  $\alpha$  is **large**, then the resulting **instance** of the **TSP** may not satisfy the triangle inequality, and has another important property.

This important **gap** property implies that, unless **P = NP**, no **approximation algorithm** is possible.

12. Any Problem  $\rightarrow$  SAT

home reading!

# Referred Materials

- Content of this lecture comes from Section 8.2, 8.3 in [DPV07], and Section 34.2, 34.3 in [CLRS09].