



NUS
National University
of Singapore

| **Computing**

CS3230

eric_han@nus.edu.sg
<https://eric-han.com>

Computer Science

T12 – Week 13

Reductions and Com. Complexity (cont.)

CS3230 – Design and Analysis of Algorithms

- 1 Please check:
 - » Tutorials - Attendance & Participation
 - » Assignments - Best Seven
- 2 Last lecture this week – Revision & Gradient Descent.
- 3 Some parting advice: \$ is important but not everything.
 - » GES 2016: \$3500, \$4000, \$5000
 - » CS3230 skills can help you land a good technical job.
- 4 FYP - AI & Machine Learning
- 5 CS3230 Practical Exercises

GES 2024

NUS GES 2024 Numbers

- › Bachelor of Computing (Computer Science)
 - » 25/50/75 percentile: \$6500, \$5600, \$7500
 - » Mean: \$6788
 - » Employed: 89.1%

Student Feedback on Teaching (SFT)

NUS Student Feedback <https://blue.nus.edu.sg/blue/>:

- › Don't Mix module/grading/project feedback - **feedback only for teaching.**
- › Feedback is confidential to university and anonymous to us.
- › Feedback is optional but highly encouraged.
- › Past student feedback improves teaching; see <https://www.eric-han.com/teaching>
 - ›› ie. Telegram access, More interactivity.
- › Your feedback is important to me, and will be used to improve my teaching.
 - ›› Good > Positive feedback > Encouragement
 - Teaching Awards (nominate)
 - Steer my career path
 - ›› Bad > Negative feedback (nicely pls) > Learning
 - Improvement
 - Better learning experience

› **P**: solvable in polynomial time

› **NP**: verifiable in polynomial time

Prove SOMETHING is in NP:

- 1 Provide a certificate for 'Yes' instance.
- 2 Verify certificate in polynomial time.

› **NP-hard**: polynomial-time reducible from all NP problems

Prove SOMETHING is in NP-hard:

- 1 Show it's at least as hard as a known NP-hard problem.
- 2 Show reduction: A-PROVEN-NP-HARD-PROBLEM \leq_p SOMETHING .

› **NP-complete**: both in NP and NP-hard

Prove SOMETHING is in NP-complete:

- 1 Prove it's in NP.
- 2 Prove it's NP-hard.

Which of the following imply $P = NP$?

- a. There is a problem in P that is also NP-complete
- b. There is a problem in P that is also in NP
- c. There is a problem in NP that is also NP-hard

Answer 1a

- › Suppose there exists a problem in P that is also in NP-complete
- › Since all NP-complete problems reduce to each other,
- › **every** NP problem would also be solvable in polynomial time.
- › So $P = NP$.

However, as of 2024, nobody has proven this yet!

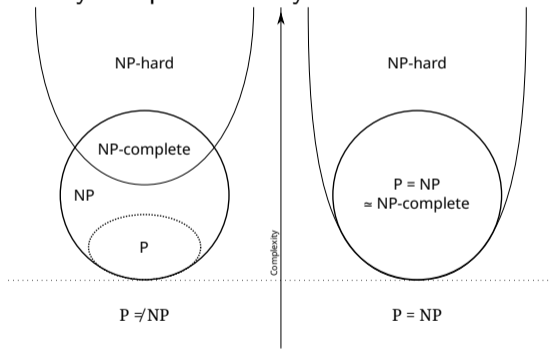


Figure 1: NP-complete with different assumptions ([Wiki](#))

Answer 1b

- › Every problem in P is also in NP (i.e., $P \subseteq NP$).
- › $P \subseteq NP$ does **not** imply $P = NP$.

Answer 1b

- › Every problem in P is also in NP (i.e., $P \subseteq NP$).
- › $P \subseteq NP$ does **not** imply $P = NP$.

Answer 1c

- › A problem in NP and NP -hard is a NP -complete problem.
- › Does **not** imply $P = NP$.

Given a multiset S of n integers (usually non-negative), $S = \{S_1, S_2, \dots, S_n\}$, and a target integer W . Is there exists a subset $I \subseteq \{1, 2, \dots, n\}$ such that $\sum_{i \in I} S_i = W$?

(See Subset Sum on Visualgo)

Example

- › Given $n = 5$, $S = \{5, 1, 5, 1, 4\}$, and $W = 7$
- › YES-instance, with certificate indices $\{0, 1, 3\}$ (values $\{5, 1, 1\}$), summing to 7.

We want to prove SUBSET-SUM is NP-complete.

Prove that SUBSET-SUM is in NP.

Prove that SUBSET-SUM is in NP.

Answer

- › **Certificate:** Subset I itself (the indices of S summing to W).
- › **Verification:** Check if $\sum_{i \in I} S_i = W$ in $O(n)$ time.

Prove that SUBSET-SUM is in NP.

Answer

- › **Certificate:** Subset I itself (the indices of S summing to W).
- › **Verification:** Check if $\sum_{i \in I} S_i = W$ in $O(n)$ time.

Pro-tips

- 1 Do **NOT** leave this question blank in the final.
- 2 Your verifier only needs to run in polynomial time w.r.t. input size—it does **NOT** need to be the fastest possible.

Prove that `SUBSET-SUM` is NP-hard.

Hint: Reduce from `3-SAT` ([see on Visualgo](#)).

Prove that **SUBSET-SUM** is NP-hard.

Hint: Reduce from **3-SAT** ([see on Visualgo](#)).

Answer

Given a **3-SAT** instance, we can reduce it to a **SUBSET-SUM** instance.

We build this reduction incrementally, with different attempts:

- 1 Modelling just variables.
- 2 Modelling Variables and Constraints.
- 3 Modelling Variables, Constraints, and Literals

Attempt 1: Modelling just variables

Given a **3-SAT** instance with variables $\{x_1, \dots, x_N\}$, we use symbols (literals) v_i and v'_i :

- › v_i : assign $x_i = \mathbf{TRUE}$
- › v'_i : assign $x_i = \mathbf{FALSE}$

To ensure single assignment per variable, we use a base-10 bitmask:

- › Set target sum W as a bitmask $1 \dots 1_{10}$ (length N).
- › Each literal corresponds to an integer in set S , with the appropriate bit set to 1.

This guarantees each variable is assigned exactly once but doesn't yet consider clause constraints.

We need to consider constraints.

Example

	x_1	x_2	x_3	Value in S	Meaning
$v_1 =$	1	0	0	100_{10}	$x_1 = \text{TRUE}$
$v'_1 =$	1	0	0	100_{10}	$x_1 = \text{FALSE}$
$v_2 =$	0	1	0	10_{10}	$x_2 = \text{TRUE}$
$v'_2 =$	0	1	0	10_{10}	$x_2 = \text{FALSE}$
$v_3 =$	0	0	1	1_{10}	$x_3 = \text{TRUE}$
$v'_3 =$	0	0	1	1_{10}	$x_3 = \text{FALSE}$
$W =$	1	1	1	111_{10}	

Table 1: Example with three variables $\{x_1, x_2, x_3\}$: Integers $S = \{100_{10}, 100_{10}, 10_{10}, 10_{10}, 1_{10}, 1_{10}\}$ with target sum $W = 111_{10}$.

Attempt 2: Modelling Variables and Constraints

Now, include the constraints (clauses) in the **3-SAT** formula Φ :

- › When assigning TRUE/FALSE for x_i ,
- › we can know which clause will be true.

Use the bitmask approach again,

- › setting a bit to 1 if literal satisfies clause C_j .
- › In polynomial time, we can verify all M clauses have at least one TRUE literal.

However, determining a suitable W for clauses (C_1, \dots, C_M) isn't straightforward, as the number of satisfied literals varies (1–3 per clause).

We can't test multiple values of W within **SUBSET-SUM** (eg. Testing 1231 vs. 3212).

Example

	x_1	x_2	x_3	C_1	C_2	C_3	C_4	Value in S	Meaning
$v_1 =$	1	0	0	1	0	0	1	1001001_{10}	
$v'_1 =$	1	0	0	0	1	1	0	1000110_{10}	$x_1 = \text{FALSE}$ satisfy C_2 and C_3
$v_2 =$	0	1	0	0	0	0	1	100001_{10}	
$v'_2 =$	0	1	0	1	1	1	0	101110_{10}	$x_2 = \text{FALSE}$ satisfy $C_1, C_2,$ and C_3
$v_3 =$	0	0	1	0	0	1	1	10011_{10}	$x_3 = \text{TRUE}$ satisfy C_3 and C_4
$v'_3 =$	0	0	1	1	1	0	0	11100_{10}	
$W =$	1	1	1	?	?	?	?	$111????_{10}$	$C_1/C_2/C_3/C_4$ has 1/2/3/1 satisfied literals

Table 2: Example of **3-SAT** from CLRS:

$\Phi = (x_1 \vee \overline{x_2} \vee \overline{x_3}) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_3}) \wedge (\overline{x_1} \vee \overline{x_2} \vee x_3) \wedge (x_1 \vee x_2 \vee x_3)$. This is a YES-instance with certificate $x_1 = x_2 = \text{FALSE}$ and $x_3 = \text{TRUE}$.

Attempt 3: Modelling Variables, Constraints, and Literals

For each of the M clauses in the **3-SAT** instance, introduce two slack symbols:

- › s_j : adds slack +1 to clause C_j
- › s'_j : adds slack +2 to clause C_j

This ensures variables, constraints, and literals are fully modelled, and sets a clear target sum for each clause, completing the reduction to **SUBSET-SUM**.

Example

	x_1	x_2	x_3	C_1	C_2	C_3	C_4	Value in S	Meaning
$v_1 =$	1	0	0	1	0	0	1	1001001_{10}	
$v'_1 =$	1	0	0	0	1	1	0	1000110_{10}	$x_1 = \text{FALSE}$ satisfy C_2 and C_3
$v_2 =$	0	1	0	0	0	0	1	100001_{10}	
$v'_2 =$	0	1	0	1	1	1	0	101110_{10}	$x_2 = \text{FALSE}$ satisfy $C_1, C_2,$ and C_3 $x_3 = \text{TRUE}$ satisfy C_3 and C_4
$v_3 =$	0	0	1	0	0	1	1	10011_{10}	
$v'_3 =$	0	0	1	1	1	0	0	11100_{10}	
$s_1 =$	0	0	0	1	0	0	0	1000_{10}	Take both +1 slack and +2 slacks for C_1
$s'_1 =$	0	0	0	2	0	0	0	2000_{10}	
$s_2 =$	0	0	0	0	1	0	0	100_{10}	
$s'_2 =$	0	0	0	0	2	0	0	200_{10}	Take only +2 slacks for C_2 Take only +1 slack for C_3
$s_3 =$	0	0	0	0	0	1	0	10_{10}	
$s'_3 =$	0	0	0	0	0	2	0	20_{10}	
$s_4 =$	0	0	0	0	0	0	1	1_{10}	Take both +1 slack and +2 slacks for C_4
$s'_4 =$	0	0	0	0	0	0	2	2_{10}	
$W =$	1	1	1	4	4	4	4	1114444_{10}	$C_1/C_2/C_3/C_4$ has target 4/4/4/4

Table 3: Example of 3-SAT from CLRS.

Summary

We've reduced (any) Φ from **3-SAT** to a corresponding **SUBSET-SUM** instance. Some technical details (omitted, see CLRS 34.5.5) include:

- › Reduction runs in polynomial time.
- › YES-instance of **3-SAT** \implies YES-instance of **SUBSET-SUM**.
- › YES-instance of **SUBSET-SUM** \implies YES-instance of **3-SAT**.

Pro-tip

This detailed proof seems long, but NP-completeness proofs can be short—don't skip such questions! (See next question.)

- › **Undirected Bipartite Graph:** $G = (L \cup R, E)$
Bipartite graph has disjoint vertex sets L, R with edges between L and R .
- › **Siblings:** $u, v \in L$
If there exists a vertex $r \in R$ such that edges (u, r) and (v, r) both exist.
- › **Family:** $F \subseteq L$
A subset is a **family** if for all distinct vertices in $u, v \in F$ are siblings.
- › **Decision Problem (FIND-FAMILY):**
Given a bipartite graph $G = (L \cup R, E)$ and integer k , does there exist a **family** of size $\geq k$?

Example

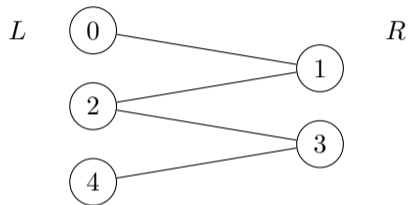


Figure 2: Example bipartite graph with $L = \{0, 2, 4\}$ and $R = \{1, 3\}$. Here, 0 and 2 are siblings, 2 and 4 are siblings, but $\{0, 2, 4\}$ is not a family since 0 and 4 are not siblings.

Prove that `FIND-FAMILY` is in NP.

Prove that `FIND-FAMILY` is in NP.

Answer

- › **Certificate:** The family set F itself.
- › **Verification:** For each pair $u, v \in F$, check if there exists $r \in R$ adjacent to both. Runs in $O(|F| \cdot |R|) = O(|L|^2 \cdot |R|)$ — polynomial in input size.

Prove that `FIND-FAMILY` is NP-hard.

Prove that `FIND-FAMILY` is NP-hard.

Answer

- 1 Decide which ... NP-hard problem to use.
- 2 Given ..., we transform to `FIND-FAMILY`

Decide which ... NP-hard problem to use

We aim to solve an instance of another NP-hard problem using **FIND-FAMILY**.

Family definition: A subset $F \subseteq L$ is a **family** if every pair $u, v \in F$ are siblings.

Problem	Condition on subset F
FIND-FAMILY	'... for every pair of vertices in F , they are <i>siblings</i> '
CLIQUE	'... for every pair of vertices in F , they are <i>adjacent</i> '

Try reduction: **CLIQUE** \leq_p **FIND-FAMILY**

Given ..., we transform to **FIND-FAMILY**

Polynomial time reduction

- 1 Given a **CLIQUE** instance $G = (V, E)$: is there a clique of size $\geq k$?
 - 2 Construct a **FIND-FAMILY** instance with $L = V$ and $R = E$.
 - 3 For each edge $(u, v) \in E$, connect u and v in L to node (u, v) in R .
- Total time to build the bipartite graph is polynomial $O(|V| + |E|)$.

Proof **CLIQUE** \iff **FIND-FAMILY**

Straightforward to argue YES-instance of **CLIQUE** corresponds (\iff) to a YES-instance of **FIND-FAMILY** :

- › (\implies) Suppose G has a clique of size k
- › (\impliedby) Suppose there is a family of size k in the bipartite graph. ...

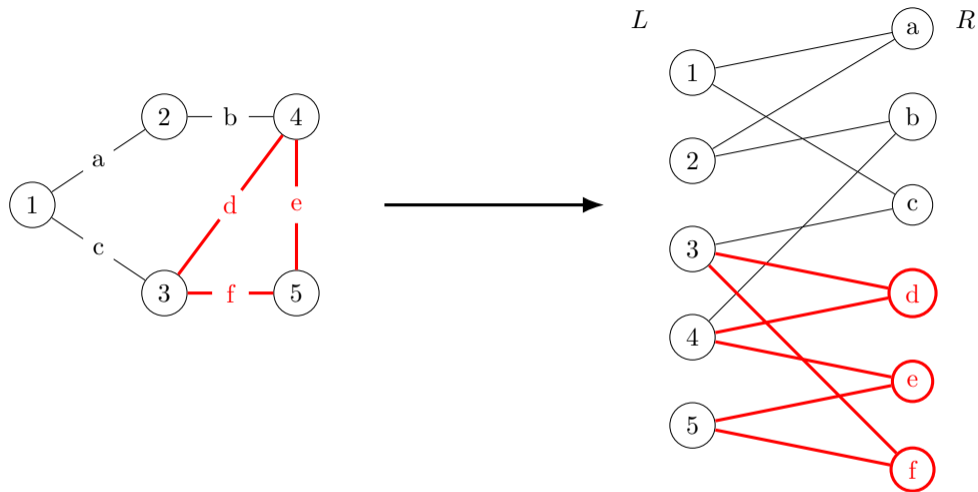


Figure 3: Edges d, e, f (in red) form a size-3 clique $\{3, 4, 5\}$ in G , which corresponds to a size-3 family $F = \{3, 4, 5\}$ in the `FIND-FAMILY` instance.

Give yourself a pat on the back for completing CS3230:

- › It has been a privilege teaching you this semester!
- › Where to go from here:
 - ›› Hate theory: possibly in your life, most difficult theory course
 - ›› Dislike theory, love concepts: CS3233 - Competitive Programming, ICPC
 - ›› Love theory:
 - CS5339 - Theory and Algorithms for Machine Learning ([scarlett](#))
 - Algorithms & Theory focus area