# Introduction to Algorithms and Data Structures Tutorial 9

your tutor

University of Edinburgh

10th-14th March, 2025

Q1(a) Show that SAT and 3-SAT are in NP

- Solution/certificate" for a SAT instance φ = C<sub>1</sub> ∧...C<sub>m</sub> will be an assignment to the logical variables {x<sub>1</sub>,...,x<sub>n</sub>}
- Assignment can be written as binary string b<sub>1</sub>...b<sub>n</sub> of length n (so polynomial in the size of the input formula φ).

• To verify  $b_1 \dots b_n$  against  $\phi$ , verify each  $C_j$  in turn

- For each literal l = xi in Cj, check whether bi = 1?
   For each literal l = xi in Cj, check whether bi = 0?
   As long as one of these tests passes, Cj is satisfied by b1...bn.
   It takes O(|Ci|) "lookups" to check Ci.
- ► To check all  $C_j$  are satisfied takes  $O\left(\sum_{j=1}^m |C_j|\right)$  lookups in total, which is polynomial-time, in the size of our input instance  $\phi$ .

Hence  ${\rm SAT}$  is in NP. Same argument works for  $3\mathchar`-{\rm SAT}.$ 

Q1(b) Show that SAT  $\leq_P 3$ -SAT.

Instance  $\Phi = C_1 \land \ldots \land C_m$  of SAT, over boolean variables  $\{x_1, \ldots, x_n\}$ .

Assume that no  $C_j$  includes any *complementary pair* of literals  $x_i$ ,  $\bar{x}_i$  (these would be trivially satisfied - just delete the clause).

Will convert  $\Phi$  to an equivalent 3-SAT formula "clause by clause". Take into account the size of each  $C_j$ . 4 cases:

► 
$$|C_j| = 1$$

$$|C_j| = 2$$

•  $|C_j| = 3 \dots$  just leave these exactly as they are!

Q1(b) SAT  $\leq_P$  3-SAT cont'd.

 $|C_j| = 1$ : Suppose  $C_j = (\ell_{j,1})$  for that specific literal  $\ell_{j,1}$ . Create *two* dummy variables  $y_{j,1}, y_{j,2}$ , then we will replace  $C_j$  by the following four clauses:

$$C_{j,1} = (\ell_{j,1} \lor y_{j,1} \lor y_{j,2}) \qquad C_{j,2} = (\ell_{j,1} \lor \overline{y_{j,1}} \lor y_{j,2}) \\ C_{j,3} = (\ell_{j,1} \lor y_{j,1} \lor \overline{y_{j,2}}) \qquad C_{j,4} = (\ell_{j,1} \lor \overline{y_{j,1}} \lor \overline{y_{j,2}})$$

Claim: The " $\land$ " of these 4 clauses is equivalent to  $C_j = (\ell_{j,1})$ .

Why? Note that *no matter what* values get assigned to  $y_{j,1}, y_{j,2}$ , there will be one pairing (from the options  $\{y_{j,1}, \overline{y_{j,1}}\} \times \{y_{j,2}, \overline{y_{j,2}}\}$ ) that has *both of* the  $y_{j,.}$  literals fail. So this clause will force  $\ell_{j,1}$  to be true, equivalent to  $C_j$ .

Q1(b) SAT  $\leq_P$  3-SAT cont'd.

 $|C_j| = 2$ : Suppose  $C_j = (\ell_{j,1} \vee \ell_{j,2})$  for its two specific literals. Use *one* dummy variable  $y_i$  to replace  $C_j$  by the following two clauses:

$$C_{j,1} = (\ell_{j,1} \vee \ell_{j,2} \vee y_j), C_{j,2} = (\ell_{j,1} \vee \ell_{j,2} \vee \overline{y_j}).$$

For similar reasons to the  $|C_j| = 1$  case,

$$(\ell_{j,1} \vee \ell_{j,2} \vee y_j) \wedge (\ell_{j,1} \vee \ell_{j,2} \vee \overline{y_j})$$

is true in exactly the same circs as the original  $C_i$ .

Q1(b) SAT  $\leq_P$  3-SAT cont'd.

 $|C_j| = 2$ : Suppose  $C_j = (\ell_{j,1} \vee \ell_{j,2})$  for its two specific literals. Use *one* dummy variable  $y_i$  to replace  $C_j$  by the following two clauses:

$$C_{j,1} = (\ell_{j,1} \vee \ell_{j,2} \vee y_j), C_{j,2} = (\ell_{j,1} \vee \ell_{j,2} \vee \overline{y_j}).$$

For similar reasons to the  $|C_j| = 1$  case,

$$(\ell_{j,1} \vee \ell_{j,2} \vee y_j) \wedge (\ell_{j,1} \vee \ell_{j,2} \vee \overline{y_j})$$

is true in exactly the same circs as the original  $C_i$ .

Q1(b) SAT  $\leq_P$  3-SAT cont'd.

 $|C_j| > 3$ : We will add  $|C_j| - 3$  new dummy variables  $y_{j,2}, \dots y_{j,|C_j|-2}$  (so 1 dummy variable if  $|C_j|$  was 4, two dummy variables if  $|C_j|$  was 5, ...)

Suppose  $C_j = (\ell_{j,1} \vee \ell_{j,2} \vee \ldots \vee \ell_{j,|C_j|}).$ 

We then replace  $C_j$  with the following clauses  $C_{j,1}, \ldots, C_{j,|C_j|-2}$  defined as follows:

$$C_{j,i} = \begin{cases} (\ell_{j,1} \lor \ell_{j,2} \lor y_{j,2}) & i = 1\\ (\overline{y_{j,i}} \lor \ell_{j,i+1} \lor y_{j,i+1}) & i, 1 < i < |C_j| - 2\\ (\overline{y_{j,|C_j|-2}} \lor \ell_{j,|C_j|-1} \lor \ell_{j,|C_j|}) & i = |C_j| - 2 \end{cases}$$

Claim:  $C_{j,1} \wedge \ldots \wedge C_{j,|C_i|-2}$  is satisfiable  $\Leftrightarrow C_j$  is satisfiable.

Q1(b) SAT  $\leq_P$  3-SAT cont'd.

$$C_{j,i} = \begin{cases} (\ell_{j,1} \lor \ell_{j,2} \lor y_{j,2}) & i = 1\\ (\overline{y_{j,i}} \lor \ell_{j,i+1} \lor y_{j,i+1}) & i, 1 < i < |C_j| - 2\\ (\overline{y_{j,|C_j|-2}} \lor \ell_{j,|C_j|-1} \lor \ell_{j,|C_j|}) & i = |C_j| - 2 \end{cases}$$

Claim:  $C_{j,1} \wedge \ldots \wedge C_{j,|C_j|-2}$  is satisfiable  $\Leftrightarrow C_j$  is satisfiable.

⇒ direction: Suppose  $C_{j,1} \land \ldots \land C_{j,|C_j|-2}$  is satisfiable. Let *i*\* be the first such that  $y_{j,i*} = 0$ . If *i*\* = 2, then  $\ell_{j,1} \lor \ell_{j,2}$  must be true ... if  $2 < i* \le |C_j| - 2$ , then  $\ell_{j,i*}$  must be true ... if  $y_{j,i} = 1$  for all *i*, then  $\ell_{j,|C_i|-1} \lor \ell_{j,|C_i|}$  must be true.

 $\Leftarrow$  direction: Choose any literal  $\ell_{j,i*}$  of  $C_j$  made true by the satisfying assignment. Set  $y_{j,i} = 1$  for i < i\* and  $y_{j,i} = 0$  for  $i \ge i*$ , this satisfies all  $C_{j,i}$ 

Note: Need "fresh" dummy variables for each  $C_i$  with  $|C_i|$ .

Q1(b) SAT  $\leq_P$  3-SAT cont'd.

We argued "equivalence" on a clause-by-clause basis.

By using "fresh" dummy variables for each  $C_j$ , this equivalence extends to the entire logical formula (no interdependence except among the original variables).

We have created an instance of 3-SAT of total size at most 12 times our original problem (if size is counted in "total number of literals").

Each of the conversions to 3-CNF are methodological, and can be done in time linear in the size of  $C_j$ .

Hence SAT  $\leq_P 3$ -SAT.

Want specific assignment to satisfy  $Y \ge \frac{7}{8}9$ , ie at least 8, of the clauses in the following  $\Phi$ .

$$\Phi = (x_1 \lor x_2 \lor x_3) \land (\bar{x_1} \lor \bar{x_2} \lor \bar{x_3}) \land (\bar{x_1} \lor x_2 \lor x_3) \land (x_1 \lor \bar{x_2} \lor \bar{x_3}) \land (x_1 \lor \bar{x_2} \lor x_4) \land (x_2 \lor x_3 \lor \bar{x_4}) \land (\bar{x_1} \lor \bar{x_3} \lor \bar{x_4}) \land (\bar{x_2} \lor \bar{x_3} \lor x_4) \land (\bar{x_1} \lor x_3 \lor x_4).$$

**variable**  $x_1$ : Two options:  $x_1 \leftarrow 0$  and  $x_1 \leftarrow 1$ .

To compute  $Exp_0 = E[Y | x_1 \leftarrow 0]$ , the expected number of satisfied clauses *conditional on*  $x_1$  *being 0*, we notice that  $\Phi$  has

- 4 clauses containing the negative literal  $\bar{x_1}$
- 3 clauses containing the positive literal x<sub>1</sub>
- 2 clauses not involving this variable at all.

$$\Phi = (x_1 \lor x_2 \lor x_3) \land (\bar{x_1} \lor \bar{x_2} \lor \bar{x_3}) \land (\bar{x_1} \lor x_2 \lor x_3) \land (x_1 \lor \bar{x_2} \lor \bar{x_3}) \land (x_1 \lor \bar{x_2} \lor x_4) \land (x_2 \lor x_3 \lor \bar{x_4}) \land (\bar{x_1} \lor \bar{x_3} \lor \bar{x_4}) \land (\bar{x_2} \lor \bar{x_3} \lor \bar{x_4}) \land (\bar{x_1} \lor \bar{x_3} \lor \bar{x_4}) \land (\bar{x_1} \lor x_3 \lor x_4).$$

variable  $x_1$ : Two options:  $x_1 \leftarrow 0$  and  $x_1 \leftarrow 1$ .

If we set  $x_1 \leftarrow 0$ , we satisfy all clauses with  $\bar{x_1}$ , we delete the  $x_1$  literal from the clauses containing it (probability drops to  $\frac{3}{4}$ )... the two other clauses still have probability  $\frac{7}{8}$ . This gives

$$\mathsf{E}[Y \mid x_1 \leftarrow 0] = 4 + 3 \cdot \frac{3}{4} + 2 \cdot \frac{7}{8} = 8.$$

To compute  $Exp_1 = E[Y | x_1 \leftarrow 1]$ , the circs for positive literals (3) and negative literals (4) are reversed, hence the value  $Exp_1$  can be computed as

$$\mathsf{E}[Y \mid x_1 \leftarrow 1] = 3 + 4 \cdot \frac{3}{4} + 2 \cdot \frac{7}{8} = 7.75.$$

Verdict:  $x_1 \leftarrow 0$ 

$$\Phi' = (\underset{\mathbf{x_1} \lor \mathbf{x_2} \lor \mathbf{x_3}}{(\mathbf{x_1} \lor \mathbf{x_2} \lor \mathbf{x_3})} \land (\overline{x_1} \lor \overline{x_2} \lor \overline{x_3}) \land (\overline{x_1} \lor \mathbf{x_2} \lor \mathbf{x_3}) \land (\mathbf{x_1} \lor \mathbf{x_2} \lor \mathbf{x_4}) \land (\mathbf{x_2} \lor \mathbf{x_3} \lor \mathbf{x_4}) \land (\overline{x_1} \lor \overline{x_3} \lor \overline{x_4}) \land (\overline{x_1} \lor \overline{x_3} \lor \overline{x_4}) \land (\overline{x_2} \lor \overline{x_3} \lor \mathbf{x_4}) \land (\overline{x_1} \lor x_3 \lor \mathbf{x_4})$$

4 clauses already satisfied. Only 5 "active" clauses remain.

Variable x<sub>2</sub>:

 $x_2 \leftarrow 0$ : we have  $\bar{x_2}$  in 3 active clauses (satisfied by  $x_2 \leftarrow 0$ ),  $x_2$  in one length-2 clause (length-1 after  $x_2 \leftarrow 0$ ), one length-3 clause (becomes length 1). Hence

$$\mathsf{E}[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 0] = 4 + 3 + 1 \cdot \frac{1}{2} + 1 \cdot \frac{3}{4} = 8.25$$

(the 4 is from the previously satisfied clauses).  $x_2 \leftarrow 1$ : For E[Y |  $x_1 \leftarrow 0, x_2 \leftarrow 1$ ], just observe

$$\mathsf{E}[Y \mid x_1 \leftarrow 0] = \frac{\mathsf{E}[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 0] + \mathsf{E}[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 1]}{2}.$$

By  $E[Y \mid x_1 \leftarrow 0] = 8$  and  $E[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 0] = 8.25$ , we know  $E[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 1] < E[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 0]$ . Verdict:  $x_2 \leftarrow 0$ 

$$\Phi' = (\underbrace{\mathbf{x_1} \lor \mathbf{x_2} \lor \mathbf{x_3}}_{(x_1 \lor \bar{x_2} \lor \bar{x_3}) \land (\bar{x_1} \lor x_2 \lor x_3) \land}_{(x_1 \lor \bar{x_2} \lor \bar{x_3}) \land (x_1 \lor \bar{x_2} \lor x_4) \land (\mathbf{x_2} \lor \mathbf{x_3} \lor \bar{x_4}) \land}_{(\bar{x_1} \lor \bar{x_3} \lor \bar{x_4}) \land (\bar{x_2} \lor \bar{x_3} \lor x_4) \land (\bar{x_1} \lor x_3 \lor x_4)}$$

7 clauses satisfied, just 2 active clauses.

Variable  $x_3$ : Both remaining clauses have  $x_3$  as a positive literal. Hence:

$$\begin{split} \mathsf{E}[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 0, x_3 \leftarrow 1] &= \mathbf{7} + 2 = \mathbf{9} \\ \mathsf{E}[Y \mid x_1 \leftarrow 0, x_2 \leftarrow 0, x_3 \leftarrow 0] &= \mathbf{7} + \mathbf{0} + \mathbf{1} \cdot \frac{1}{2} = \mathbf{7.5} \end{split}$$

Verdict:  $x_3 \leftarrow 1$ 

Choose either value for  $x_4$ , doesn't matter which. Overall assignment is  $x_1 \leftarrow 0, x_2 \leftarrow 0, x_3 \leftarrow 1, x_4 \in \{0, 1\}$ .

## Q3 (b): derandomization for general SAT

Suppose we want to do the same process for general CNF?

Obs 1:  $E[Y] = \frac{7}{8}m$  no longer fits.

- ► Can use *linearity of expectation* for E[Y], but clauses have variable length
  - Let m<sub>k</sub> be the number of clauses of length k in Φ (each k = 1,..., n)
    Then E[Y] is

$$E[Y] = \sum_{k=1}^{n} m_k (1 - \frac{1}{2^k})$$

Obs 2: Can do a derandomization to achieve at least E[Y] satisfied clauses.

- While calculating the E[Y | x<sub>1</sub> = b<sub>1</sub>,..., x<sub>j</sub> = b<sub>j</sub>] values, we will be working with a Φ' formula with clauses of varying sizes.
- Probability is  $(1 \frac{1}{2^k})$  for clauses with k > 3 active literals.
- Calculations are still feasible.

Won't necessarily satisfy  $\geq \frac{7}{8}m$  clauses, because E[Y] might have been smaller (especially if  $\Phi$  has a lot of 1-literal and/or 2-literal clauses).

- ▶  $\mathfrak{I}$  is an *Independent Set* of *G* if for every  $u \in \mathfrak{I}, v \in \mathfrak{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .
- K is a Vertex Cover of G if for every e = (u, v) ∈ E, either u ∈ K or v ∈ K.

- ▶ J is an *Independent Set* of G if for every  $u \in J, v \in J \setminus \{u\}$ , that  $(u, v) \notin E$ .
- ▶  $\mathcal{K}$  is a *Vertex Cover* of *G* if for every  $e = (u, v) \in E$ , either  $u \in \mathcal{K}$  or  $v \in \mathcal{K}$ .

**proof:** By definition, the set  $\mathcal{I}$  is an Independent set if (and only if) for every  $u \in \mathcal{I}, v \in \mathcal{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .

- ▶  $\mathfrak{I}$  is an *Independent Set* of *G* if for every  $u \in \mathfrak{I}, v \in \mathfrak{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .
- ▶  $\mathcal{K}$  is a *Vertex Cover* of *G* if for every  $e = (u, v) \in E$ , either  $u \in \mathcal{K}$  or  $v \in \mathcal{K}$ .

**proof:** By definition, the set  $\mathcal{I}$  is an Independent set if (and only if) for every  $u \in \mathcal{I}, v \in \mathcal{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .

This is the case if and only if for every  $(u, v) \in E$ , at least one of u, v is *not* in  $\mathfrak{I}$ .

- ▶  $\mathfrak{I}$  is an *Independent Set* of *G* if for every  $u \in \mathfrak{I}, v \in \mathfrak{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .
- ▶  $\mathcal{K}$  is a *Vertex Cover* of *G* if for every  $e = (u, v) \in E$ , either  $u \in \mathcal{K}$  or  $v \in \mathcal{K}$ .

**proof:** By definition, the set  $\mathcal{I}$  is an Independent set if (and only if) for every  $u \in \mathcal{I}, v \in \mathcal{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .

This is the case if and only if for every  $(u, v) \in E$ , at least one of u, v is *not* in  $\mathcal{I}$ . This is the case if and only if if for every  $(u, v) \in E$ , either  $u \in V \setminus \mathcal{I}$  or  $v \in V \setminus \mathcal{I}$ .

- ▶ J is an *Independent Set* of G if for every  $u \in J, v \in J \setminus \{u\}$ , that  $(u, v) \notin E$ .
- ▶  $\mathcal{K}$  is a *Vertex Cover* of *G* if for every  $e = (u, v) \in E$ , either  $u \in \mathcal{K}$  or  $v \in \mathcal{K}$ .

**proof:** By definition, the set  $\mathcal{I}$  is an Independent set if (and only if) for every  $u \in \mathcal{I}, v \in \mathcal{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .

This is the case if and only if for every  $(u, v) \in E$ , at least one of u, v is *not* in  $\mathfrak{I}$ . This is the case if and only if if for every  $(u, v) \in E$ , either  $u \in V \setminus \mathfrak{I}$  or  $v \in V \setminus \mathfrak{I}$ . This is the case (by definition) if and only if  $V \setminus \mathfrak{I}$  is a Vertex Cover for G.

- ▶ J is an *Independent Set* of G if for every  $u \in J, v \in J \setminus \{u\}$ , that  $(u, v) \notin E$ .
- ▶  $\mathcal{K}$  is a *Vertex Cover* of *G* if for every  $e = (u, v) \in E$ , either  $u \in \mathcal{K}$  or  $v \in \mathcal{K}$ .

**proof:** By definition, the set  $\mathcal{I}$  is an Independent set if (and only if) for every  $u \in \mathcal{I}, v \in \mathcal{I} \setminus \{u\}$ , that  $(u, v) \notin E$ .

This is the case if and only if for every  $(u, v) \in E$ , at least one of u, v is *not* in  $\mathfrak{I}$ . This is the case if and only if if for every  $(u, v) \in E$ , either  $u \in V \setminus \mathfrak{I}$  or  $v \in V \setminus \mathfrak{I}$ . This is the case (by definition) if and only if  $V \setminus \mathfrak{I}$  is a Vertex Cover for G.

Implications for the two decision problems:

*G* has an Independent Set  $\mathfrak{I}$  of size  $|\mathfrak{I}| \ge k$   $\Leftrightarrow$  *G* has a Vertex Cover  $V \setminus \mathfrak{I}$  such that  $|\mathfrak{I}| \ge k$  $\Leftrightarrow$  *G* has a Vertex Cover  $\mathfrak{K}$  such that  $|\mathfrak{K}| \le (n-k)$ .

- ▶ Very straightforward "reduction" from INDEPENDENT SET to VERTEX COVER, and vice versa.
- Really simple reduction, graph doesn't change, just flip between k and n-k for size parameter.

Therefore, INDEPENDENT SET is NP-complete  $\Leftrightarrow$  VERTEX COVER is NP-complete.

Rare to have "reductions" which work in both directions.

#### Implications for the approximation problems:

Hypothesis: We have an approximation algorithm for VERTEX COVER with approximation ration of  $\alpha$ , for  $\alpha > 1$ .

So the algorithm will return  $\ell$  satisfying  $\ell \leq \alpha \cdot OPT_{VC}(G)$ , where  $OPT_{VC}(G)$  is the optimum/minimum size of a VC for G.

What does  $n - \ell$  mean in relation to the maximum Independent Set for G? (remember maximum has size  $n - OPT_{VC}(G)$ ). We **know** that

$$n - \ell \geq n - \alpha \cdot OPT_{VC}(G) = (n - OPT_{VC}(G)) - (\alpha - 1)OPT_{VC}(G)$$

We would like  $n - \ell \geq \frac{1}{\beta} \left( n - OPT_{VC}(G) \right)$  for some  $\beta > 1$ 

But imagine  $\alpha$  of the Vertex Cover is  $\alpha = 2$ . Then taking  $n - \ell$  for Independent Set gives the bound

$$n - \ell \geq (n - OPT_{VC}(G)) - OPT_{VC}(G)$$

However, there may be graphs where  $OPT_{VC}(G)$  is n/2 or even greater.

So  $n - \ell$  may be arbitrarily close to 0.

Problem is the - in the conversion between the two problems: subtraction does not preserve approximation.