# Universal Algebra and Computational Complexity Lecture 2 

Ross Willard<br>University of Waterloo, Canada

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## Summary of Lecture 1

Recall from yesterday:

$$
\begin{array}{cccc}
L \subseteq P \subseteq & \text { PSPACE } & \subseteq \text { EXPTIME } \\
\Psi & \Psi & \Psi \\
F V A T H & 3 C O L & C L O
\end{array}
$$

Topics for today:

## Summary of Lecture 1

Recall from yesterday:


Topics for today:

- "Nondeterministic" complexity classes
- Reductions
- Complete problems


## "Nondeterministic polynomial time": an example

## Recall

## Graph 3-Colorability problem (3COL)

INPUT: a finite graph $G=(V, E)$.
QUESTION: Does $G$ have a 3 -coloring?

Recall that we only know $3 C O L \in E X P T I M E$ (and PSPACE).
Most complexity theorists conjecture that 3COL is not tractable.

## "Nondeterministic polynomial time": an example

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HOWEVER, if we are GIVEN a 3 -coloring of $G$, it is easy (tractable) to VERIFY the correctness of the 3 -coloring (and thus know that $G$ is 3-colorable).

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HOWEVER, if we are GIVEN a 3 -coloring of $G$, it is easy (tractable) to VERIFY the correctness of the 3 -coloring (and thus know that $G$ is 3-colorable).

Informally, 3COL is a projection of a problem in $P$.

## 3 COL as a projection of a problem in $P$

Identify 3COL with set

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\{G: 3 C O L \text { answers "YES" on input } G\} .
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Similarly with other decision problems.

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3 C O L-T E S T=\{(G, \chi): \chi \text { is a 3-coloring of } G\} .
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Clearly 3COL-TEST is tractable (in $\operatorname{TIME}\left(N^{2}\right)$, hence in $P$ ).
And

$$
G \in 3 C O L \Leftrightarrow \exists \chi[(G, \chi) \in 3 C O L-T E S T] .
$$

## Certificates for 3 COL

If $(G, \chi) \in 3 C O L-T E S T$, then we call $\chi$ a certificate for " $G \in 3 C O L$."
We say that:

- 3COL-TEST is a polynomial-time certifier for 3COL.
- 3COL is polynomial-time certifiable.
- 3COL is in Nondeterministic Polynomial Time (or NP).


## Nondeterministic Polynomial Time (NP)

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## More examples of NP problems

The following problems are all in NP (and not known to be in $P$ ). (1) $4 C O L, 5 C O L$, etc.

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The following problems are all in NP (and not known to be in $P$ ).
(1) $4 C O L, 5 C O L$, etc.
(2) SAT:

- INPUT: a boolean formula $\varphi$.
- QUESTION: is $\varphi$ satisfiable?
- Certificate: an assignment of values to the variables making $\varphi$ true.
- Polynomial-time certifier: given $(\varphi, \mathbf{c})$, decide if $\varphi(\mathbf{c})=1$ (i.e., $F V A L$ ).


## More examples of $N P$ problems

The following problems are all in NP (and not known to be in $P$ ).
(1) $4 C O L, 5 C O L$, etc.
(2) $S A T$ :

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(3) ISO:
- INPUT: two finite graphs $G_{1}, G_{2}$.
- QUESTION: are $G_{1}$ and $G_{2}$ isomorphic?
- Certificate: an isomorphism from $G_{1}$ to $G_{2}$.
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(c) HAMPATH:
- INPUT: a finite directed graph $G$.
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- Grad student reader can only move RIGHT.


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If $\square$ is a complexity class, then a decision problem $D$ is in $N \square$ iff there exists a decision problem $E$ in two inputs $(x, z)$, and there exists a certifying Turing machine $M$, such that

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- Moreover, $\forall(x, z), M$ decides whether $(x, z) \in E$ with resource usage as defined by $\square$, measured as a function of $N=$ the length of $x$.


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- Moreover, $\forall(x, z), M$ decides whether $(x, z) \in E$ with resource usage as defined by $\square$, measured as a function of $N=$ the length of $x$.
- Exercise: this defines NP equivalently.
- $N L=$ "Nondeterministic LOGSPACE"
- NSPACE = "Nondeterministic PSPACE"
- NEXPTIME = "Nondeterministic EXPTIME"


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Define
PATH-TEST $=\{(G, \pi): G$ is a directed graph with $V=\{0, \ldots, n-1\}$, and $\pi=\left(v_{0}, v_{1}, \ldots, v_{k}\right)$ is a path from 0 to 1 in $\left.G\right\}$

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Clearly PATH is a projection of PATH-TEST.

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While the certifying student traverses $\pi$, the R/W Tape 1 student copies and remembers the last two vertices traversed, and checks the input tape to see if they form an edge.

## Certifying PATH-TEST

We can build a certifying Turing machine which solves PATH-TEST ...
Input (ROM):

Certif. (ROM) R/W Tape 1:


While the certifying student traverses $\pi$, the R/W Tape 1 student copies and remembers the last two vertices traversed, and checks the input tape to see if they form an edge.

Only LOGSPACE (as a function of the length of the input $G$ ) is needed.

## Comparing deterministic and nondeterministic classes

Let $f: \mathbb{N} \rightarrow \mathbb{N}$ be "nice" and such that $f(N) \geq \log N$.

Theorem
(1) $\operatorname{TIME}(f(N)) \subseteq \operatorname{NTIME}(f(N))$ and similarly for SPACE.

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(3) (Savitch's Theorem): $\operatorname{NSPACE}(f(N)) \subseteq \operatorname{SPACE}\left(f(N)^{2}\right)$.

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Since PATH $\in N L$, Savitch's theorem shows PATH $\in \operatorname{SPACE}\left((\log N)^{2}\right)$.
(Our algorithm showed only that $\operatorname{PATH} \in \operatorname{SPACE}(\sqrt{N})$.)

## Summary of complexity classes

| $L \subseteq N L \subseteq P \subseteq N P \subseteq P S P A C E \subseteq E X P T I M E \subseteq \text { NEXPTIME }$ |  |  |
| :---: | :---: | :---: |
| $\begin{aligned} & \quad P A T H \\ & \text { FVAL, } \\ & 2 C O L \end{aligned}$ | 3 COL , | CLO |
|  | 4COL, etc. |  |
|  | SAT, |  |
|  | ISO, |  |
|  | HAMPATH |  |

## Summary of complexity classes



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## Summary of complexity classes



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## Summary of complexity classes



## Summary of complexity classes


$10^{6}$ USD prize (Clay Mathematics Institute) for answering $P \stackrel{?}{=} N P$.

## Reductions

Suppose $C, D$ are decision problems.
Suppose $f: C_{i n p} \rightarrow D_{\text {inp }}$ is a function.
We say that

$$
f \text { reduces } C \text { to } D,
$$

and write

$$
C \leq_{f} D,
$$

if for all $x \in C_{i n p}$,

$$
x \in C \Leftrightarrow f(x) \in D .
$$

## Picture of $C \leq_{f} D$



Intuition: if $C \leq_{f} D$, then

## Picture of $C \leq_{f} D$



Intuition: if $C \leq_{f} D$, then

- Algorithms for $D$ and $f$ can be used to solve $C$.
- Hence $D$ is at least as hard as $C$ (modulo the cost of computing $f$ ).


## Example

Recall the problems $3 C O L$ and $S A T$ :
3COL
INPUT: a finite graph $G=(V, E)$.
QUESTION: is G 3-colorable?

## SAT

INPUT: a boolean formula $\varphi$. QUESTION: is $\varphi$ satisfiable?

Let's find a function $f$ which reduces $3 C O L$ to $S A T$.

## A reduction of $3 C O L$ to $S A T$

Given a finite graph $G=(V, E)$, we want a boolean formula $\varphi_{G}$ such that $G$ is 3 -colorable $\Leftrightarrow \varphi_{G}$ is satisfiable.

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Given a finite graph $G=(V, E)$, we want a boolean formula $\varphi_{G}$ such that $G$ is 3-colorable $\Leftrightarrow \varphi_{G}$ is satisfiable.

- The variables of $\varphi_{G}$ will be all $x_{V}^{\mathbf{c}} \quad(v \in V, \mathbf{c} \in\{\mathbf{r}, \mathbf{g}, \mathbf{b}\})$. - Think of $x_{v}^{c}$ as representing the assertion " $v$ is colored $\mathbf{c}$."


## A reduction of $3 C O L$ to SAT

Given a finite graph $G=(V, E)$, we want a boolean formula $\varphi_{G}$ such that $G$ is 3-colorable $\Leftrightarrow \varphi_{G}$ is satisfiable.

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- Think of $x_{v}^{\mathbf{c}}$ as representing the assertion " $v$ is colored $\mathbf{c}$."
- For each $v \in V$ let $\alpha_{v}$ be the formula " $v$ has exactly one color," i.e.,

$$
\left(x_{v}^{\mathbf{r}} \vee x_{v}^{\mathbf{g}} \vee x_{v}^{\mathbf{b}}\right) \wedge \neg\left(x_{v}^{\mathbf{r}} \wedge x_{v}^{\mathbf{g}}\right) \wedge \neg\left(x_{v}^{\mathbf{r}} \wedge x_{v}^{\mathbf{b}}\right) \wedge \neg\left(x_{v}^{\mathbf{b}} \wedge x_{v}^{\mathbf{b}}\right)
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## A reduction of $3 C O L$ to $S A T$

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- For $v, w \in V$ let $\beta_{v, w}$ be the formula " $v$ and $w$ have different colors," i.e.,

$$
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## A reduction of $3 C O L$ to SAT

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$$

- Let

$$
\varphi_{G}=\left(\bigwedge_{v \in V} \alpha_{v}\right) \wedge\left(\bigwedge_{(v, w) \in E} \beta_{v, w}\right)
$$

This clearly works.

## Picture of $3 \mathrm{COL} \leq_{f}$ SAT

Define $f: G \mapsto \varphi_{G}$. Then $3 C O L \leq_{f} S A T$.


Formulas

## Picture of $3 C O L \leq_{f} S A T$

Define $f: G \mapsto \varphi_{G}$. Then $3 C O L \leq_{f} S A T$.


Formulas

Thus SAT is at least as hard as $3 C O L$, modulo the cost of computing $\varphi_{G}$.
What is the cost of computing $\varphi_{G}$ ?

## Computing $f$ with a functional Turing machine

Idea: replace the output bit with an output write-only tape.


At the start.

## Computing $f$ with a functional Turing machine

Idea: replace the output bit with an output write-only tape.


At the end.

## Computing $f$ with a functional Turing machine

Idea: replace the output bit with an output write-only tape.


Exercise: Can compute $\varphi_{G}$ from $G$ in $\operatorname{TIME}\left(N^{2}\right)$ and $\operatorname{SPACE}(\log N)$.

## Complexity of computing $f$

In general:

## Definition

a functional Turing machine is a Turing machine whose output bit is replaced by an output tape (write-only).

- Output tape grad student can only move RIGHT.

Let $C, D$ be decision problems with appropriately encoded input sets $C_{i n p}, D_{i n p}$ respectively.

## Definition

A function $f: C_{i n p} \rightarrow D_{\text {inp }}$ is computed by a functional Turing Machine $M$ if whenever $M$ is started with input $x \in C_{i n p}$, it eventually halts with $f(x)$ written on its output tape.

## $X$-computable functions

Let $X$ be a complexity class (such as $P, L$, etc.).

## Definition

We say that a function $f: C_{i n p} \rightarrow D_{i n p}$ is computable in $X$ if there exists a functional Turing Machine which computes $f$ and on input $x$ requires no more resources than those permitted by the definition of $X$.

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Example: the function $f: G \mapsto \varphi_{G}$ in our example showing $3 C O L \leq_{f} S A T$ is $P$-computable.

- (In fact, it is L-computable.)


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## Lemma

For any decent complexity class $X$, if $C \leq_{f} D \in X$ and $f$ is $X$-computable, then $C \in X$.

## X-reductions

Suppose $X, Y$ are complexity classes with $X \subseteq Y$.
Let $C, D$ be decision problems with $C, D \in Y$.

## Definition

(1) We say that $C$ reduces to $D(\bmod X)$ and write

$$
C \leq x D
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if there exists an $X$-computable function $f: C_{i n p} \rightarrow D_{\text {inp }}$ which reduces $C$ to $D$.

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if there exists an $X$-computable function $f: C_{i n p} \rightarrow D_{\text {inp }}$ which reduces $C$ to $D$.
(2) We write $C \equiv_{x} D$ if both $C \leq_{x} D$ and $D \leq_{x} C$.

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(2) We write $C \equiv_{x} D$ if both $C \leq_{x} D$ and $D \leq_{x} C$.

This turns the $\equiv x$-classes of $Y$ into a poset.
Most widely used when $X=P$.

## The picture of $N P(\bmod P)$

## Theorem

The poset (NP/ $\equiv_{P}, \leq_{P}$ ) has ...
(1) a least element (consisting of all the elements of $P$ ), and
(2) (S. Cook, '71; L. Levin, '73) a greatest element, namely, the $\equiv_{p}$-class containing SAT.


Jargon: SAT is NP-complete (for $\leq_{P}$ reductions).

## Definition

A decision problem $D$ is $N P$-complete if:

- $D \in N P$, and
- $C \leq_{p} D$ for all $C \in N P$.

Equivalently (by Cook-Levin), $D$ is $N P$-complete iff $D \equiv{ }_{P} S A T$.


## Karp's Theorem

## Theorem (R. Karp, '72)

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Examples:

- 3COL, 4COL, etc.
- HAMPATH
- 3SAT (the restriction of SAT to formulas in CNF, each conjunct being a disjunction of at most 3 literals)
(Exercise: check that our proof we gave for $3 C O L \leq_{p} S A T$ also shows $3 C O L \leq_{P} 3 S A T$.)


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## Theorem (R. Ladner, '75)

If $P \neq N P$, then $\left|N P / \equiv_{P}\right| \geq 3$.

In fact, if $P \neq N P$, then $N P / \equiv_{P}$ is order dense.

## The picture of EXPTIME $(\bmod P)$



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- (H. Friedman '82, unpubl.; C. Bergman, D. Juedes \& G. Slutzki, '99) CLO is EXPTIME-complete (for $\leq_{p}$ reductions).


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- (D. Kozen, '77) 1-CLO is PSPACE-complete (for $\leq_{P}$ reductions).


## The picture of $N P(\bmod L)$



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- (R. Ladner, '75) CVAL is . . $\quad P$-complete (for $\leq_{L}$ reductions).
- (???) HORN-SAT and HORN-3SAT are also P-complete.


## Summary

| $L$ $U$ | $\begin{aligned} & N L \\ & ש \end{aligned}$ | $\subseteq \begin{gathered} P \\ ש \end{gathered}$ | $\underset{\Psi}{N P} \subseteq$ | PSPACE <br> ${ }^{*}$ | $\subseteq \underset{u}{\subseteq}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| FVAL, | PATH, | CVAL, | SAT, | 1-CLO | CLO |
| 2 COL | 2SAT | HORN- | 3SAT, |  |  |
|  |  | 3SAT | 3COL, |  |  |
|  |  |  | 4COL, etc. |  |  |
|  |  |  | HAMPATH |  |  |

Moreover, each problem listed above is "hardest in its class," i.e., is complete with respect to either $\leq_{P}$ or $\leq_{L}$ reductions.

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Moreover, each problem listed above is "hardest in its class," i.e., is complete with respect to either $\leq_{P}$ or $\leq_{L}$ reductions.

In Thursday's lecture: some problems from universal algebra.

