

# CoCoNuT - Complexity - Lecture 3

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# Outline

Complementation of Languages

Search Problems

Average and Worst Case

# co-C Recap

## Complimentation of Languages

Let  $\mathcal{L}$  be a language, we define  $\overline{\mathcal{L}}$  as

$$\overline{\mathcal{L}} = \{x \in \Sigma^* : x \notin \mathcal{L}\}.$$

## co Classes

If  $\mathcal{C}$  is a class of languages then co-C is the set

$$\text{co-C} = \{\overline{\mathcal{L}} : \mathcal{L} \in \mathcal{C}\}$$

We have  $P = \text{co-P}$

# NP and co-NP

Recall NP is the set of languages which have short proofs.

So co-NP is the set of languages which have short refutations.

Our earlier example FACTORING was in NP and co-NP.

It is believed that  $\text{NP} \cap \text{co-NP}$  is larger than P

- ▶ e.g.  $\text{FACTORING} \in \text{NP} \cap \text{co-NP}$  but we suspect that  $\text{FACTORING} \notin \text{P}$ .

It is unknown whether  $\text{NP} = \text{co-NP}$ .

- ▶ Take SAT formulae  $\phi$ .
- ▶ Easy to present a proof that  $\phi$  is satisfiable.
- ▶ Hard to see how to present a proof that  $\phi$  is not-satisfiable.

# Recall Definition of NP

Go back to our witness definition of NP.

NP

$\mathcal{L} \in \text{NP}$  means  $\exists V$  s.t. :  $x \in \mathcal{L}$  iff  $\exists w \in \{0, 1\}^{p(|x|)}$  s.t.  $V(x, w) = 1$ .

# Alternative Definition of co-NP

We can define co-NP via

co-NP: Defn 1

$$\text{co-NP} = \{\overline{\mathcal{L}} : \mathcal{L} \in \text{NP}\}$$

or via

co-NP: Defn 2

$\mathcal{L} \in \text{co-NP}$  means  $\exists V$  s.t. :  $x \in \mathcal{L}$  iff  $\forall w \in \{0, 1\}^{p(|x|)}$  s.t.  $V(x, w) = 1$ .

We will now show that both definitions are equivalent:

# Alternative Definition of co-NP

## Defn 2 $\implies$ Defn 1

Suppose  $\mathcal{L}$  a language such that

$$y \in \mathcal{L} \text{ iff } \forall w \in \{0, 1\}^{p(|y|)} \text{ s.t. } V(y, w) = 1.$$

Let  $x \in \overline{\mathcal{L}}$

- ▶ Implies  $\exists w$  such that  $V(x, w) = 0$ .
- ▶ Define the machine  $V'(x, w) = 1 - V(x, w)$ .
- ▶ Then  $V'$  accepts the language  $\overline{\mathcal{L}}$  thus  $\overline{\mathcal{L}} \in \text{NP}$ .
- ▶ i.e.  $\mathcal{L} = \overline{\overline{\mathcal{L}}}$  satisfies Defn 1.

# Alternative Definition of co-NP

## Defn 1 $\implies$ Defn 2

Now suppose  $\mathcal{L}$  is such that

$$\overline{\mathcal{L}} \in \text{NP}$$

Let  $x \in \mathcal{L}$

- ▶  $\exists V'$  and  $\forall w$  s.t.  $V'(x, w) = 0$ .
  - ▶ Otherwise  $V'$  would accept a false proof.
- ▶ Now define  $V(x, w) = 1 - V'(x, w)$
- ▶ Then  $\forall w$  we have  $V(x, w) = 1$ .



# The Polynomial Hierarchy

We can play games with  $\exists$  and  $\forall$  for a long time

## The class $\Sigma_i$

$\mathcal{L} \in \Sigma_i$  if there is a verifier  $V$  such that  $\exists V$  such that

$$x \in \mathcal{L} \text{ iff } \exists w_1 \in \{0, 1\}^{p(|x|)}, \forall w_2 \in \{0, 1\}^{p(|x|)}, \exists w_3 \in \{0, 1\}^{p(|x|)}, \\ \dots Q_i w_i \in \{0, 1\}^{p(|x|)}, \text{ s.t. } V(x, w_1, w_2, \dots, w_i) = 1.$$

where  $Q_i = \exists$  if  $i$  is odd and  $Q_i = \forall$  if  $i$  is even.

The polynomial hierarchy is defined by

$$\text{PH} = \cup_{i \geq 1} \Sigma_i.$$

Believed that  $\Sigma_i \neq \Sigma_{i+1}$ .

# The Polynomial Hierarchy

## Example

Given a graph  $G$  and an integer  $k$  determine if the largest clique in  $G$  has size exactly  $k$ .

- ▶ In a clique every two vertices are connected by an edge.

Call this problem LARGEST-CLIQUE

A “witness” for this problem would be

- ▶ A set of  $k$  vertices  $S$  s.t.  $S$  is a clique
- ▶ For all sets of  $k + 1$  vertices  $S'$  we have  $S'$  is not a clique

Thus LARGEST-CLIQUE  $\in \Sigma_2$ .

# QSAT<sub>*i*</sub>

Given a boolean formulae  $\phi$  in  $n$  variables  $x_1, \dots, x_n$  which are partitioned into  $i$  sets  $X_1, \dots, X_i$ .

## Quantified SAT, or QSAT<sub>*i*</sub>

QSAT<sub>*i*</sub> is the problem to determine whether the following statement is true.

$$\exists X_1 \forall X_2 \exists X_3 \forall X_4 \dots Q_i X_i \phi,$$

where  $Q_i = \exists$  if  $i$  is odd and  $Q_i = \forall$  if  $i$  is even.

QSAT<sub>*i*</sub> is  $\Sigma_i$ -complete.

# Search vs Decision

Up to now we have focused on **decision problems**.

This can seem a bit artificial, but it is the tradition of complexity theory.

- ▶ Nice philosophy about theorems, proofs etc
- ▶ Simplifies the discussions a lot
- ▶ Generalises to other questions
- ▶ Useful notion in cryptography.

In practice we care about **search problems**

- ▶ We want to **find** the variables which satisfy SAT
- ▶ We want to **find** the factors of a number  $N$
- ▶ ...

We can build up virtually the same theory for search problems:

- ▶ See Goldreich's book for an extensive study on this.

# Search Problems

Let  $R$  denote a (poly-bounded) relation

- ▶ Set of values  $(x, y) \in \{0, 1\}^* \times \{0, 1\}^*$
- ▶ Such that  $|y| < p(|x|)$  for some poly  $p$ .

Think of  $x$  as the problem and  $y$  the solution.

The class  $P_{\text{search}}$  is the set of relations  $R$  such that

- ▶ There is an algorithm  $A$  (depending on  $R$ )
- ▶ For all  $x$  the algorithm  $A(x)$  outputs  $y$
- ▶ Such that  $(x, y) \in R$  or  $y = \perp$  and there is no such pair in  $R$ .
- ▶ The algorithm  $A$  runs in poly-time

Clearly  $P_{\text{search}} \subset P$ .

- ▶ In some vague undefined sense we will not worry about

# Search Problems

Just as NP denotes the class of efficiently checkable decision problems

We can define  $\text{NP}_{\text{search}}$  as the class of efficiently checkable search problems.

The class  $\text{NP}_{\text{search}}$  is the set of relations  $R$  such that

- ▶ There is an algorithm  $A$  (depending on  $R$ )
- ▶ We have  $(x, y) \in R$  iff  $A(x, y)$  outputs 1
- ▶ The algorithm  $A$  runs in poly-time

## $P_{search} \subset NP_{search}?$

Surprisingly no...

Let  $R = \{(0, 0), (0, b) | b = 1 \text{ if God exists, } b = 0 \text{ otherwise}\}$ .

Then  $R$  is a polynomially-bounded relation.

$R$  is in  $P_{search}$  since

- ▶ On input 0, output 0
- ▶ On input  $x \neq 0$ , output fail

is a polynomial-time search algorithm for  $R$ .

But  $R$  is not in  $NP_{search}$ , assuming I cannot (dis)prove the existence of God by computer, since any algorithm that could verify whether or not  $(0, 1) \in R$  would be such a proof.

# Search To Decision Reductions

We can relate some search problems to **other** decision problems.

## Example:

Given a group  $G$  of prime order  $q$  and two elements  $g$  and  $h = g^x$  for some (unknown)  $x$

Write down an algorithm to find  $x$  (This is called the Discrete Log Problem)

- ▶ Suppose you are given an algorithm which on input of  $(g', h') \in G$  finds the least significant bit of the discrete log  $y'$  such that  $h' = g^{y'}$ ,
  - ▶ i.e. it **decides** if  $y$  is odd or even.



# Search To Decision Reductions

Some search problems are believed to be **much** harder than their associated decision ones.

- ▶ Bit esoteric, but in Crypto one has instances for which search Diffie–Hellman is hard but decision Diffie–Hellman is easy.

Given a group  $G$  of prime order  $g$

## Computational/Search Diffie–Hellman:

Given  $g$ ,  $g^x$  and  $g^y$  (but not  $x$  or  $y$ ) determine  $g^{xy}$ .

## Decision Diffie–Hellman:

Given  $g$ ,  $g^x$  and  $g^y$  and either ( $Z = g^z$  or  $Z = g^{xy}$  with 50 percent probability) (but not  $x$ ,  $y$  or  $z$ ) determine whether  $Z = g^z$  or  $Z = g^{xy}$ .

In general it is believed that for **most** groups if the DLP is hard, then these two problems are equivalent.

- ▶ But there are some groups for which this does not hold.

# Search To Decision Reductions

We can relate some search problems to **the associated** decision problems.

**Example:** Consider the two problems

## SEARCH-KNAPSACK:

Given  $x_1, \dots, x_n$  and an  $S$  find  $b_i \in \{0, 1\}$  such that

$$\sum b_i \cdot x_i = S.$$

## DECISION-KNAPSACK:

Given  $x_1, \dots, x_n$  and an  $S$  find **if there exists**  $b_i \in \{0, 1\}$  such that

$$\sum b_i \cdot x_i = S.$$

Show that the two problems are poly-time equivalent.

# Average and Worst Case

Another problem with the traditional view of complexity theory is that it only deals with worst case problems.

Consider the problem

FACTORING

$= \{ \langle x, y \rangle \mid x \text{ is an integer with a prime factor lower than } y \}$ .

We know  $\text{FACTORING} \in \text{NP} \cap \text{co-NP}$ .

But we do not know whether  $\text{FACTORING} \in \text{P}$ .

The reason FACTORING is hard is because some numbers are hard to factor!

- ▶ The worst case is hard

# Average and Worst Case

But **on average** FACTORING is easy.

Fifty percent of the time we read in  $x$  and  $y$ , and as long as  $y > 2$  we output 1.

- ▶ Since fifty percent of all random numbers  $x$  are even!

So FACTORING is easy **on average**.

Indeed most advanced factoring algorithms use a trick of factoring the hard number by finding lots of factors of easy numbers.

# Average and Worst Case

Even NP-complete problems are not hard on average

In industry most SAT problems one encounters can be easily solved.

Graph 3-colourability is NP-complete

- ▶ In worst case colouring the nodes such that no edge connects two nodes with the same colour is very hard.
- ▶ On average (with a specific definition of what is a random graph) can solve this in **constant time**.
- ▶ Most graphs are not 3-colourable, and we can quickly determine they are not.

# Random Self-Reductions

Are there some problems which are as bad in the worst case as they are on average?

Consider the *DDH* problem considered earlier

Suppose DDH is easy on average, i.e. given a random DDH instance  $(A, B, C, D) = (g, g^x, g^y, Z)$  there is an algorithm  $A$  which will solve the DDH problem.

Now suppose we are given a hard instance  $(A', B', C', D')$ , we can turn it into a random instance by setting

$$A = A'^{r_1}, \quad B = B'^{r_1 \cdot r_2}, \quad C = C'^{r_1 \cdot r_3}, \quad D = D'^{r_1 \cdot r_2 \cdot r_3}.$$

such  $(A', B', C', D')$  is a DDH tuple iff  $(A, B, C, D)$  is a DDH tuple.

- Pick  $r_1, r_2, r_3 \in \{0, \dots, |G| - 1\}$ .

Thus the worst case problem can be reduced to the average case