# Introduction to Computational Complexity A 10-lectures Graduate Course

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#### Introduction

### Administrative Meta-Information

- 5-Day course, Monday (13.7.) to Friday (17.7.)
- Schedule:

```
Mon (13.7.): 10:00 - 12:00, 16:30 - 18:30
Tue (14.7.): 10:00 - 12:00, 14:00 - 16:00
Wed (15.7.): 10:00 - 12:00, 14:00 - 16:00
Thu (16.7.): 10:00 - 12:00, 14:00 - 16:00
Fri (17.7.): 10:00 - 12:00, 16:30 - 18:30
```

• Lecture notes avaiable at:

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http://www.it.uu.se/katalog/marst984/cc-st09
```

- Some small assignments at end of day
- Course credits: ??
- Course is interactive, so:

Any questions so far?

# What is Computational Complexity?

- Studies intrinsic complexity of computational tasks
- Absolute Questions:
  - How much time is needed to perform the task?
  - How much resources will be needed?
- Relative Questions:
  - More difficult than other tasks?
  - Are there "most difficult" tasks?
- (Surprisingly: Many relative answers, only few absolute ones..)
- Rigorous treatment:
  - Mathematical formalisms, minimize "hand-waving"
  - Precise definitions
  - Theorems have to be proved

After all: Complexity Theory

# What is Computational Complexity? (Cont.)

### • Basis: Computability Theory

- Provides models of computation
- Explores their strength (expressiveness)
- Question: "What can be computed (at all)?"
- Then: Complexity Theory
  - Tries to find meaningful complexity measures
  - Tries to classify and relate problems
  - Tries to find upper and lower complexity bounds
  - Question: "What can efficiently be computed?"
- One core concern:
  - What does "efficiently" actually mean?
  - Proving vs. Verifying ( $P \stackrel{?}{=} NP$  problem)

# Course Outline

- Introduction
  - Basic Computability Theory
    - Formal Languages
    - Model of Computation: Turing Machines
    - Decidability, Undecidability, Semi-Decidability

### 2 Complexity Classes

- Landau Symbols: The  $\mathcal{O}(\cdot)$  Notation
- Time and Space Complexity
- Relations between Complexity Classes
- Feasible Computations: P vs. NP
  - Proving vs. Verifying
  - Reductions, Hardness, Completeness
  - Natural NP-complete problems
  - Advanced Complexity Concepts
    - Non-uniform Complexity
    - Probabilistic Complexity Classes
    - Interactive Proof Systems

This is a theoretical course

– expect a lot of "math"!

### Problems as Formal Languages

- Start with very high-level model of computation
- Assume a machine with input and output
- Formal notation for format:
  - $\Sigma = \{\sigma_1, \dots, \sigma_k\}$  is a finite set of *symbols*
  - $w = (w_1, \ldots, w_l)$  is a word over  $\Sigma: \forall i : w_i \in \Sigma$

★ Write also just w₁w₂...w₁

- I is the *length* of w, also denoted |w|
- $\varepsilon$  is the *empty word*, i.e.,  $|\varepsilon| = 0$
- $\Sigma^k$  is the set of words of length k
- $\Sigma^* = \bigcup_{k \ge 0} \Sigma^k$  are all words over  $\Sigma$
- A language is a set  $L \subseteq \Sigma^*$
- Let  $L_1, L_2 \subseteq \Sigma^*$ , language operations:
  - ★  $L_1 \cup L_2$  (union),  $L_1 \cap L_2$  (intersection),  $L_1 L_2$  (difference)
  - ★  $\overline{L} := \Sigma^* L$  (complement)
  - ★  $L_1L_2 := \{w \mid \exists w_1 \in L_1, w_2 \in L_2 : w = w_1w_2\}$  (concatenation)

# Problems as Formal Languages (Example)

#### Example: NAT

- Let  $\Sigma:=\{0,1,\ldots,9\}$
- $\Sigma^{\ast}$  is all strings with digits
- Let  $[n]_{10}$  denote decimal representation of  $n \in \mathbb{N}$
- NAT :=  $\{[n]_{10} \mid n \in \mathbb{N}\} \subsetneq \Sigma^*$  all representations of naturals •  $010 \in \Sigma^* - NAT$
- Machine for calculating square:

Input:  $w = [n]_{10} \in \Sigma^*$ Output:  $v \in \Sigma^*$  with  $v = [n^2]_{10}$ (Plus error-handling for  $w \notin NAT$ )

### Problems as Formal Languages (2nd Example)

### Example: PRIMES

- $\bullet$  Let  $\Sigma:=\{0,1,\ldots,9\}$  and NAT as before
- PRIMES := { $[p]_{10} | p$  is a prime number}
- Clearly:  $\mathsf{PRIMES} \subsetneq \mathsf{NAT}$
- Machine *M* for checking primality:

Input:  $w = [n]_{10} \in \Sigma^*$ Output: 1 if *n* is prime, 0 otherwise

- This is a *decision problem*:
  - PRIMES are the positive instances,
  - >  $\Sigma^*$  PRIMES (everything else) the *negative instances*
  - M has to distinguish both (it decides PRIMES)

Important concept, will come back to that later!

### Model of Computation

- Input/Output format defined. What else?
- Model of Computation should:
  - Define what we mean with "computation", "actions", ...
  - Be simple and easy to use
  - ... but yet powerful
- Models: Recursive Functions, Rewriting Systems, Turing Machines, ...
- All equally powerful:

#### Church's Thesis

All "solvable" problems can be solved by any of the above formalisms.

#### • We will focus on the *Turing Machine*.

### **Turing Machine**

• Turing Machines are like simplified computers containing:

- A tape to read/write on
  - ★ Contains squares with one symbol each
  - ★ Is used for input, output and temporary storage
  - ★ Unbounded
- A read/write head
  - \* Can change the symbol on the tape at current position
  - ★ Moves step by step in either direction
- A finite state machine
  - ★ Including an initial state and final states
- Looks simple, but is very powerful
- Standard model for the rest of the course

# Turing Machine: Definition

#### Definition (Turing Machine)

A Turing machine M is a five-tuple  $M = (Q, \Gamma, \delta, q_0, F)$  where

- Q is a finite set of *states*;
- $\Gamma$  is the *tape alphabet* including the blank:  $\Box \in \Gamma$ ;
- $q_0$  is the *initial state*,  $q_0 \in Q$ ;
- *F* is the set of final states,  $F \subseteq Q$ ;
- $\delta$  is the transition function,  $\delta : (Q F) \times \Gamma \rightarrow Q \times \Gamma \times \{R, N, L\}.$

Operation:

- Start in state  $q_0$ , input w is on tape, head over its first symbol
- Each step:
  - Read current state q and symbol a at current position
  - Lookup  $\delta(q, a) = (p, b, D)$
  - Change to state p, write b, move according to D
- Stop as soon as  $q \in F$ . Left on tape: Output

### Turing Machine: Configuration

- Configuration (w, q, v) denotes status after each step:
  - ► Tape contains wv (with infinitely many □ around)
  - Head is over first symbol of v
  - Machine is in state q
- Start configuration:  $(\varepsilon, q_0, w)$  if input is w
- End configuration: (v, q, z) for a  $q \in F$ 
  - Output is z, denoted by M(w)
  - In case machine doesn't halt (!): M(w) = ↗

### Turing Machine: Step Relation

### • Step relation: Formalizes semantics of Turing machine

### Definition (Step Relation)

Let  $M = (Q, \Gamma, \delta, q_0, F)$ , define  $\vdash$  for all  $w, v \in \Gamma^*, a, b \in \Gamma$  and  $q \in Q$  as:

$$(wa, q, bv) \vdash \begin{cases} (wac, p, v) & \text{if } \delta(q, b) = (p, c, R), \\ (wa, p, cv) & \text{if } \delta(q, b) = (p, c, N), \\ (w, p, acv) & \text{if } \delta(q, b) = (p, c, L). \end{cases}$$

- $\alpha$  reaches  $\beta$  in 1 step:  $\alpha \vdash \beta$
- $\alpha$  reaches  $\beta$  in k steps:  $\alpha \vdash^k \beta$
- $\alpha$  reaches  $\beta$  in any number of steps:  $\alpha \vdash^* \beta$

### Turing Machine: The Universal Machine

- Turing machine model is quite simple
- Can be easily simulated by a human
  - Provided enough pencils, tape space and patience
- Important result: Machines can simulate machines
  - Turing machines are *finite* objects!
  - Effective encoding into words over an alphabet
  - Also configurations are *finite*! Encode them also
- Simulator machine U only needs to
  - Receive an encoded M as input
  - Input of *M* is *w*, give that also to *U*
  - U maintains encoded configurations of M and applies steps

# Turing Machine: The Universal Machine (Cont.)

• Let  $\langle M \rangle$  be encoding of machine M.

Theorem (The Universal Machine)

There exists a universal Turing machine U, such that for all Turing machines M and all words  $w \in \Sigma^*$ :

 $U(\langle M \rangle, w) = M(w)$ 

In particular, U does not halt iff  $^1$  M does not halt.

(Without proof.)

<sup>1</sup> "if and only if"

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### Turing Machine: Transducers and Acceptors

- Definition so far: Receive input, compute output
- We call this a *transducer*:
  - Interpret a TM M as a function  $f: \Sigma^* \to \Sigma^*$
  - All such f are called computable functions
  - Partial functions may be undefined for some inputs w
    - ★ In case *M* does not halt for them  $(M(w) = \nearrow)$
  - Total functions are defined for all inputs
- For decision problems L: Only want a positive or negative answer
- We call this an *acceptor*:
  - Interpret M as halting in
    - ★ Either state  $q_{yes}$  for positive instances  $w \in L$
    - ★ Or in state  $q_{no}$  for negative instances  $w \notin L$
  - Output does not matter, only final state
  - ► *M* accepts the language *L*(*M*):

$$L(M) := \{w \in \Sigma^* \mid \exists y, z \in \Gamma^* : (\varepsilon, q_0, w) \vdash^* (y, q_{yes}, z)\}$$

• Rest of the course: Mostly acceptors

### Turing Machine: Multiple Tapes

- Definition so far: Machine uses one tape
- More convenient to have k tapes (k is a constant)
  - As dedicated input/output tapes
  - To save intermediate results
  - To precisely measure used space (except input/output space)
- Define this as *k*-tape Turing machines
  - Still only one state, but k heads
  - Equivalent to 1-tape TM in terms of expressiveness (Encode a "column" into one square)
  - Could be more efficient, but not much
- Rest of the course: *k*-tape TM with dedicated input/output tapes

### Turing Machine: Non-determinism

- Definition so far: Machine is deterministic
  - Exactly one next step possible
- Extension: Allow different possible steps

 $\delta: (Q - F) \times \Gamma \to \mathcal{P}(Q \times \Gamma \times \{R, N, L\})$ 

- Machine chooses non-deterministically which step to do
  - Useful to model uncertainty in a system
  - Imagine behaviour as a *computation tree*
  - Each path is one possible computation
  - Accepts w iff there is a path to q<sub>yes</sub> (accepting path)
- Not a real machine, rather a theoretical model
- Will see another characterization later
- Expressiveness does not increase in general (see following Theorem)

### Turing Machine: Non-determinism (Cont.)

#### Theorem

Given a non-deterministic TM N, one can construct a deterministic TM M with L(M) = L(N). Further, if N(w) accepts after t(w) steps, then there is c such that M(w)accepts after at most  $c^{t(w)}$  steps.

#### Remark

- Exponential blowup concerning speed
- Ignoring speed, expressiveness is the same
- Note that *N* might not terminate on certain inputs

# Turing Machine: Non-determinism (Cont. 2)

### Proof (Sketch).

- Given a non-deterministic N and an input w
- Search the computation tree of N
- Breadth-first technique: Visit all "early" configurations first
  - Since there may be infinite paths
  - For each  $i \ge 0$ , visit all configurations up to depth i
  - If *N* accepts *w*, we will find accepting configuration at a depth *t* and halt in  $q_{yes}$
  - If *N* rejects *w*, we halt in  $q_{no}$  or don't terminate
- Let d be maximal degree of non-determinism (choices of  $\delta$ )
- Above takes at most  $\sum_{i=0}^{t} d^{i}$  steps
- Can be bounded from above by  $c^t$  with a suitable constant c

# Summary (Turing Machine)

- Simple model of computation, but powerful
- Clearly defined syntax and semantics
- May accept languages or compute functions
- May use *multiple tapes*
- Non-determinism does not increase expressiveness
- A Universal Machine exists, simulating all other machines

### Remark

The machines we use from now on

- are deterministic
- are acceptors,
- with k tapes

(except stated otherwise).

### Deciding a Problem

#### • Recall: Turing Machines running with input w may

- halt in state q<sub>yes</sub>,
- halt in state q<sub>no</sub>, or
- run without halting.

• Given problem L and instance w, want to decide whether  $w \in L$ :

- Using a machine M
- If  $w \in L$ , M should halt in  $q_{yes}$
- If  $w \notin L$ , *M* should halt in  $q_{no}$
- In particular: Always terminate! (Little use otherwise...)

### Decidability and Undecidability

### Definition

*L* is called *decidable*, if there exists a TM *M* with L(M) = L that *halts on all inputs*. REC is the set of all decidable languages.

- We can *decide* the status of w by just running M(w).
- Termination guaranteed, we won't wait infinitely
- "M decides L"
- If  $L \notin \text{REC}$ , then L is undecidable

## Decidability and Undecidability: Example

### Example (PRIMES $\in$ REC)

- Recall PRIMES :=  $\{[p]_{10} | p \text{ is a prime number}\}$
- Can be decided:
  - Given  $w = [n]_{10}$  for some n
  - Check for all  $i \in (1, n)$  whether n is multiple of i
  - If an i found: Halt in q<sub>no</sub>
  - Otherwise, if all i negative: Halt in q<sub>yes</sub>
- Can be implemented with a Turing machine
- Always terminates (only finitely many i)
- Thus:  $\mathsf{PRIMES} \in \mathsf{REC}$

### Semi-Decidability

#### Definition

L is called *semi-decidable*, if there exists a TM M with L(M) = L. RE is the set of all semi-decidable languages.

- Note the missing "halts on all inputs"!
- We can only "half-decide" the status of a given w:
  - Run M, wait for answer
  - If  $w \in L$ , M will halt in  $q_{yes}$
  - If  $w \notin L$ , M may not halt
  - We don't know:  $w \notin L$  or too impatient?
- "M semi-decides L"



### **Class Differences**

- Questions at this point:
  - Are there undecidable problems?
  - 2 Can we at least semi-decide some of them?
  - In the second second
- Formally: REC  $\stackrel{?}{\subsetneq}$  RE  $\stackrel{?}{\subsetneq} \mathcal{P}(\Sigma^*)$
- Subtle difference between REC and RE: Termination guarantee

### Properties of Complementation

#### Theorem

L ∈ REC ⇐⇒ L̄ ∈ REC. ("closed under taking complements")
L ∈ REC ⇐⇒ (L ∈ RE ∧ L̄ ∈ RE).

### Proof (First part).

- Direction " $\Longrightarrow$ ":
  - Assume M decides L and halts always
  - Construct M': Like M, but swap  $q_{yes}$  and  $q_{no}$
  - M' decides L and halts always!
- Direction "←":
  - Exact same thing.

### Properties of Complementation

#### Theorem

L ∈ REC ⇐⇒ L̄ ∈ REC. ("closed under taking complements")
L ∈ REC ⇐⇒ (L ∈ RE ∧ L̄ ∈ RE).

### Proof (Second part).

- Direction " $\Longrightarrow$ ":
  - Follows from  $\mathsf{REC} \subseteq \mathsf{RE}$  and first part
- Direction "←":
  - Let  $M_1, M_2$  with  $L(M_1) = L$  and  $L(M_2) = \overline{L}$
  - Given w, simulate  $M_1(w)$  and  $M_2(w)$  step by step, in turns
  - Eventually one of them will halt in q<sub>yes</sub>
  - If it was M<sub>1</sub>, halt in q<sub>yes</sub>
  - It it was M<sub>2</sub>, halt in q<sub>no</sub>
  - Thus, we always halt (and decide L)!

# The Halting Problem

- Approach our three questions:
  - Are there undecidable problems?
  - ② Can we at least semi-decide some of them?
  - In the second second
- Classical problem: Halting Problem
  - Given a program M (Turing machine!) and an input w
  - ► Will *M*(*w*) terminate?
  - Natural problem of great practical importance
- Formally: Let  $\langle M \rangle$  be an encoding of M

### Definition (Halting Problem)

*H* is the set of all Turing machine encodings  $\langle M \rangle$  and words *w*, such that *M* halts on input *w*:

$$H := \{(\langle M \rangle, w) \mid M(w) \neq \nearrow\}$$

### Undecidability of the Halting Problem

#### Theorem

 $H \in \mathsf{RE} - \mathsf{REC}$ 

### Proof (First part).

We show  $H \in RE$ :

- Need to show: There is a TM M', such that
  - Given *M* and *w*
  - If M(w) halts, M' accepts (halts in  $q_{yes}$ )
  - If M(w) doesn't halt, M' halts in  $q_{no}$  or doesn't halt
- Construct M': Just simulate M(w)
  - If simulation halts, accept (i.e. halt in  $q_{yes}$ )
  - If simulation doesn't halt, we also won't
- Thus: L(M') = H

## Undecidability of the Halting Problem (Cont.)

#### Theorem

 $H \in \mathsf{RE} - \mathsf{REC}$ 

### Proof (Second part).

#### We show $H \notin \text{REC}$ :

- Need to show: There is no TM  $M_H$ , such that
  - Given M and w
  - If M(w) halts,  $M_H$  accepts (halts in  $q_{yes}$ )
  - If M(w) doesn't halt,  $M_H$  rejects (halts in  $q_{no}$ )
  - Note: M<sub>H</sub> always halts!
- We can't use simulation!
  - What if it doesn't halt?
- New approach: Indirect proof
  - Assume there is  $M_H$  with above properties
  - Show a contradiction

# Undecidability of the Halting Problem (Cont. 2)

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 $H \in \mathsf{RE} - \mathsf{REC}$ 

#### Proof (Second part, cont.)

#### We show $H \notin \text{REC}$ : Assume there is $M_H$ that always halts

- Build another machine N:
  - On input w, simulate  $M_H(w, w)$
  - If simulation halts in  $q_{yes}$ , enter infinite loop
  - If simulation halts in q<sub>no</sub>, accept (i.e. halt in q<sub>yes</sub>)
- *N* is Turing machine and  $\langle N \rangle$  its encoding. *Does*  $N(\langle N \rangle)$  *halt?*
- Assume "yes,  $N(\langle N \rangle)$  halts":
  - By construction of N,  $M_H(\langle N \rangle, \langle N \rangle)$  halted in  $q_{no}$
  - Definition of H:  $N(\langle N \rangle)$  does not halt. Contradiction!
- Assume "no,  $N(\langle N \rangle)$  doesn't halt":
  - By construction of N,  $M_H(\langle N \rangle, \langle N \rangle)$  halted in  $q_{yes}$
  - Definition of H:  $N(\langle N \rangle)$  does halt. Contradiction!
- N can not exist!  $\implies M_H$  can not exist.

# Class Differences: Results

- Know now:  $H \in RE REC$ , thus: REC  $\subsetneq RE$
- What about RE and  $\mathcal{P}(\Sigma^*)$ ?
  - Is there an  $L \subseteq \Sigma^*$  that's not even semi-decidable?
- Counting argument:
  - ▶ RE is *countably* infinite: Enumerate all Turing machines
  - P(Σ\*) is uncountably infinite: Σ\* is countably infinite

### Corollary

 $\mathsf{REC} \subsetneq \mathsf{RE} \subsetneq \mathcal{P}(\Sigma^*)$ 

#### Remark

- Actually, we even know one of those languages:  $\overline{H} \notin RE$
- Otherwise, H would be decidable:  $(H \in \mathsf{RE} \land \overline{H} \in \mathsf{RE}) \implies H \in \mathsf{REC}$

### Reductions

- We saw: Some problems are harder than others
- Possible to compare them directly?
- Concept for this: *Reductions* 
  - Given problems A and B
  - Assume we know how to solve A using B
  - Then: Sufficient to find out how to solve B for solving A
  - We reduced A to B
  - Consequence: A is "easier" than B
- Different formal concept established
  - Differ in how B is used when solving A
  - We use Many-one reductions

# Reductions: Definition

### Definition (Many-one Reduction)

 $A \subseteq \Sigma^*$  is many-one reducible to  $B \subseteq \Sigma^*$   $(A \leq_m B)$ , if there is  $f : \Sigma^* \to \Sigma^*$  (computable and total), such that

$$\forall w \in \Sigma^* : w \in A \iff f(w) \in B$$

f the reduction function.

- f maps positive to positive instances, negative to negative
- Impact on decidability:
  - Given problems A and B with  $A \leq_m B$
  - And given  $M_f$  calculating reduction f
  - And given M<sub>B</sub> deciding B
  - Decide A by simulating M<sub>f</sub> and on its output M<sub>B</sub>

### **Reductions:** Properties

#### Lemma

For all A, B and C the following hold: a  $A \leq_m B \land B \in \text{REC} \implies A \in \text{REC}$ a  $A \leq_m B \land B \in \text{RE} \implies A \in \text{REC}$ b  $A \leq_m B \land B \leq_m C \implies A \leq_m C$ c  $A \leq_m B \iff \overline{A} \leq_m \overline{B}$ 

(Closedness of REC under  $\leq_m$ ) (Closedness of RE under  $\leq_m$ ) (Transitivity of  $\leq_m$ )

### Proof.

- First two: We just discussed this
- Second two: Easy exercise
### Reductions: Example

#### Example (The Problems)

- Need to introduce two problems: REACH and REG-EMPTY
- First Problem: The reachability problem:

 $\mathsf{REACH} := \{ (G, u, v) \mid \text{there is a path from } u \text{ to } v \text{ in } G \}$ 

- G is a finite directed graph; u, v are nodes in G
- Question: "Is v reachable from u?"
- $\blacktriangleright$  Easily solvable using standard breath first search: REACH  $\in$  REC
- Second Problem: Emptiness problem for regular languages

 $\mathsf{REG}\mathsf{-}\mathsf{EMPTY} := \{ \langle D \rangle \mid L(D) = \emptyset \}$ 

- D encodes a Deterministic Finite Automaton
- Question: "Is the language D accepts empty?"

### Reductions: Example (Cont.)

#### Example (The Reduction)

- Will reduce REG-EMPTY to REACH
- Idea: Interpret DFA D as a graph
  - Is a final state reachable from initial state?
  - Thus: Start node u is initial state
  - Problem: Want just one target node v, but many final states possible
  - Solution: Additional node v with edges from final states

• Result: 
$$f$$
 with  $\langle D \rangle \mapsto (G, u, v)$ 

- L(D) empty  $\iff u$  can not reach v
- Thus: REG-EMPTY  $\leq_m \overline{\text{REACH}}$
- Remark: Implies REG-EMPTY ∈ REC (Closedness of REC under complement!)

## A Second Example: Halting Problem with empty input

#### Lemma

The Halting Problem with empty input is undecidable, i.e.:

 $H_{\varepsilon} := \{ \langle M \rangle \mid M(\varepsilon) \neq \nearrow \} \notin \mathsf{REC}$ 

#### Proof.

- Already know:  $H \notin \text{REC}$
- Sufficient to find a reduction  $H \leq_m H_{\varepsilon}$  (Closedness!)
- Given is  $(\langle M \rangle, w)$ : A machine M with input w
- Idea: Encode w into the states
- Construct a new machine M':
  - Ignore input and write w on tape (is encoded in states of M')
  - Simulate M
- $f: (\langle M \rangle, w) \mapsto \langle M' \rangle$  is computable: Simple syntactical manipulations!
- Reduction property by construction:
  - If  $(\langle M \rangle, w) \in H$ , then M' terminates with all inputs (also with empty input)
  - ▶ If  $(\langle M \rangle, w) \notin H$ , then M' doesn't ever terminate (also not with empty input)

### Rice's Theorem: Introduction

- We know now: Halting is undecidable for Turing machines
- Even for just empty input!
- Are other properties undecidable?
- (Maybe halting is just a strange property..)
- Will see now: No "non-trivial" behavioural property is decidable!
  - For Turing machines
  - Simpler models behave better (DFA..)
  - Non-trivial: Some Turing machines have it, some don't
- High practical relevance:
  - Either have to restrict model (less expressive)
  - Or only approximate answers (less precise)
- Formally: Rice's Theorem

### Rice's Theorem: Formal formulation

#### Theorem (Rice's Theorem)

Let C be a non-trivial class of semi-decidable languages, i.e.,  $\emptyset \subsetneq C \subsetneq RE$ . Then the following  $L_C$  is undecidable:

 $L_{\mathcal{C}} := \{ \langle M \rangle \mid L(M) \in \mathcal{C} \}$ 

#### Proof (Overview).

- First assume  $\emptyset \notin C$
- Then there must be a non-empty  $A \in C$  (since C is non-empty)
- We will reduce H to  $L_C$
- Idea:
  - We are given M with input w
  - Simulate M(w)
  - If it halts, we will semi-decide A
    - If it doesn't halt, we will semi-decide  $\emptyset$  (never accept)
  - This is the reduction!

### Rice's Theorem: Formal formulation (Cont.)

#### Theorem (Rice's Theorem)

Let C be a non-trivial class of semi-decidable languages, i.e.,  $\emptyset \subsetneq C \subsetneq RE$ . Then the following  $L_C$  is undecidable:

$$L_{\mathcal{C}} := \{ \langle M \rangle \mid L(M) \in \mathcal{C} \}$$

#### Proof (Details).

- Recall:  $\emptyset \notin C$ ,  $A \in C$ , let  $M_A$  be machine for A
- Construct a new machine M':
  - 1 Input y, first simulate M(w) on second tape
  - 2 If M(w) halts, simulate  $M_A(y)$
- Reduction property by construction:
  - If  $(\langle M \rangle, w) \in H$ , then L(M') = A, thus  $\langle M' \rangle \in L_{\mathcal{C}}$
  - If  $(\langle M \rangle, w) \notin H$ , then  $L(M') = \emptyset$ , thus  $\langle M' \rangle \notin L_{\mathcal{C}}$
- What about the case  $\emptyset \in C$ ? Similar construction showing  $H \leq_m \overline{L_C}$

### Rice's Theorem: Examples

#### Example

• The following language is *undecidable*:

 $L := \{ \langle M \rangle \mid L(M) \text{ contains at most 5 words} \}$ 

- $\bullet$  Follows from Rice's Theorem since  $\mathcal{C} \neq \emptyset$  and  $\mathcal{C} \neq \mathsf{RE}$
- Thus: For any k, can't decide if an M only accepts at most k inputs

#### Example

• The following language is *decidable*:

 $L := \{ \langle M \rangle \mid M \text{ contains at most 5 states} \}$ 

- Easy check by looking at encoding of M
- Not a *behavioural* property

### Summary Computability Theory

- Defined a model of computation: Turing machines
- Explored properties:
  - Decidability and Undecidability
  - Semi-Decidability
  - Example: The Halting problem is undecidable
- Reductions as a *relative* concept
- Closedness allows using them for *absolute* results
- Rice's Theorem:

All non-trivial behavioural properties of TM are undecidable.

### Course Outline

- Introduction
- Basic Computability Theory
  - Formal Languages
  - Model of Computation: Turing Machines
  - Decidability, Undecidability, Semi-Decidability

#### Complexity Classes

- Landau Symbols: The  $\mathcal{O}(\cdot)$  Notation
- Time and Space Complexity
- Relations between Complexity Classes
- Feasible Computations: P vs. NP
  - Proving vs. Verifying
  - Reductions, Hardness, Completeness
  - Natural NP-complete problems
- Advanced Complexity Concepts
  - Non-uniform Complexity
  - Probabilistic Complexity Classes
  - Interactive Proof Systems

### **Restricted Resources**

- Previous Chapter: Computability Theory
  - "What can algorithms do?"
- Now: Complexity Theory
  - "What can algorithms do with restricted resources?"
  - Resources: Runtime and memory
- Assume the machines always halt in q<sub>yes</sub> or q<sub>no</sub>
  - But after how many steps?
  - How many tape positions were necessary?

### Landau Symbols

- Resource bounds will depend on input size
- Described by functions  $f : \mathbb{N} \to \mathbb{N}$
- Need ability to express "grows in the order of"
  - Consider  $f_1(n) = n^2$  and  $f_2(n) = 5 \cdot n^2 + 3$
  - Eventually,  $n^2$  dominates for large n
  - Both express "quadratic growth"
  - Want to see all  $c_1 \cdot n^2 + c_2$  equivalent
  - Asymptotic behaviour
- Formal notation for this:  $\mathcal{O}(n^2)$
- Will provide a kind of upper bound of asymptotic growth

### Landau Symbols: Definition

#### Definition

Let  $g : \mathbb{N} \to \mathbb{N}$ .  $\mathcal{O}(g)$  denotes the set of all functions  $f : \mathbb{N} \to \mathbb{N}$  such that there are  $n_0$  and c with

$$\forall n \geq n_0 : f(n) \leq c \cdot g(n).$$

We also just write f(n) = O(g(n)).

# Lemma (Alternative characterization) For $f, g : \mathbb{N} \to \mathbb{N}_{>0}$ the following holds: $f \in \mathcal{O}(g) \iff \exists c > 0 : \limsup_{n \to \infty} \frac{f(n)}{g(n)} \le c$

(Without proof.)

#### Landau Symbols

### Landau Symbols: Examples

• We have 
$$5 \cdot n^2 + 3 = \mathcal{O}(n^2)$$

- One even writes  $\mathcal{O}(n) = \mathcal{O}(n^2)$  (meaning " $\subseteq$ ")
- Both is abuse of notation! Not symmetric:  $\mathcal{O}(n^2) \neq \mathcal{O}(n)!$

#### Examples

- $n \cdot \log(n) = \mathcal{O}(n^2)$
- $n^c = \mathcal{O}(2^n)$  for all constants c
- O(1) are the bounded functions
  - $n^{\mathcal{O}(1)}$  are the functions bounded by a polynomial
- Other symbols exist for lower bounds  $(\Omega)$ , strict bounds  $(o, \omega)$  and "grows equally" ( $\Theta$ )

### Proper complexity functions

- Landau-Symbols classify functions according to growth
- Which functions to consider for resource bounds?
- Only "proper" ones:

#### Definition

Let  $f : \mathbb{N} \to \mathbb{N}$  be a computable function.

- f is time-constructible if there exists a TM which on input  $1^n$  stops after  $\mathcal{O}(n + f(n))$  steps.
- If is space-constructible if there exists a TM which on input 1<sup>n</sup> outputs 1<sup>f(n)</sup> and does not use more than O(f(n)) space.
  - This allows us to assume "stopwatches"
  - All common "natural" functions have these properties

#### Resource measures

#### Definition

- The runtime time<sub>M</sub>(w) of a TM M with input w is defined as: time<sub>M</sub>(w) := max{t ≥ 0 | ∃y, z ∈ Γ\*, q ∈ F : (w, q<sub>0</sub>, ε) ⊢<sup>t</sup> (y, q, z)}
- ② If, for all inputs w and a  $t : \mathbb{N} \to \mathbb{N}$  it holds that time<sub>M</sub>(w) ≤ t(|w|), then M is t(n)-time-bounded. Further:

 $\mathsf{DTIME}(t(n)) := \{L(M) \mid M \text{ is } t(n)\text{-time-bounded}\}\$ 

- The required space space<sub>M</sub>(w) of a TM M with input w is defined as: space<sub>M</sub>(w) := max{n ≥ 0 | M uses n squares on a working tape}
- ④ If for all inputs *w* and an *s* :  $\mathbb{N} \to \mathbb{N}$  it holds that space<sub>*M*</sub>(*w*) ≤ *s*(|*w*|), then *M* is *s*(*n*)-space-bounded. Further:

 $\mathsf{DSPACE}(s(n)) := \{L(M) \mid M \text{ is } s(n)\text{-space-bounded}\}$ 

### Resource measures (Cont.)

#### Definition

I For functions, we have:

 $\mathsf{FTIME}(t(n)) := \{f \mid \exists M \text{ being } t(n)\text{-time-bounded and computing } f\}$ 

For non-deterministic M, time and space are as above, and we have: NTIME(t(n)) := {L(M) | M is non-det. and t(n)-time-bounded} NSPACE(s(n)) := {L(M) | M is non-det. and s(n)-space-bounded}

- Recall: Non-deterministic machines can choose different next steps
  - Can be imagined as a *computation tree*
  - Time and space bounds for all paths in the tree
- Note: space M(w) is for the working tapes
  - Only they "consume memory" during the computation
  - Input (read-only) and output (write-only) should not count
  - Allows notion of *sub-linear space*, e.g., log(|w|)

### Common Complexity Classes

- Deterministic *time complexity* classes:
  - Linear time:

$$\mathsf{LINTIME} := \bigcup_{c \ge 1} \mathsf{DTIME}(cn + c) = \mathsf{DTIME}(\mathcal{O}(n))$$

Polynomial time:

$$\mathsf{P} := \bigcup_{c \ge 1} \mathsf{DTIME}(n^c + c) = \mathsf{DTIME}(n^{\mathcal{O}(1)})$$

Polynomial time functions:

$$\mathsf{FP} := \bigcup_{c \ge 1} \mathsf{FTIME}(n^c + c) = \mathsf{FTIME}(n^{\mathcal{O}(1)})$$

Exponential time

$$\mathsf{EXP} := \bigcup_{c \ge 1} \mathsf{DTIME}(2^{n^c + c}) = \mathsf{DTIME}\left(2^{n^{\mathcal{O}(1)}}\right)$$

### Common Complexity Classes (Cont.)

- Deterministic space complexity classes:
  - Logarithmic space:

$$L := \mathsf{DSPACE}(\mathcal{O}(\mathsf{log}(n)))$$

Polynomial space:

$$\mathsf{PSPACE} := \mathsf{DSPACE}(n^{\mathcal{O}(1)})$$

Exponential space:

$$\mathsf{EXPSPACE} := \mathsf{DSPACE}\left(2^{n^{\mathcal{O}(1)}}\right)$$

 Non-deterministic classes defined similarly: NLINTIME, NP, NEXP, NL, NPSPACE and NEXPSPACE

### Common Complexity Classes: Example

#### Example (REACH)

• Consider again the *reachability problem*:

 $\mathsf{REACH} := \{(G, u, v) \mid \text{there is a path from } u \text{ to } v \text{ in } G\}$ 

- Decidable but how much space is needed?
- Non-deterministically:  $REACH \in NL$ 
  - Explore graph beginning with u
  - Choose next node non-deterministically, for at most n steps
  - If there is a path to v, it can be found that way
  - Space: For step counter and number of current node:  $O(\log(n))$
- Deterministically:  $\mathsf{REACH} \in \mathsf{DSPACE}(\mathcal{O}(\log(n)^2))$ 
  - Sophisticated recursive algorithm
  - Split path p of length  $\leq n$ :
    - $p = p_1 p_2$  with  $p_1, p_2$  of length  $\leq n/2$
    - Iterate over all intermediate nodes
    - Space: Recursion stack depth log(n) and elements log(n):  $O(log(n)^2)$

### Complexity Class Relations

• Clear from definitions:

#### $\mathsf{LINTIME} \subseteq \mathsf{P} \subseteq \mathsf{EXP}$

• Same relation for non-deterministic classes:

 $\mathsf{NLINTIME} \subseteq \mathsf{NP} \subseteq \mathsf{NEXP}$ 

- Only inclusion, no separation yet:
  - Know that LINTIME  $\subseteq$  P
  - But is there  $L \in P LINTIME$ ?
  - Such an L would separate LINTIME and P
- Will now see a very "fine-grained" separation result

### Hierarchy Theorem

#### Theorem (Hierarchy Theorem)

• Let  $f:\mathbb{N}\to\mathbb{N}$  be time-constructible and  $g:\mathbb{N}\to\mathbb{N}$  with

$$\liminf_{n\to\infty}\frac{g(n)\cdot\log(g(n))}{f(n)}=0.$$

Then there exists  $L \in \mathsf{DTIME}(f(n)) - \mathsf{DTIME}(g(n))$ .

• Let  $f:\mathbb{N}\to\mathbb{N}$  be space-constructible and  $g:\mathbb{N}\to\mathbb{N}$  with

$$\liminf_{n\to\infty}\frac{g(n)}{f(n)}=0$$

Then there exists  $L \in DSPACE(f(n)) - DSPACE(g(n))$ .

(Without proof.)

### Hierarchy Theorem: Examples

#### Example

- Let  $C_k := \mathsf{DTIME}(\mathcal{O}(n^k))$
- Using time hierarchy theorem:

 $\mathcal{C}_1 \subsetneq \mathcal{C}_2 \subsetneq \mathcal{C}_3 \subsetneq \dots \qquad (\text{Infinite hierarchy})$ 

- Means: Let p(n) and q(n) be polynomials, deg  $p < \deg q$
- Then there is *L* such that:
  - Decidable in  $\mathcal{O}(q(n))$  time
  - Not decidable in  $\mathcal{O}(p(n))$  time
- Remark: Theorem states "more time means more power"
- Also the case with REC  $\subsetneq$  RE:
  - REC: Time bounded: Always halt
  - RE: May not halt, "infinite time"

### Determinism vs. Non-determinism

#### Theorem

For each space-constructible function  $f : \mathbb{N} \to \mathbb{N}$ , the following holds:

```
\mathsf{DTIME}(f) \subseteq \mathsf{NTIME}(f) \subseteq \mathsf{DSPACE}(f) \subseteq \mathsf{NSPACE}(f)
```

#### Proof (Overview).

- First and third clear: Determinism is special case
- Now show  $NTIME(f) \subseteq DSPACE(f)$
- Time bounded by f(n) implies space bounded by f(n)
- Still need to remove non-determinism
- Key idea:
  - Time bound f(n): At most f(n) non-deterministic choices
  - Computation tree at most f(n) deep
  - Represent paths by strings of size f(n)
  - Simulate all paths by enumerating the strings

### Determinism vs. Non-determinism (Cont.)

#### Theorem

For each space-constructible function  $f : \mathbb{N} \to \mathbb{N}$ , the following holds:

```
\mathsf{DTIME}(f) \subseteq \mathsf{NTIME}(f) \subseteq \mathsf{DSPACE}(f) \subseteq \mathsf{NSPACE}(f)
```

#### Proof (Details).

- Want to show  $NTIME(f) \subseteq DSPACE(f)$
- Let L ∈ NTIME(f) and N corresponding machine
- Let d be maximal degree of non-determinism
- Build new machine M:
  - **①** Systematically generate words  $c \in \{1, \ldots, d\}^{f(n)}$
  - 2 Simulate N with non-deterministic choices c
  - Seperate until all words generated (overwrite c each time)
- Simulation is deterministic and needs only  $\mathcal{O}(f(n))$  space
  - (But takes exponentially long!)

### Deterministic vs. Non-deterministic Space

• Theorem implies:

#### $\mathsf{P}\subseteq\mathsf{NP}\subseteq\mathsf{PSPACE}\subseteq\mathsf{NPSPACE}$

- Thus, in context of polynomial bounds:
  - Non-determinism "beats" determinism
  - Space "beats" time
- But are these inclusions strict?
- Will now see: PSPACE = NPSPACE
- Recall:  $\mathsf{REACH} \in \mathsf{DSPACE}(\mathcal{O}(\log(n)^2))$

### Deterministic vs. Non-deterministic Space (Cont.)

Theorem (Savitch)

For each space-constructible function  $f : \mathbb{N} \to \mathbb{N}$ , the following holds:

```
\mathsf{NSPACE}(f) \subseteq \mathsf{DSPACE}(f^2)
```

Proof (Sketch).

- Let  $L \in \mathsf{NSPACE}(f)$  and  $M_L$  corresponding non-deterministic TM
- Consider configuration graph of  $M_L$  for an input w
  - Each node is a configuration
  - Edges are given by step relation  $\vdash$
  - $M_L$  space bounded, thus only  $c^{f(|w|)}$  configurations
- Assume just one final accepting configuration
- Question: "Is there a path from initial to final configuration?"
- Reachability problem!

• Solve it with 
$$\mathcal{O}\left(\log\left(c^{f(n)}
ight)^{2}
ight)=\mathcal{O}(f(n)^{2})$$
 space

### Polynomial Complexity Classes

#### Corollary

#### $\mathsf{P}\subseteq\mathsf{NP}\subseteq\mathsf{PSPACE}=\mathsf{NPSPACE}$

- Previous theorem implies NPSPACE  $\subseteq$  PSPACE
- First two inclusions: Difficult, next chapter!
- Following concept will be of use:

#### Definition

Let  $\mathcal{C} \subseteq \mathcal{P}(\Sigma^*)$  be a class of languages. We define:

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

• For deterministic  $C \subseteq \mathsf{REC}$ :  $C = \mathsf{co-}C$ 

#### Complementary Classes: Asymmetries

- Consider RE and co- RE:
  - For RE the TM always halts on the positive inputs
    - ★ "For  $x \in L$  there is a finite path to  $q_{yes}$ "
  - For co- RE it always halts on the negative inputs
    - ★ "For  $x \notin L$  there is a finite path to  $q_{no}$ "
  - RE  $\neq$  co- RE (Halting Problem, ..)
  - REC = co- REC and REC = RE  $\cap$  co- RE
- Consider NPSPACE and co- NPSPACE:
  - ▶ We know PSPACE = NPSPACE and PSPACE = co- PSPACE
  - Thus NPSPACE = co- NPSPACE
- What about P, NP and co- NP?
  - Looks like RE situation:
    - ★ NP: "For  $x \in L$  there is a bounded path to  $q_{yes}$ "
    - ★ co-NP: "For  $x \notin L$  there is a *bounded* path to  $q_{no}$ "
  - Surprisingly: Relationship not known!

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- Introduction
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#### 2 Complexity Classes

- Landau Symbols: The  $\mathcal{O}(\cdot)$  Notation
- Time and Space Complexity
- Relations between Complexity Classes
- Feasible Computations: P vs. NP
  - Proving vs. Verifying
  - Reductions, Hardness, Completeness
  - Natural NP-complete problems
  - Advanced Complexity Concepts
    - Non-uniform Complexity
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### **Feasible Computations**

- Will now focus on classes P and NP
- Polynomial time bounds as "feasible", "tractable", "efficient"
  - Polynomials grow only "moderately"
  - Many practical problems polynomial
  - Often with small degrees  $(n^2 \text{ or } n^3)$

### Recall P and NP

- Introduced P and NP via Turing machines:
  - Polynomial time bounds
  - Deterministic vs. non-deterministic operation
- Recall P: For  $L_1 \in P$ 
  - Existence of a deterministic TM M
  - Existence of a polynomial  $p_M(n)$
  - For each input  $x \in \Sigma^*$  runtime  $\leq p_M(|x|)$
- Recall NP: For  $L_2 \in NP$ 
  - Existence of a non-deterministic TM N
  - Existence of a polynomial  $p_N(n)$
  - For each input  $x \in \Sigma^*$  runtime  $\leq p_N(|x|)$
  - For all computation paths
- Theoretical model practical significance?
- Introduce now a new characterization of NP

### A new NP characterization

#### Definition

Let  $R \in \Sigma^* \times \Sigma^*$  (binary relation). *R* is *polynomially bounded*, if there exists a polynomial p(n), such that:

$$\forall (x,y) \in R : |y| \le p(|x|)$$

#### Lemma

NP is the class of all L such that there exists a polynomially bounded  $R_L \in \Sigma^* \times \Sigma^*$  satisfying:

•  $R_L \in P$ , and

• 
$$x \in L \iff \exists w : (x, w) \in R_L.$$

We call w a witness (or proof) for  $x \in L$  and  $R_L$  the witness relation.

### Proving vs. Verifying

- For  $L \in P$ :
  - Machine must decide membership of x in polynomial time
  - Interpret as "finding a proof" for  $x \in L$
- For  $L \in NP$ : (new characterization)
  - Machine is provided a witness w
  - Interpret as "verifying the proof" for  $x \in L$
- *Efficient* proving and verifying procedures:
  - ► For P, runtime is bounded
  - For NP, also witness size is bounded
- Write  $L \in NP$  as:

$$L = \{x \in \Sigma^* \mid \exists w \in \Sigma^* : (x, w) \in R_L\}$$

### Proving vs. Verifying (Cont.)

- P-problems: solutions can be efficiently found
- NP-problems: solutions can be efficiently checked
- Checking certainly a prerequisite for finding (thus  $P \subseteq NP$ )
- But is finding *more* difficult?
  - Intuition says: "Yes!"
  - Theory says: "We don't know." (yet?)
- Formal formulation:

$$P \stackrel{?}{=} NP$$

- One of the most important questions of computer science!
  - ► Many proofs for either "=" or "≠"
  - None correct so far
  - Clay Mathematics Institute offers \$1.000.000 prize

## A new NP characterization (Cont.)

#### Lemma (Revisited)

NP is the class of all L such that there exists a polynomially bounded  $R_L \in \Sigma^* \times \Sigma^*$  satisfying:

- $R_L \in P$ , and
- $x \in L \iff \exists w : (x, w) \in R_L$ . (w is a witness)

#### Proof (First part).

- First, let  $L \in NP$ . Let N be the machine with  $p_N(n)$  time bound.
- Want to show: R<sub>L</sub> as above exists
- Idea:
  - On input x, all computations do  $\leq p_N(|x|)$  steps
    - $x \in L$  iff an *accepting* computation exists
  - $\blacktriangleright$  Encode computation (non-deterministic choices) into w
  - All such pairs (x, w) define  $R_L$
- R<sub>L</sub> has all above properties
# A new NP characterization (Cont. 2)

#### Lemma (Revisited)

NP is the class of all L such that there exists a polynomially bounded  $R_L \in \Sigma^* \times \Sigma^*$  satisfying:

- $R_L \in P$ , and
- $x \in L \iff \exists w : (x, w) \in R_L$ . (w is a witness)

#### Proof (Second part).

- Now, let L as above, using  $R_L$  bounded by p(n)
- Want to show: Non-deterministic N exists, polynomially bounded
- Idea to construct N:
  - R<sub>L</sub> bounds length of w by p(|x|)
  - $R_L \in P$ : There is a *M* for checking  $R_L$
  - N can "guess" w first
  - Then simulate M for checking  $(x, w) \in R_L$
  - ▶ Accepting path exists iff  $\exists w : (x, w) \in R_L$
- N is polynomially time bounded

## A new co- NP characterization

#### Remark

• Recall: All  $L \in NP$  can now be written as:

$$L = \{x \in \Sigma^* \mid \exists w \in \Sigma^* : (x, w) \in R_L\}$$

- Read this as:
  - ▶ Witness relation R<sub>L</sub>
  - For each positive instance, there is a proof w
  - For no negative instance, there is a proof w
  - The proof is efficiently checkable
- Similar characterization for all  $L' \in \text{co-NP}$ :

$$L' = \{x \in \Sigma^* \mid \forall w \in \Sigma^* : (x, w) \notin R_{L'}\}$$

- Read this as:
  - Disproof relation R<sub>L'</sub>
  - For each negative instance, there is a disproof w
  - For no positive instance, there is a disproof w
  - The disproof is efficiently checkable

### **Boolean Formulas**

#### Definition

Let  $X = \{x_1, \ldots, x_N\}$  be a set of variable names.

- Define boolean formulas BOOL inductively:
  - ►  $\forall i : x_i \in \mathsf{BOOL}.$
  - ▶  $\varphi_1, \varphi_2 \in \mathsf{BOOL} \implies (\varphi_1 \land \varphi_2), (\neg \varphi_1) \in \mathsf{BOOL} \ (\textit{conjunction and negation})$

• A truth assignment for the variables in X is a word  $\alpha_1 \dots \alpha_N \in \{0,1\}^N$ .

• The value  $\varphi(\alpha)$  of  $\varphi$  under  $\alpha$  is defined inductively:

$$\frac{\varphi: x_i \quad \neg \psi \quad \psi_1 \wedge \psi_2}{\varphi(\alpha): \alpha_i \quad 1 - \psi(\alpha) \quad \psi_1(\alpha) \cdot \psi_2(\alpha)}$$

#### Shorthand notations:

$$\begin{array}{l} \varphi_1 \lor \varphi_2 \text{ (disjunction) for } \neg(\neg \varphi_1 \land \neg \varphi_2), \\ \varphi_1 \to \varphi_2 \text{ (implication) for } \neg \varphi_1 \lor \varphi_2 \\ \varphi_1 \leftrightarrow \varphi_2 \text{ (equivalence) for } (\varphi_1 \to \varphi_2) \land (\varphi_2 \to \varphi_1) \end{array}$$

 $\alpha$ 

# Example: XOR Function

#### Example

• Consider the *exclusive* or XOR with *m* arguments:

$$\mathsf{XOR}(z_1,\ldots,z_m) := \bigvee_{i=1}^m z_i \wedge \bigwedge_{1 \leq i < j \leq m} \neg (z_i \wedge z_j)$$

• XOR $(z_1, \ldots, z_m) = 1 \iff z_j = 1$  for exactly one j

• Can also be used as shorthand notation.

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# Example for NP: The Satisfiability Problem

#### Example

• Consider 
$$\psi_1 = (x_1 \vee \neg x_2) \wedge x_3$$
 and  $\psi_2 = (x_1 \wedge \neg x_1)$ :

$$\psi_1(lpha)=1$$
 for  $lpha=$  011

- $\psi_2(lpha) = 0$  for all lpha
- $\varphi \in \mathsf{BOOL}$  is called *satisfiable*, if  $\exists \alpha : \varphi(\alpha) = 1$
- $\bullet\,$  Can encode boolean formula into words over fixed alphabet  $\Sigma$
- Language of all satisfiable formulas, the *satisfiability problem*:

 $\mathsf{SAT} := \{ \langle \varphi \rangle \mid \varphi \in \mathsf{BOOL} \text{ is satisfiable} \}$ 

- Obviously, SAT  $\in$  NP:
  - Witness for positive instance  $\langle \varphi \rangle$  is  $\alpha$  with  $\varphi(\alpha) = 1$
  - Size of witness: linearly bounded in  $|\langle \varphi \rangle|$
  - Validity check efficient
- Unknown, whether  $SAT \in P!$

## **Bounded Reductions**

- Comparing P and NP by directly comparing problems
- Assume  $A, B \in \mathsf{NP}$  and  $C \in \mathsf{P}$ 
  - How do A and B relate?
  - ▶ Is C "easier" than A and B?
  - Maybe we just didn't find good algorithms for A or B?
- Recall: *Reductions* 
  - Given problems A and B
  - Solve A by reducing it to B and solving B
  - Tool for that: Reduction function f
  - Consequence: A is "easier" than B
- Used many-one reductions in unbounded setting
- Now: *Bounded setting*, so *f* should be also bounded!
  - Introduce "Cook reductions"

## Polynomial Reduction (Cook Reduction)

#### Definition

 $A \subseteq \Sigma^*$  is polynomially reducible to  $B \subseteq \Sigma^*$  (written  $A \leq_m^p B$ ), if there  $f \in FP$ , such that

$$\forall w \in \Sigma^* : w \in A \iff f(w) \in B$$

#### Lemma

For all A, B and C the following hold:(Closedness of P under  $\leq_m^p$ )A  $\leq_m^p B \land B \in P \Longrightarrow A \in P$ (Closedness of P under  $\leq_m^p$ )A  $\leq_m^p B \land B \in NP \Longrightarrow A \in NP$ (Closedness of NP under  $\leq_m^p$ )A  $\leq_m^p B \land B \leq_m^p C \Longrightarrow A \leq_m^p C$ (Transitivity of  $\leq_m^p$ )A  $\leq_m^p B \iff \overline{A} \leq_m^p \overline{B}$ 

## Hardness, Completeness

- Can *compare* problems now
- Introduce now "hard" problems for a class  $\mathcal{C}$ :
  - Can solve whole  $\mathcal{C}$  if just one of them
  - $\blacktriangleright$  Are more difficult then everything in  ${\cal C}$

### Definition

- A is called *C*-hard, if:  $\forall L \in C : L \leq_m^p A$
- If A is C-hard and  $A \in C$ , then A is called C-complete
- NPC is the class of all NP-complete languages
- NPC: "Most difficult" problems in NP
- Solve one of them, solve whole NP
- Solve one of them efficiently, solve whole NP efficiently

### Hardness, Completeness: Properties

#### Lemma

- A is C-complete if and only if  $\overline{A}$  is co-C-complete.

#### Proof (First part).

- Let A be  $\mathcal{C}$ -complete, and  $L \in \operatorname{co-} \mathcal{C}$
- Want to show:  $L \leq_m^p \overline{A}$
- Indeed:  $L \in \text{co-}\mathcal{C} \iff \overline{L} \in \mathcal{C} \implies \overline{L} \leq_m^p A \iff L \leq_m^p \overline{A}$
- Other direction similar (symmetry)

### Hardness, Completeness: Properties (Cont.)

#### Lemma

- A is C-complete if and only if  $\overline{A}$  is co-C-complete.

#### Proof (Second part).

- Assume  $A \in P \cap NPC$  and let  $L \in NP$
- Want to show:  $L \in P$  (since then P = NP)
- $L \in \mathsf{NP} \implies L \leq_m^p A \text{ since } A \in \mathsf{NPC}$
- $L \leq_m^p A \implies L \in \mathsf{P}$  since  $A \in \mathsf{P}$

### Hardness, Completeness: Properties (Cont. 2)

#### Lemma

- **1** A is C-complete if and only if  $\overline{A}$  is co-C-complete.

#### Proof (Third part).

- Assume  $A \in NPC$ ,  $B \in NP$ ,  $A \leq_m^p B$  and  $L \in NP$
- Want to show:  $L \leq_m^p B$  (since then, B is NP-complete)
- $L \in \mathsf{NP} \implies L \leq_m^p A$  since  $A \in \mathsf{NPC}$
- $L \leq_m^p A \implies L \leq_m^p B$  since  $A \leq_m^p B$  (transitivity!)

# A first NP-complete Problem

• Do NP-complete problems actually exist? Indeed:

Lemma

The following language is NP-complete:

NPCOMP := {( $\langle M \rangle, x, 1^n$ ) | *M* is NTM and accepts x after  $\leq n$  steps}

("NTM" means "non-deterministic Turing machine".)

• How to prove a problem A is NP-complete? 2 parts:

1. Membership: Show  $A \in NP$ 

(Directly or via  $A \leq_m^p B$  for a  $B \in NP$ )

2. Hardness: Show  $L \leq_m^p A$  for all  $L \in NP$ (Directly or via  $C \leq_m^p A$  for a C which is NP-hard)

# A first NP-complete Problem (Cont.)

#### Lemma

The following language is NP-complete:

NPCOMP := {( $\langle M \rangle, x, 1^n$ ) | M is NTM and accepts x after  $\leq n$  steps}

#### Proof (First part).

- Want to show:  $\mathsf{NPCOMP} \in \mathsf{NP}$
- Given  $(\langle M \rangle, x, 1^n)$
- If M accepts x in  $\leq n$  steps, then at most n non-deterministic choices
- For each x, these choices are witness w!
  - Exactly the positive instances x have one w
  - |w| is bounded by n
  - Efficient check by simulating that path
- All (x, w) are witness relation  $R_L$ , so NPCOMP  $\in$  NP

# A first NP-complete Problem (Cont. 2)

#### Lemma

The following language is NP-complete:

NPCOMP := {( $\langle M \rangle, x, 1^n$ ) | M is NTM and accepts x after  $\leq n$  steps}

#### Proof (Second part).

- Want to show now: NPCOMP is NP-hard
- Let  $L \in NP$ , decided by  $M_L$ , bound p(n)
- Show  $L \leq_m^p$  NPCOMP with reduction function:

$$f: x \mapsto (\langle M_L \rangle, x, 1^{p(|x|)})$$

- ►  $f \in \mathsf{FP}$
- ▶ If  $x \in L$ , then  $M_L$  accepts x within p(|x|) steps
- If  $x \notin L$ , then  $M_L$  never accepts x
- Thus:  $x \in L \iff f(x) \in \mathsf{NPCOMP}$

# NP-completeness of SAT

- Know now: There is an NP-complete set
- Practical relevance?
- Are there "natural" NP-complete problems?
- Recall the satisfiability problem:

 $\mathsf{SAT} := \{ \langle \varphi \rangle \mid \varphi \in \mathsf{BOOL} \text{ is satisfiable} \}$ 

- We saw that  $SAT \in NP$ :
  - A satisfying truth assignment  $\alpha$  is witness
- Even more, it's one of the most difficult NP-problems:

Theorem (Cook, Levin)

SAT is NP-complete.

## NP-completeness of SAT: Proof ideas

- We will show NPCOMP  $\leq_m^p$  SAT
- Need reduction function  $f \in FP$  such that:
  - ▶ Input  $(\langle M \rangle, x, 1^n)$ : Machine *M*, word *x*, runtime bound *n*
  - Output  $\psi$ : Boolean formula such that

 $(\langle M \rangle, x, 1^n) \in \mathsf{NPCOMP} \iff \psi \in \mathsf{SAT}.$ 

- Assume *M* has just one tape
- If M accepts x, then within n steps
- Only 2n + 1 tape positions reached!
- Central idea:
  - Imagine a configuration as a line,  $\mathcal{O}(n)$  symbols
  - Whole computation as a matrix with n lines
  - $\blacktriangleright\,$  Encode matrix into formula  $\psi$
  - $\psi$  satisfiable iff computation reaches  $q_{yes}$
  - Formula size = Matrix size =  $O(n^2)$

# NP-completeness of SAT: Proof ideas (Cont.)

- Note: *M* is *non-deterministic* 
  - Different computations possible for each x
  - Different paths in computation tree
- Matrix represents one path to q<sub>yes</sub>
- If  $x \in L(M)$  then there is at least one path to  $q_{yes}$ 
  - Each path described by one matrix
  - Thus, at least one matrix!
- If  $x \notin L(M)$  then there *no path* to  $q_{yes}$ 
  - Thus, there is no matrix!
- Formula  $\psi$  describes a matrix which
  - Represents a computation path
  - Of length at most n
  - ► To q<sub>yes</sub>

• Thus:  $\psi$  satisfiable iff accepting computation path exists!

### NP-completeness of SAT: Proof details

- Describe now the formula  $\psi$
- Given is M, states  $Q = \{q_0, \ldots, q_k\}$ , tape alphabet  $\Gamma = \{a_1, \ldots, a_l\}$
- Final state  $q_{yes} \in Q$
- Used boolean variables:
  - Q<sub>t,q</sub> for all t ∈ [0, n] and q ∈ Q.

     Interpretation: After step t, the machine is in state q.
  - *H*<sub>t,i</sub> for all t ∈ [0, n] and i ∈ [-n, n].
     Interpretation: After step t, the tape head is at position i.
  - T<sub>t,i,a</sub> for all t ∈ [0, n], i ∈ [-n, n] and a ∈ Γ. Interpretation: After step t, the tape contains symbol a at position i.
- Number of variables:  $\mathcal{O}(n^2)$
- Structure of  $\psi$ :

$$\psi := \mathit{Conf} \land \mathit{Start} \land \mathit{Step} \land \mathit{End}$$

### $\psi := Conf \land Start \land Step \land End$

- Part *Conf* of  $\psi$ :
  - ► Ensures: Satisfying truth assignments describe valid computations
- Again, 3 parts:

$$Conf := Conf_Q \wedge Conf_H \wedge Conf_T$$

$$Conf_Q := \bigwedge_{t=0}^n XOR(Q_{t,q_0}, \dots, Q_{t,q_k})$$

$$Conf_H := \bigwedge_{t=0}^n XOR(H_{t,-n}, \dots, H_{t,n})$$

$$Conf_T := \bigwedge_{t=0}^n \bigwedge_{i=-n}^n XOR(T_{t,i,a_1}, \dots, T_{t,i,a_l})$$

## $\psi := Conf \land \mathsf{Start} \land \mathsf{Step} \land \mathsf{End}$

- Part *Start* of  $\psi$ :
  - Ensures: At t = 0, machine is in start configuration
- One single formula:

$$Start := Q_{0,q_0} \land H_{0,0} \land \bigwedge_{i=-n}^{-1} T_{0,i,\square} \land \bigwedge_{i=0}^{|x|-1} T_{0,i,x_{i+1}} \land \bigwedge_{i=|x|}^{n} T_{0,i,\square}$$

## $\psi := Conf \wedge Start \wedge Step \wedge End$

- Part Step of  $\psi$ :
  - ▶ Ensures: At each step, machine executes a legal action
  - Only one tape field changed; head moves by one position
  - Consistency with  $\delta$

$$\begin{split} Step &:= Step_1 \land Step_2\\ Step_1 &:= \bigwedge_{t=0}^{n-1} \bigwedge_{i=-n}^n \bigwedge_{a \in \Gamma} \left( \left( \neg H_{t,i} \land T_{t,i,a} \right) \to T_{t+1,i,a} \right) \\ Step_2 &:= \bigwedge_{t=0}^{n-1} \bigwedge_{i=-n}^n \bigwedge_{a \in \Gamma} \bigwedge_{p \in Q} \left( \left( Q_{t,p} \land H_{t,i} \land T_{t,i,a} \right) \\ & \to \bigvee_{(q,b,D) \in \delta(p,a)} \left( Q_{t+1,q} \land H_{t+1,i+D} \land T_{t+1,i,b} \right) \right) \end{split}$$

## $\psi := Conf \wedge Start \wedge Step \wedge End$

- Part *End* of  $\psi$ :
  - ► Ensures: Eventually, machine reaches an accepting configuration
- One single formula:

$$\mathit{End}:=\bigvee_{t=0}^n Q_{t,q_{yes}}$$

- Completes proof:
  - ▶ By construction,  $\psi \in \mathsf{SAT} \iff (\langle M \rangle, x, 1^n) \in \mathsf{NPCOMP}$
  - Construction is efficient

# co-NP-completeness of UNSAT

#### Remark

- SAT is NP-complete
- Consider its complement:

UNSAT := { $\langle \varphi \rangle \mid \varphi \in \text{BOOL} \text{ is } not \text{ satisfiable}$ } =  $\overline{\text{SAT}}$ 

- Clearly, UNSAT  $\in$  co- NP:
  - Disproof for  $\langle \varphi \rangle$  is  $\alpha$  with  $\varphi(\alpha) = 1$
  - Can be checked efficiently, like for SAT
  - Follows from SAT  $\in$  NP anyway
- SAT is NP-complete  $\iff$  UNSAT is co-NP-complete

#### • Will now study some more NP-complete problems!

# CIRSAT: Satisfiability of Boolean Circuits

#### Definition (Boolean Circuit)

Let  $X = \{x_1, \ldots, x_N\}$  be a set of variable names.

• A boolean circuit over X is a sequence  $c = (g_1, \ldots, g_m)$  of gates:

$$g_i \in \{\perp, \top, x_1, \ldots, x_N, (\neg, j), (\wedge, j, k)\}_{1 \leq j, k < i}$$

• Each  $g_i$  represents a boolean function  $f_c^{(i)}$  with N inputs  $\alpha \in \{0, 1\}^N$ :

$g_i(lpha)$ :	$\perp$	Т	xi	$(\neg, j)$	$(\wedge, j, k)$
$f_c^{(i)}(\alpha)$ :	0	1	$\alpha_i$	$1-f_c^{(j)}(\alpha)$	$f_c^{(j)}(\alpha) \cdot f_c^{(k)}(a)$

- Use  $a \lor b$  as shorthand for  $\neg(\neg a \land \neg b)$
- Whole circuit c represents boolean function  $f_c(\alpha) := f_c^{(m)}(\alpha)$ .
- c is satisfiable if  $\exists \alpha \in \{0,1\}^N$  such that  $f_c(\alpha) = 1$ .

# CIRSAT: Satisfiability of Boolean Circuits (Cont.)

- Practical question: "Is circuit ever 1?"
  - Find unused parts of circuits (like dead code)
- Formally:
  - (Assume again some fixed encoding  $\langle c \rangle$  of circuit c)

### Definition

The circuit satisfiability problem is defined as:

 $\mathsf{CIRSAT} := \{ \langle c \rangle \mid c \text{ is a satisfiable circuit} \}$ 

# CIRSAT: Satisfiability of Boolean Circuits (Cont. 2)

#### Lemma

CIRSAT is NP-complete.

#### Proof.

- CIRSAT  $\in$  NP: Satisfying input is witness w
  - Size N for N variables
  - Verifying: Evaluating all gates is efficient
- SAT  $\leq_m^p$  CIRSAT: Transform formula  $\varphi$  to circuit c
- Remark: Transformation circuit to equivalent formula not efficient
  - Circuit can "reuse" intermediate results
  - CIRSAT  $\leq_m^p$  SAT anyway (SAT is NP-complete!)
  - Transformation produces satisfiability equivalent formula

# CNF: Restricted Structure of Boolean Formulas

#### Definition (CNF)

Let  $X = \{x_1, \ldots, x_N\}$  be a set of variable names.

- A literal I is either  $x_i$  (variable) or  $\neg x_i$  (negated variable, also  $\overline{x_i}$ )
- A *clause* is a disjunction  $C = I_1 \vee \ldots \vee I_k$  of literals
- A boolean formula in *conjunctive normal form* (CNF) is a conjunction of clauses φ = C<sub>1</sub> ∧ ... ∧ C<sub>m</sub>
- Set of all CNF formulas:

$$\mathsf{CNFBOOL} := \left\{ \bigwedge_{i=1}^{m} \bigvee_{j=1}^{k(i)} \sigma_{i,j} \mid \sigma_{i,j} \text{ are literals} \right\}$$

• CNF formulas where the clauses contain only k literals: k-CNF

$$k\operatorname{-SAT} := \{ \langle \varphi \rangle \mid \varphi \in k\operatorname{-CNFBOOL} \text{ is satisfiable} \}$$

## *k*-SAT: NP-complete for $k \ge 3$

#### Lemma

- $\textcircled{0} 1-SAT, 2-SAT \in \mathsf{P}$
- 3-SAT is NP-complete.

#### Proof (Overview).

- First part: Exercise
- Second part:
  - → 3-SAT  $\in$  NP clear: 3-SAT  $\leq_m^p$  SAT (special case)
  - Then show CIRSAT  $\leq_m^p$  3-SAT
  - Figure a circuit  $c = (g_1, \dots, g_m)$ , construct a 3-CNF formula  $\psi_c$
  - Variables in formula: One for each input and each gate
  - $x_1, \ldots, x_N$  for inputs of circuit
  - $y_1, \ldots, y_m$  for gates
  - Clauses (size 3) enforce values of gates

## NP-completeness of 3-SAT

#### Lemma

2 3-SAT is NP-complete.

#### Proof (Details).

Gate g <sub>i</sub>	Clause	Semantics
$\perp$	$\{\overline{y_i}\}$	$y_i = 0$
Т	$\{y_i\}$	$y_i = 1$
Xj	$\{\overline{y_i}, x_j\}, \{\overline{x_j}, y_i\}$	$y_i \leftrightarrow x_j$
$(\neg, j)$	$\{\overline{y_i},\overline{y_j}\},\{y_i,y_j\}$	$y_i \leftrightarrow \overline{y_j}$
$(\wedge, j, k)$	$\{\overline{y_i}, y_j\}, \{\overline{y_i}, y_k\}, \{\overline{y_j}, \overline{y_k}, y_i\}$	$y_i \leftrightarrow (y_j \wedge y_k)$

- Finally, add  $\{y_m\}$
- All clauses together form  $\psi_{\textit{c}}$

# NP-completeness of 3-SAT (Cont.)

#### Lemma

2 3-SAT is NP-complete.

### Proof (Details, Cont.).

- If c is satisfiable, then also  $\psi_c$ :
  - Use assignment  $\alpha$  of c for  $x_1, \ldots, x_N$
  - Value  $f_c^{(j)}(\alpha)$  at gate  $g_j$  as value for  $y_j$
  - By construction, all clauses true, thus  $\psi_c$  satisfied
- If c not satisfiable, then neither  $\psi_c$ :

$$f_c(\alpha) = 0$$
 for all  $\alpha$ 

- Thus,  $f_c^{(m)}(lpha)$  always 0 ("top level gate")
- If all clauses satisfied,  $y_m = f_c^{(m)}(lpha)$ , but then  $\{y_m\}$  not satisfied
- Thus,  $c \in \mathsf{CIRSAT} \iff \psi_c \in \operatorname{3-SAT}$

## Graph problems

#### • So far: Satisfiability problems

- For boolean formulas (SAT, 3-SAT)
- For boolean circuits (CIRSAT)
- Now: Graph problems
  - Undirected graph: G = (V, E)
  - V are the nodes
  - $E \subseteq \binom{V}{2}$  are the *edges*
  - Efficient encoding possible (adjacency matrix or list)
- Problems consider different properties of graphs

# Independent Set Problem

- First problem: Independent set problem
  - Given: Undirected graph G = (V, E) and number k
  - Question: Is there  $I \subseteq V$  such that
    - **1** ||I|| = k, and
    - 2 No two nodes in I are connected?
- Formally:

### Definition

The *independent set problem* is defined as:

$$\mathsf{INDEPSET} := \left\{ (G,k) \mid \exists I \subseteq V(G) : \|I\| = k \land \binom{I}{2} \cap E(G) = \emptyset \right\}$$

• Turns out: Very difficult (i.e. NP-complete)

# INDEPSET is NP-complete

#### Lemma

INDEPSET is NP-complete.

Proof (Sketch).

- INDEPSET  $\in$  NP: Set I is the witness
- NP-completeness via 3-SAT  $\leq_m^p$  INDEPSET:
  - Given  $\varphi$  with k clauses, construct G
  - Each literal is a node
  - Connect literals from same clause (triangles)
  - Connect complementary literals
- If  $\varphi$  satisfiable:

Choose one satisfied literal in each clause for I

• If G has k-independent set:

Represents a satisfying truth assignment

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# CLIQUE is NP-complete

#### • Clique problem

- Given: Undirected graph G = (V, E) and number k
- Question: Is there  $C \subseteq V$  such that
  - $\|C\| = k?$
  - All nodes in C are pairwise connected (a "k-clique")
- Formally:

### Definition (Clique Problem)

The *clique problem* is defined as:

$$\mathsf{CLIQUE} := \left\{ (G,k) \mid \exists C \subseteq V(G) : \|C\| = k \land \binom{C}{2} \subseteq E \right\}$$

• NP-complete! (See exercises)

# NODECOVER is NP-complete

#### • Node cover problem

- Given: Undirected graph G = (V, E) and number k
- Question: Is there  $N \subseteq V$  such that

• Formally:

Definition (Node Cover Problem)

The node cover problem is defined as:

 $\mathsf{NODECOVER} := \{ (G, k) \mid \exists N \subseteq V(G) : \|N\| = k \land \forall e \in E(G) : e \cap N \neq \emptyset \}$ 

N

• NP-complete! (See exercises)

## HAMILTONPATH is NP-complete

#### • Hamilton path problem

- Given: Undirected graph G = (V, E)
- Question: Is there a path  $p = (p_0, \ldots, p_k)$  in G such that:
  - All nodes p<sub>i</sub> are pairwise different? ("Hamilton path")
- Formally:

#### Definition (Hamilton Path Problem)

The Hamilton Path Problem is defined as:

HAMILTONPATH := { $G \mid \exists p : p \text{ is Hamilton path in } G$ }

#### • NP-complete!
## HITTINGSET is NP-complete

#### • Hitting set problem

- Given:
  - ★ A set A
  - ★ A collection  $C = (C_1, \ldots, C_m)$  of subsets of A:  $\forall i : C_i \subseteq A$
  - ★ A number k
- Question: Is there a set  $H \subseteq A$  such that:
  - $\bigcirc ||H|| = k$
  - 2 *H* contains an element from each  $C_i \in C$  ("*k*-hitting set")
- Not a graph problem, but related
- Formally:

#### Definition (Hitting Set Problem)

The *hitting set problem* is defined as:

 $\mathsf{HITTINGSET} := \{ (A, C, k) \mid \exists H \subseteq A : \|H\| = k \land \forall C_i \in C : H \cap C_i \neq \emptyset \}$ 

• NP-complete! (See exercises)

# TSP is NP-complete

- Travelling Salesman Problem
  - Given:
    - ★ *n* cities with a distance matrix  $D \in \mathbb{N}^{n \times n}$
    - ★ A number k
  - Question: Is there a tour through all cities such that
    - Each city is visited exactly once, and
    - 2 The distance sum of the tour is at most k?
- Formally:

#### Definition (Travelling Salesman Problem)

The Travelling salesman problem is defined as:

$$\mathsf{TSP} := \left\{ (D,k) \mid \exists \pi : \sum_{i=1}^n D[\pi(i),\pi(i+1)] \le k \right\}$$

#### • NP-complete!

# KNAPSACK is NP-complete

#### • Knapsack Problem

- Given:
  - ★ *n* items with values  $V = (v_1, ..., v_n)$  and weights  $W = (w_1, ..., w_n)$
  - ★ A lower value limit / and an upper weight limit m
- Question: Is there a selection  $S \subseteq [1, n]$  of the items, such that
  - The sum of the values is at least *I*, and
  - 2 The sum of the weights is at most m?

• Formally:

#### Definition (Knapsack problem)

The Knapsack problem is defined as:

$$\mathsf{KNAPSACK} := \left\{ (V, W, I, m) \mid \exists S \subseteq [1, n] : \sum_{i \in S} w_i \le I \land \sum_{i \in S} v_i \ge m \right\}$$

#### • NP-complete!

# ILP is NP-complete

#### • Integer linear programming problem

- ▶ Given: *n* linear inequalities in *n* variables with integer coefficients
- Question: Is there an integer solution to that system?
- Formally:

#### Definition (Integer Linear Programming)

The Integer linear programming problem is defined as:

 $\mathsf{ILP} := \{ (A, b) \mid \exists x \in \mathbb{Z}^n : Ax \le b \}$ 

#### • NP-complete!

• Remark: Linear programming (allowing rationals) is in P!

# **BINPACK** is NP-complete

- Bin packing problem
  - Given:
    - \* *n* items with sizes  $A = (a_1, \ldots, a_n)$
    - ★ b bins with capacity c each
  - Question: Is it possible to pack the items into the bins?
- Formally:

### Definition (Bin Packing)

The bin packing problem is defined as:

$$\mathsf{BINPACK} := \left\{ (A, b, c) \mid \exists \text{ partition } S_1, \dots, S_b \text{ of } [1, n] \text{ s.t. } \forall i : \sum_{j \in S_i} a_j \leq c \right\}$$

#### • NP-complete!

## Between P and NPC

- Many problems are NP-complete
- Many problems are in P
- Assume  $P \neq NP$ :
  - Are all problems either P or NP-complete?
  - ► No!

#### Lemma

```
If P \neq NP, then there is a language L \in NP - (P \cup NPC).
```

(Without proof.)

- "Too easy" for NPC, "too difficult" for P
- Will see a candidate later

## Pseudo-polynomial complexity

- Precise problem formulation may make a difference
  - Example: Integer linear programming
  - In P without restriction to integers
- Representation of problem instances may also matter:
  - KNAPSACK is NP-complete
  - But: Given n items and weight limit l, solve it in time  $O(n \cdot l)$
  - Still not polynomial in input: Input size is O(n · log(l))
  - I is represented binary (or other k-ary)
- If polynomial in input values (not size/length): Pseudo-polynomial
- Strong NP-completeness:
  - NP-complete even if input values polynomially in input size
  - Equivalent: Input values are given in unary
  - Examples: All we saw except KNAPSACK
  - Don't have pseudo-polynomial algorithms (unless P = NP)

# Unknown Relations

- NP- and co- NP-complete problems: Regarded as "difficult"
- How do they relate to each other?
  - ► Unknown: NP <sup>?</sup> = co- NP
  - ▶ NP  $\neq$  co- NP  $\implies$  P  $\neq$  co- NP, thus NP  $\neq$  co- NP stronger
  - Intuition: Efficiently verifiable proofs, no efficiently verifiable disproofs
- What about  $NP \cap co-NP?$ 
  - Unknown:  $P \stackrel{?}{\subsetneq} NP \cap co-NP$
  - Intuition: Efficiently verifiable proofs and disproofs, not efficiently provable
- "Upper end": Also NP  $\stackrel{?}{\subsetneq}$  PSPACE unknown
  - Intuition: Provable in polynomial space, but no (time-)efficiently verifiable proofs
  - Even  $P \subsetneq PSPACE$  unknown

# The Problems PRIMES and GI

•  $\mathsf{P} \neq \mathsf{NP}$  not known, thus no language proven in  $\mathsf{NP} - (\mathsf{P} \cup \mathsf{NPC})$ 

#### • Former candidate: PRIMES

- Deciding primality of a number
- Efficient probabilistic methods were known
- Shown in 2002:  $PRIMES \in P$
- Another candidate: GI (graph isomorphism)
  - Given: Two graphs  $G_1$  and  $G_2$
  - Question: Are they isomorphic?

 $\mathsf{GI} := \{ (\mathsf{G}_1, \mathsf{G}_2) \mid \exists \pi : (e \in \mathsf{E}(\mathsf{G}_1) \iff \pi(e) \in \mathsf{E}(\mathsf{G}_2)) \}$ 

- Isomorphism: Graphs "look the same" (same structure)
- High practical relevance
- Many approximations, but no exact complexity known
- Own complexity class GI: Everything reducible to GI

# Course Outline

- Introduction
  - Basic Computability Theory
    - Formal Languages
    - Model of Computation: Turing Machines
    - Decidability, Undecidability, Semi-Decidability

#### Complexity Classes

- Landau Symbols: The  $\mathcal{O}(\cdot)$  Notation
- Time and Space Complexity
- Relations between Complexity Classes
- Feasible Computations: P vs. NP
  - Proving vs. Verifying
  - Reductions, Hardness, Completeness
  - Natural NP-complete problems
  - Advanced Complexity Concepts
    - Non-uniform Complexity
    - Probabilistic Complexity Classes
    - Interactive Proof Systems

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#### Complexity Classes

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- Relations between Complexity Classes
- 3 Feasible Computations: P vs. NP
  - Proving vs. Verifying
  - Reductions, Hardness, Completeness
  - Natural NP-complete problems
  - Advanced Complexity Concepts
    - Non-uniform Complexity
    - Probabilistic Complexity Classes
    - Interactive Proof Systems

# Uniform vs. Non-uniform Models

- Turing Machine: One fixed (finite) machine for all input sizes
- "One size fits it all" approach, uniform model
- Some situations: More hardwired information when size grows
  - Cryptography: Precomputed tables for different key sizes
  - Want to model such attackers
- Model this *non-uniform* notion using *advice*:
  - Machine gets advice string a<sub>n</sub> in addition to input
  - One fixed string a<sub>n</sub> for each input size n

# Turing Machine with Advice

#### Definition (Turing Machine with Advice)

- A Turing machine with advice is a 6-tuple M = (Q, Γ, δ, q<sub>0</sub>, F, A) with Q, Γ, δ, q<sub>0</sub>, F as before and A = {a<sub>n</sub>}<sub>n≥0</sub>
- The set a A is called the *advice*.
- Language accepted by *M*:

 $L(M) := \{x \in \Sigma^* \mid \exists y, z \in \Gamma^* : (\varepsilon, q_0, x \# a_{|x|}) \vdash^* (y, q_{yes}, z)$ 

(Separation symbol  $\# \in \Gamma$ )

### Turing Machine with Advice: Remarks

- Classical Turing machine: Finite object
- Advice: Infinite object, external information to machine
- Difference to witness from NP:
  - Witness was different for each input, a specific proof
  - Only for the positive instances  $x \in L(M)$
  - Advice is for each input size, a static computational "aid"
  - ▶ Fixed, the same *a<sub>n</sub>* for each *x* of length *n*
- Very powerful without restrictions:
  - Any language L (even undecidable!) could be decided
  - Encode into  $a_n$  a large table with all words  $\Sigma^n$
  - For each word x a bit:  $x \in L$  or  $x \notin L$
  - Machine can look up in table



# Restricting the Advice: P/poly

• Restrict now advice polynomially

#### Definition

P/poly is the set of all languages L such that:

- L is decided by a TM M with advice A exists, and
- 2  $\forall n : |a_n| \le p(n)$  for some polynomial p(n).
  - Clear:  $P \subseteq P/poly$  (just ignore the advice)
  - Thus: If  $\exists L \in NP$  with  $L \notin P$ /poly then  $P \neq NP$ !
  - Difficulty: P/poly is powerful, contains still undecidable languages

# P/poly and Undecidability

#### Lemma

P/poly contains undecidable problems.

#### Proof.

First: There are unary undecidable languages
Encode *H* using one-symbol alphabet
Second: All unary languages are in P/poly
For each size *n*, there is only one instance *x<sub>n</sub>*Set *a<sub>n</sub>* = 1 if it is positive (*x<sub>n</sub>* ∈ *L*)
Set *a<sub>n</sub>* = 0 if it is negative (*x<sub>n</sub>* ∉ *L*)

• However, under reasonable assumptions:  $NP - P/poly = \emptyset$ 

## Circuit Characterization of Non-uniformity

- Defined P/poly using advice
- Alternative: Use circuits
- Recall: Circuit C<sub>n</sub>
  - Gets input string  $\alpha \in \{0,1\}^n$ , and
  - Produces output bit  $f_c(\alpha)$
  - Defines an accepted language:

$$L(C_n) := \{ \alpha \in \{0,1\}^n \mid f_{C_n}(\alpha) = 1 \}$$

- Works only with fixed instance size
  - ► Thus extend to *circuit family*  $C = \{C_n\}_{n \ge 0}$ :  $L(C) := \{\alpha \in L(C_{|\alpha|})\}$
- Non-uniformity: New device for each input size

## Circuit Size

- Circuit family may have a non-finite representation (like advice)
- Very powerful without restrictions (like advice)
- Resource bound: size(C<sub>n</sub>)
  - Defined via number of gates
  - Creates language class:

 $\mathsf{DSIZE}(s(n)) := \{L(C) \mid \forall n \ge 0 : \mathsf{size}(C_n) \le s(n)\}$ 

• TM can be simulated with quadratic circuit size:

#### Lemma

```
\mathsf{DTIME}(t(n)) \subseteq \mathsf{DSIZE}(\mathcal{O}(t(n)^2))
```

(Without proof.)

# Circuit Characterization of P/poly

#### Lemma

 $P/poly = DSIZE(n^{O(1)})$ 

#### Proof.

- First: Let  $L \in P/poly$  by a TM M with advice A
  - For each *n*, construct circuit  $C_n$  simulating *M* on  $\Sigma^n$
  - $a_n$  is only polynomially big, can be hardwired

Thus:  $L \in \mathsf{DSIZE}(n^{O(1)})$ 

- Second: Let  $L \in \mathsf{DSIZE}(n^{O(1)})$  via circuit family C
  - Use a TM for evaluating circuits
  - Advice A is the (encoded) circuit family C
  - Thus:  $L \in P/poly$

# Probabilistic Computation

- Models so far: Had exactly one outcome
  - Classical deterministic Turing machine
  - Non-determinism with witness characterization
  - Circuits
- Relax this requirement: Introduce "coin-tosses"
  - At each step, machine can choose different next steps
  - Does so with a certain *probability*
  - Introduces uncertainty: Same input, different answers
- Produces still useful results if low error probability
  - Cryptography: Attacker may be happy to succeed 1% of the time
- Produces still useful results if *low error probability*
- Syntactically the same as non-deterministic machine, but:
  - ▶ Non-deterministic model: *Theoretic* model, implicit "∃" quantification
  - Randomized model: Practical model, may be executed directly

#### Notation

• For 
$$L \subseteq \Sigma^*$$
 let  $\chi_L : \Sigma^* \to \{0, 1\}$ :  
 $\chi_L(x) := \begin{cases} 1 & \text{if } x \in L \\ 0 & \text{if } x \notin L \end{cases}$ 

 $\chi_L$  is the characteristic function of L

- Assume from now on:
  - ► *M* outputs 1 for "accept"
  - M outputs 0 for "reject"
  - Thus, only q<sub>halt</sub> instead of q<sub>yes</sub> and q<sub>no</sub>
- Allows expressions like " $M(x) = \chi_L(x)$ "
- "Coin tosses"  $r \in \{0,1\}^*$  as a second argument: M(x,r)
  - Notation of probability:  $\operatorname{Prob}_r[M(x, r) = 1]$
  - If M probabilistic, then r implicit: Prob[M(x) = 1]

#### One-sided Error: The class RP

- Didn't define accepted languages yet
- First type: One-sided error
  - Positive instances may be judged wrong (output 0 or 1)
  - Negative instances are always correct (output 0)

#### Definition

RP is the class of all L for which a polynomially time-bounded, probabilistic TM M exists, such that:

$$x \in L \implies \operatorname{Prob}[M(x) = 1] \ge 1/2$$
  
 $x \notin L \implies \operatorname{Prob}[M(x) = 1] = 0$ 

• Symmetric behaviour: co- RP (always correct for  $x \in L$ )

## P and NP vs. RP

- Clear:  $P \subseteq RP$  (Machine is always correct, no coin tosses)
- $\mathsf{RP} \subseteq \mathsf{NP}$ :
  - RP: For  $x \in L$ , at least half the paths lead to 1
  - NP: For  $x \in L$ , at least one path leads to 1
  - ▶ Both produce 0 for all runs on  $x \notin L$
- Or, using the witness-based characterization of NP:

	NP	RP
$x \in L:$ $x \notin L:$	$\exists w : (x, w) \in R_L \\ \forall w : (x, w) \notin R_L$	$\frac{Prob_r[(x,r) \in R_L] \ge 1/2}{\forall r : (x,r) \notin R_L}$

# Error Bound Robustness

- Positive instances: Error probability 1/2
- Assume: Error probability 2/3
- Can be *reduced* to 1/2:
  - Run the machine twice
  - Accept, if one of the runs accepted
  - For  $x \notin L$ : Still never acceptance
  - For  $x \in L$ : Error probability  $(2/3)^2 = 4/9 < 1/2$
- In general, using more runs:

#### Lemma

If there is a polynomially time-bounded probabilistic TM M for L and a polynomial p(n) such that:

$$x \in L \implies \operatorname{Prob}[M(x) = 1] \ge 1/p(|x|)$$
  
 $x \notin L \implies \operatorname{Prob}[M(x) = 1] = 0$ 

Then  $L \in \mathsf{RP}$ .

(Without proof.)

## **RP: Error Bound Robustness**

- Seen: Very low acceptance ratio is "boosted" to 1/2
- Can "boost" even further, very close to 1 ("almost always"):

#### Lemma

For each  $L \in RP$  and each polynomial p(n), there is a polynomially time-bounded probabilistic TM M such that:

$$x \in L \implies \operatorname{Prob}[M(x) = 1] \ge 1 - 2^{-p(|x|)}$$
  
 $x \notin L \implies \operatorname{Prob}[M(x) = 1] = 0$ 

(Without proof.)

• Thus, two equivalent characterizations:

Very weak (1/p(|x|) bound) for proof obligations
 Very strong (1 - 2<sup>-p(|x|)</sup> bound) for proof assumptions

### Two-sided Error: The class BPP

• Now allow errors for positive and negative instances

#### Definition

BPP is the class of all L for which a polynomially time-bounded, probabilistic TM M exists, such that:

$$\forall x \in L : \mathsf{Prob}[M(x) = \chi_L(x)] \ge 2/3$$

• Notation from before:

$$x \in L \implies \operatorname{Prob}[M(x) = 1] \ge 2/3$$
  
 $x \notin L \implies \operatorname{Prob}[M(x) = 1] < 1/3$ 

• Symmetric definition: BPP = co- BPP

• Definition is again quite robust, regarding the 2/3

# **BPP: Error Bound Robustness**

#### Lemma (Weak characterization of BPP)

If there is a polynomially time-bounded probabilistic TM M for L, a polynomial p(n) and a computable function f(n) such that:

$$\begin{aligned} x \in L \implies \operatorname{Prob}[M(x) = 1] \geq f(|x|) + 1/p(|x|) \\ x \notin L \implies \operatorname{Prob}[M(x) = 1] < f(|x|) - 1/p(|x|) \end{aligned}$$

Then  $L \in BPP$ .

(Without proof.)

#### Lemma (Strong characterization of BPP)

For each  $L \in BPP$  and each polynomial p(n), there is a polynomially time-bounded probabilistic TM M such that:

$$\forall x \in L : \operatorname{Prob}[M(x) = \chi_L(x)] \ge 1 - 2^{-p(|x|)}$$

(Without proof.)

## Relations of BPP

- Clearly  $RP \subseteq BPP$  (no error for  $x \notin L$  with RP)
- Unknown relation between BPP and NP
- Note: Error rate can be made exponentially small
- "Efficient computation" nowadays often characterized with BPP
  - ▶ In fact, P = BPP is a popular conjecture

### Monte Carlo vs. Las Vegas

• Described machines always answer, sometimes wrong

- Monte Carlo Algorithms
- Contrast: Always answer right, but sometimes with "I don't know"
  - Las Vegas Algorithms
- Denote "I don't know" with " $\perp$ "

#### Definition

ZPP is the class of all L for which a polynomially time-bounded, probabilistic TM M exists, such that:

$$\forall x \in L : \mathsf{Prob}[M(x) = \bot] \leq 1/2$$
, and

$$\forall x \in L, r : M(x, r) \neq \bot \implies M(x, r) = \chi_L(x)$$

# More Class Relationships

#### Lemma

- $\bullet \mathsf{P} \subseteq \mathsf{ZPP} \subseteq \mathsf{RP} \subseteq \mathsf{BPP}$
- **2PP = RP \cap co- RP**
- $I BPP \subseteq P/poly$
- BPP = P if pseudo random number generators exist. (Efficient derandomization)

(Without proof.)

• Pseudo random number generators (PRNGs) output *looks* random

- No observer can tell the difference
- No observer can predict the next bit
- PRNGs exist under certain (sophisticated) assumptions

# Intuitive Notion of a Proof

- Proof: Prover and Verifier
- Prover convinces verifier of validity of some assertion
- In mathematics:
  - Prover writes down a list of steps
  - Verifier checks each step
- In general:
  - Interaction between the parties
  - Verifier asks questions (possibly adaptively)
  - Prover answers them
- Careful verifier is only convinced of valid assertions

# Formalizing the Notion of a Proof

• Interpret NP as non-interactive proofs:

- Supplied witness is the proof
- Machine checking it is the verifier, working *efficiently*
- Only true assertions ( " $x \in L$ ") have a proof
- "NP proof system"
- General notion with similar properties:
  - Efficiency of the verifier
  - ② Correctness requirement:
    - Completeness: Each true assertion has a convincing proof strategy Soundness: No false assertion has a convincing proof strategy
- Will use Interactive Turing Machines for that

## Interactive Turing Machine (ITM)

- Like ordinary Turing machine, but with
  - Communication tape and
  - Two communication states q<sub>?</sub> and q<sub>!</sub>
- Operation of two *composed machines*  $\langle M_1, M_2 \rangle$ :
  - $M_1$  starts in  $q_0$ ,  $M_2$  waits in  $q_?$
  - ▶ *M*<sub>1</sub> runs, writes message on *shared communication* tape
  - $M_1$  switches to  $q_2$ ,  $M_2$  switches to  $q_1$
  - M<sub>2</sub> runs, M<sub>1</sub> waits
  - .. And so on, back and forth ..
  - Finally, M<sub>2</sub> stops
- Period between control switches: A round
- Output: Tape contents of M<sub>2</sub> after halting

# Interactive Proof System

#### Definition (Interactive Proof System)

An *interactive proof system* for L is a pair  $\langle P, V \rangle$  of ITMs with:

- V is probabilistic and polynomially time-bounded.
- Orrectness requirement:

Completeness:  $\forall x \in L$ :  $Prob[\langle P, V \rangle(x) = 1] \ge 2/3$ Soundness:  $\forall x \notin L : \forall P^* : Prob[\langle P^*, V \rangle(x) = 1] < 1/3$ 

- Correctness is probabilistic
- Bounds:
  - Verifier is bounded
  - Prover is not bounded
- Soundness is against all provers (incl. very bad ones)
- Useful model for cryptographic protocols

# IP hierarchy

#### Definition (IP hierarchy)

Let  $r : \mathbb{N} \to \mathbb{N}$ .

- IP(r(n)) contains all L for with interactive proof systems ⟨P, V⟩ such that on common input x, at most r(|x|) rounds are used.
- IP contains all *L* having interactive proof systems:

$$\mathsf{IP} := \bigcup_r \mathsf{IP}(r(n))$$

### Properties of IP

- Clearly NP  $\subseteq$  IP:
  - Just one round
  - Prover just writes witness
  - Verifier checks it (even deterministically)
  - Thus, interaction necessary to gain expressiveness
- Also randomness necessary:
  - Let  $\langle P, V \rangle$  without random choices
  - P always knows exactly answer of V
  - Thus: Doesn't need to ask, can calculate answer itself
  - Therefore, only one final message needed
  - This is an NP proof system!
- Further:  $IP = IP(n^{\mathcal{O}(1)})$ 
  - V can only do polynomially many steps
  - Thus: Number of rounds polynomially bounded
# IP: Error Bound Robustness

• As for BPP, bounds can be minimized:

#### Lemma

For each  $L \in IP$  and each polynomial p(n), there is an interactive proof system  $\langle P, V \rangle$  with strong correctness properties: Completeness:  $\forall x \in L : Prob[\langle P, V \rangle(x) = 1] \ge 1 - 2^{-p(|x|)}$ Soundness:  $\forall x \notin L : \forall P^* : Prob[\langle P^*, V \rangle(x) = 1] < 2^{-p(|x|)}$ (Without proof.)

- Price for that:
  - Increased number of rounds (serial repetitions), or
  - Increased message sizes (parallel repetitions)

## Example: Graph Non-Isomorphism

#### Theorem

 $\mathsf{GNI} := \overline{\mathsf{GI}} \in \mathsf{IP}.$ 

### Proof (Sketch).

• Given  $G_1, G_2$ , show they are not isomorphic

Idea:

- "Shuffle" one of them
- Only possible to find which one it was, if  $G_1$ ,  $G_2$  not isomorphic

#### • Thus, protocol:

- V chooses  $i \in \{1,2\}$  and permutation  $\pi$
- V sends  $H := \pi(G_i)$ 
  - *P* finds *j* such that  $G_j$  is isomorphic to *H*, sends *j*
- V checks whether i = j
- If  $(G_1, G_2) \in \text{GNI}$ , *P* can find correct *j* (*P* is unbounded!)
- If  $(G_1, G_2) \notin \text{GNI}$ , P can only guess and fails with 50% chance

### IP: More properties

- Note: Protocol was just two rounds,  $\mathsf{GNI} \in \mathsf{IP}(2)$
- $\mathsf{GNI} \in \mathsf{co-NP}$ , not known whether in P or NP
- Indeed, IP is quite strong:

#### Theorem

```
(\mathsf{NP} \cup \mathsf{co-NP}) \subseteq \mathsf{IP}
```

```
2 IP = PSPACE
```

(Without proof.)

- Interesting extension: Zero knowledge proof systems
  - Verifier does not gain knowledge
  - Needs definition of "knowledge": Information that can not be efficiently computed
  - Used for secrecy properties in cryptography



# Course Outline

- Introduction
  - Basic Computability Theory
    - Formal Languages
    - Model of Computation: Turing Machines
    - Decidability, Undecidability, Semi-Decidability

### 2 Complexity Classes

- Landau Symbols: The  $\mathcal{O}(\cdot)$  Notation
- Time and Space Complexity
- Relations between Complexity Classes
- Feasible Computations: P vs. NP
  - Proving vs. Verifying
  - Reductions, Hardness, Completeness
  - Natural NP-complete problems
  - Advanced Complexity Concepts
    - Non-uniform Complexity
    - Probabilistic Complexity Classes
    - Interactive Proof Systems