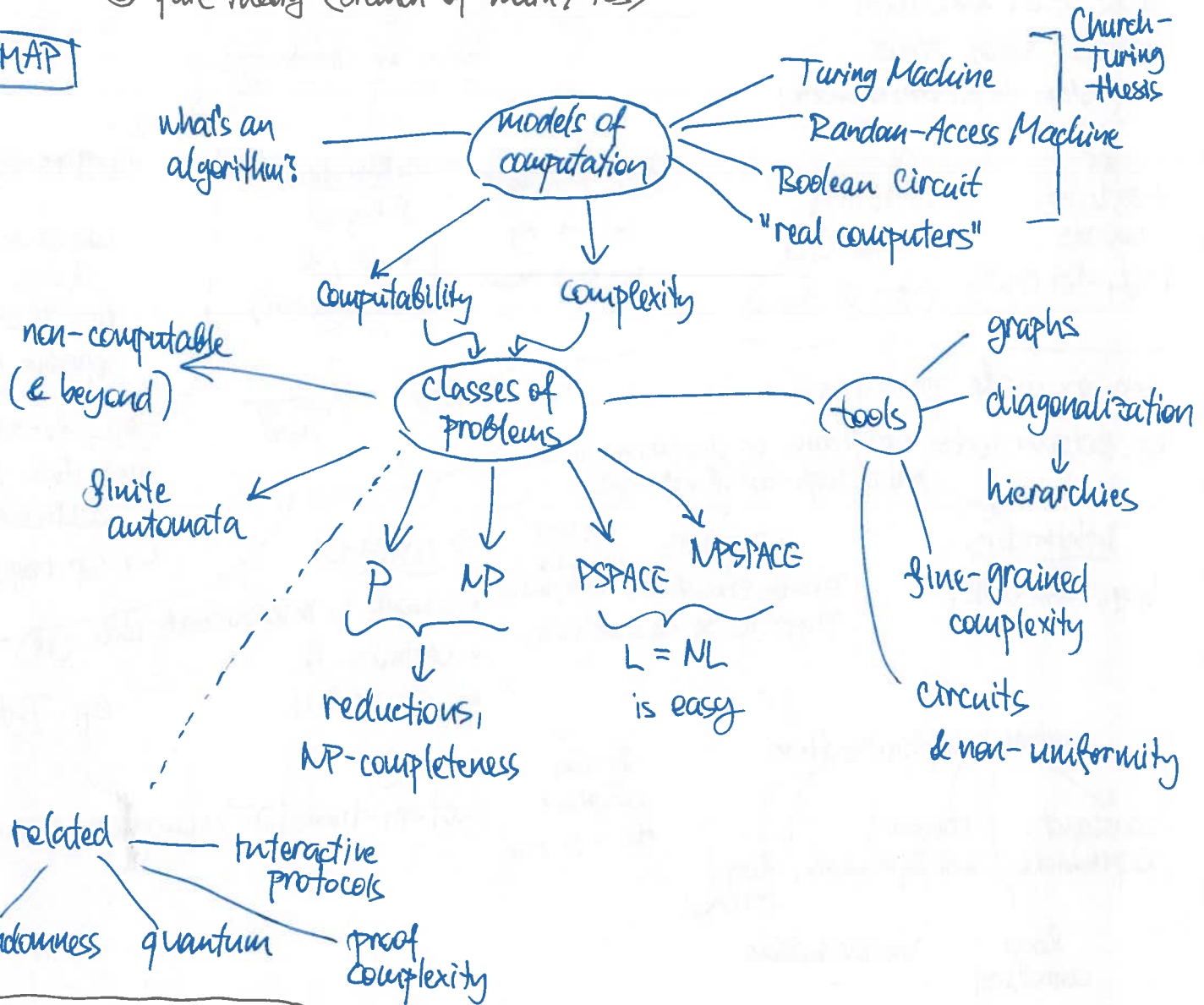


Automata & Complexity Theory winter 2022

goal: build theory of computation & hardness of problems

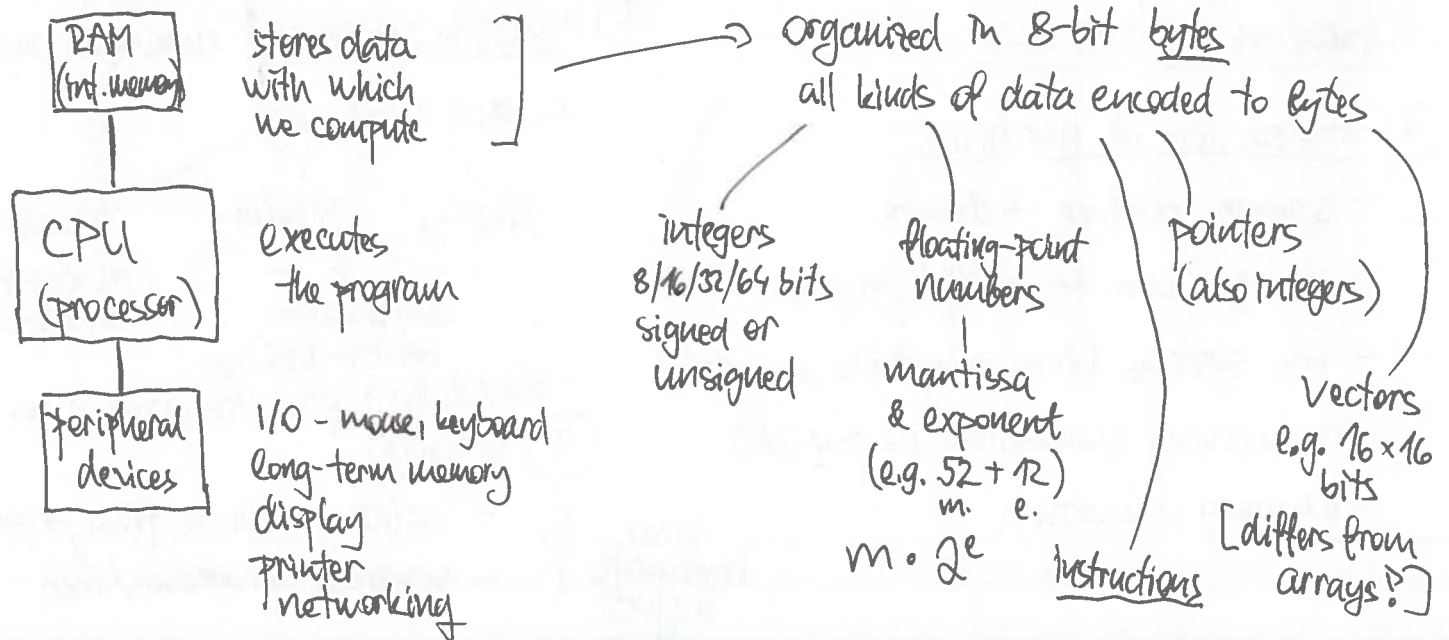
- two views:
- ① an applied theory to help us use machines more efficiently
 - ② pure theory (branch of math/TCS)

ROAD MAP



PHYSICAL COMPUTERS

& their architecture (typical case in 2022, just sketching)

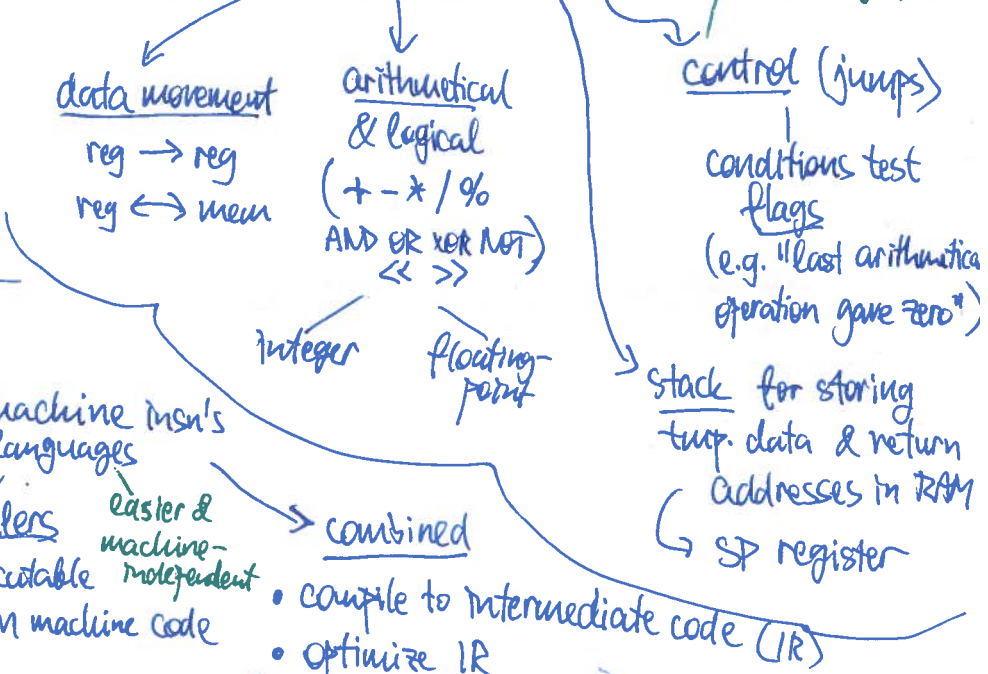


machine instructions (program) - stored in the same memory as data (Von Neumann architecture)
 - can modify itself
 - just another interpretation of bytes

they often work with several simple pieces of data (e.g., 64b numbers)

machine words ("64-bit CPU")
 registers inside CPU (10s of them)

types of instructions



see example program ...

We seldom write programs in machine insn's but in high-level languages

interpreters (e.g., UNIX shell)

compilers produce executable programs in machine code
 easier & machine-independent

combined

- compile to intermediate code (IR)
- optimize IR
- interpret IR

e.g. Python

automatic optimization

- constant expressions
- common sub-expressions
- loop reversal
- loop unrolling
- vectorization
- & many others

typical compiler: HLL → IR → MC

just-in-time (JIT) compilers - e.g. Java

Operating systems (OS)

① abstraction of hardware

- common interface → drivers
- manage access by multiple programs
- file systems (files, directories, mounts...)
- networking (connections vs. packets)
- memory allocation

② multiple processes "running at once"

- context switching
 - sleeping
 - yielding
 - timeslices (pre-emptive multitasking)
- cooperative multi-tasking
- scheduling, priorities, real-time

③ security

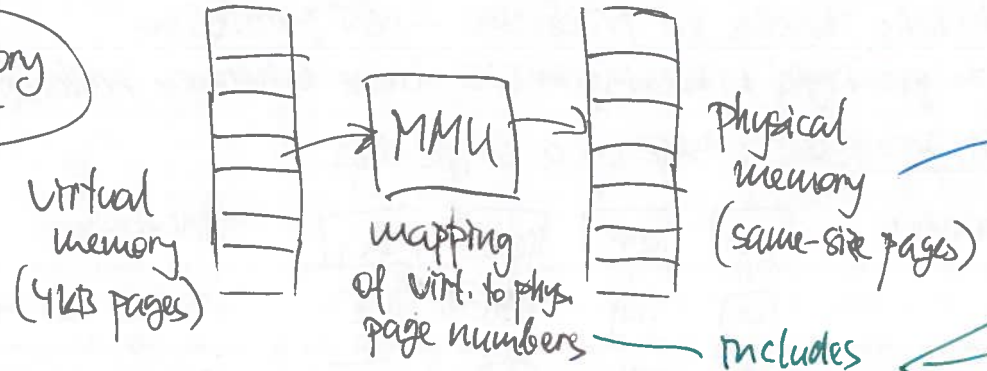
- Need hardware support
- isolate hardware from programs
 - separate programs/users

HW features for security

- privilege levels (supervisor/user mode)

→ system call instructions

Virtual memory management



Physical memory (same-size pages) → page fault exception

includes access rights

- read R
- write W
- execute code X

Supervisor only (or switching tables on mode changes...)

Uses:

- protection of processes (private memory) ... RW(x) for 1 process
- shared memory -- RW(x) in multiple processes
- shared library ... R in multiple processes
- lazy allocation ... read-only shared zeroes, copy on write
- fork ... using copy-on-write mapping
- SWapping ... store seldom-used pages to disk, read back on access

● caching - RAM is slow (CPU executes ~ 10⁹ instructions/second, RAM latency is tens of ns)

- idea: small, very fast memory inside the CPU which remembers frequently used data] called a cache | can be better only for small memory (speed of light etc.)

- caches 64B chunks of data (cache lines)
- strategy:
 - write-through vs. write-back
 - when cache fills up: evict least-recently used item (LRU)

- real caches have limited associativity → cache aliasing

- multiple levels of caches

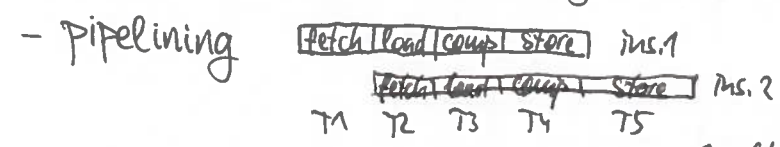
- example: accessing a matrix row by row vs. column by column

- ↳ sequential in memory
- ↳ every access is a cache miss → very slow

→ modelling of caches, cache-oblivious algorithms (not at this lecture)

● improving execution of instructions

- CPU works in cycles, historically 1 instruction took multiple cycles



all units of CPU always busy

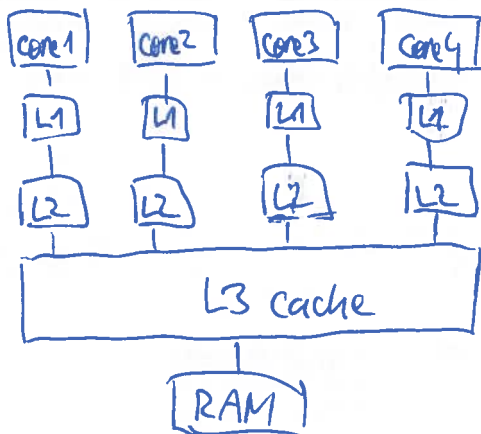
- e.g.:
- 1 fetch & decode
 - 2 load operands
 - 3 compute result
 - 4 store result

problems: dependencies & (conditional) jumps

- superscalar CPU: multiple units for different types of instructions, can run in parallel → scheduling instructions to units
 - jump prediction
- all this is transparent to SW (well, almost: Meltdown & other bugs)

- multiple processors sharing memory (SMP = Symmetric Multi-Processing)
 - OS schedules processes on processors - real parallelism
 - hard to get right: locking in SW, cache coherency protocols in HW
- multi-core processors: SMP on a single chip

For example:



typical sizes

32 KB code + 32 KB data

256 KB unified

8 MB unified

16 GB

- multi-threaded cores: two cores sharing their execution units & caches
 - unclear benefits (can even make things worse)
- virtual machines: simulating a whole machine within a process
 - including supervisor mode → the VM can run its own OS
 - including virtual peripherals
 - CPUs have special support for VMs (e.g., nested paging in MMU)

Relationship with theory

- will ignore most machine-dependent constants
- concentrate on asymptotics ⇒ all machines (roughly) equal
- use simple mathematical machines instead
- I/O and caches need special treatment

} rest of the semester

Models of Computation

history: beginning of 20th century: people asking for "mechanical procedures" for solving math problems - e.g., solving integer polynomial equations

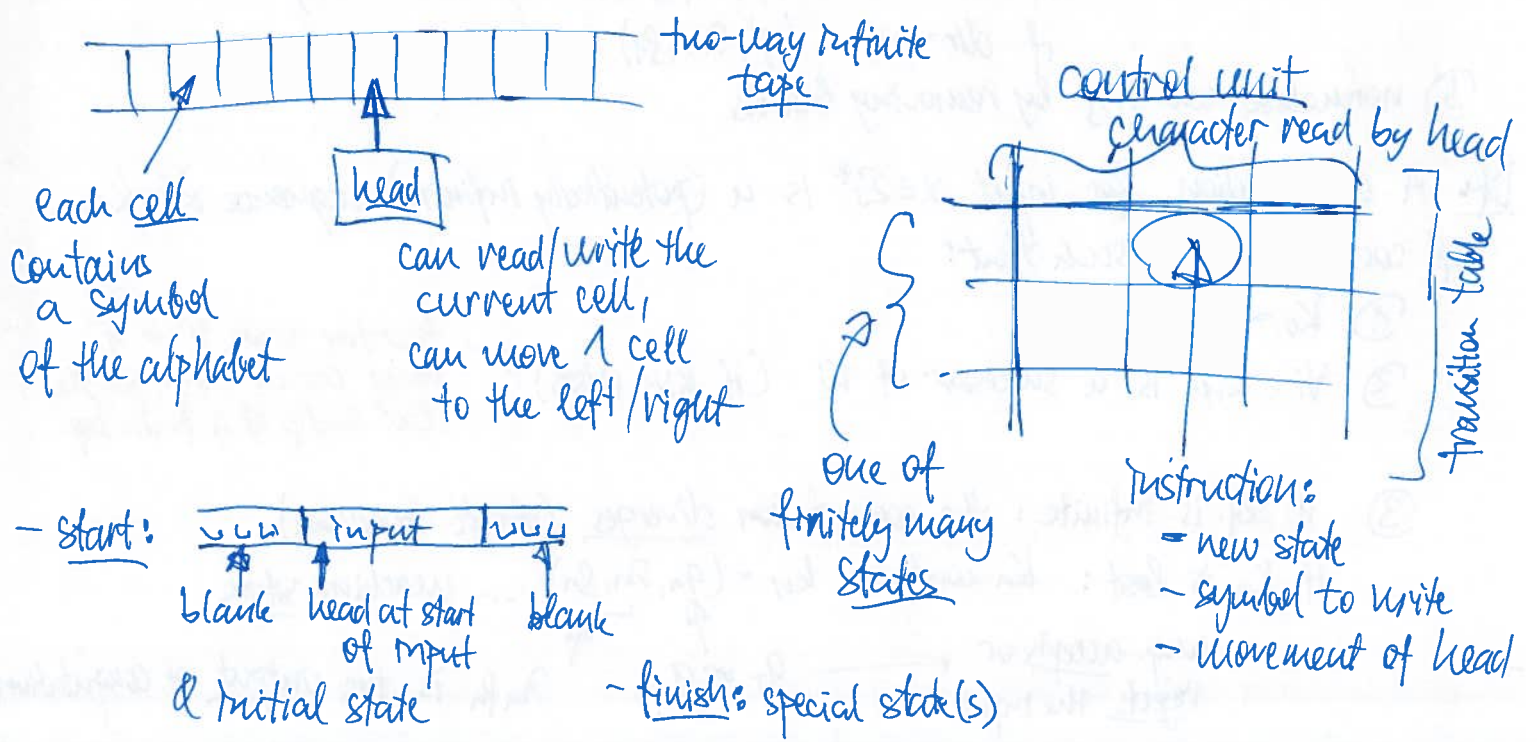
1930s: Gödel, Church, Kleene, Turing: formal definitions of computation (all of them equivalent, yet different)

What problems we want to solve?

↳ fix "language"

- Σ - finite alphabet of symbols (characters) - examples: $\{0,1\}$, $\{a \dots z\}$, math symbols
- Σ^* - set of all words (finite sequences) over Σ
 - ϵ ... empty word
 - $|x|$... length
 - $\alpha\beta$... concatenation
 - symbol \approx 1-symbol word
 - $\alpha[i]$... i-th symbol (starting with 0)
 - $\alpha[i:j]$ = $\alpha[i] \dots \alpha[j-1]$... subword
 - $\alpha[:j]$ = $\alpha[0:j]$... prefix
 - $\alpha[i:]$ = $\alpha[i:k]$... suffix
- problem: function from Σ^* to Σ^*
- decision problem: $f: \Sigma^* \rightarrow \{0,1\}$
 - ↳ also viewed as language $L \subseteq \Sigma^* : \alpha \in L \Leftrightarrow f(\alpha) = 1$ (characteristic function of a set)
- usually we find encoding of inputs (e.g. polynomials) to strings
 - concrete encoding doesn't matter (they can be converted algorithmically)
 - what happens if the input string is not a valid encoding?
 - ↳ suppose we always answer ϵ or 0 in such cases.

Turing Machines motivation: a mathematician with finite mind working on an infinite blackboard

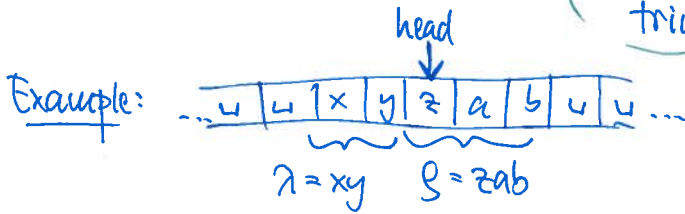
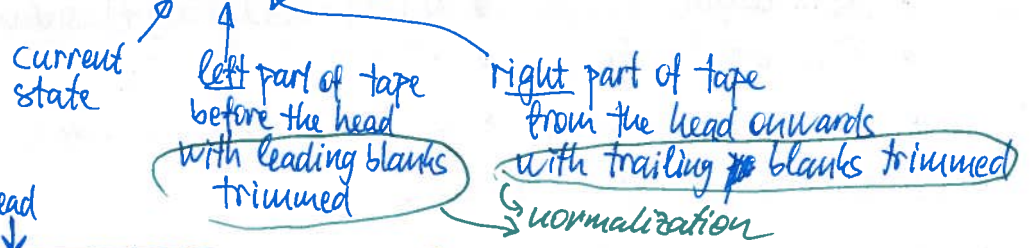


Now formally...

Df: A Turing machine consists of:

- Q ... a finite set of states
 - $q_0 \in Q$... initial state
 - $q_+, q_- \in Q$... final states (accepting & rejecting)
 - Σ ... non-empty finite input alphabet
 - $\Gamma \supseteq \Sigma$... finite work alphabet
 - $\sqcup \in \Gamma \setminus \Sigma$... blank symbol
 - $\delta : (Q \setminus \{q_+, q_-\}) \times \Gamma \rightarrow Q \times \Gamma \times \{\leftarrow, \circ, \rightarrow\}$... transition function
- } q_0, q_+, q_- all distinct

Df: Configuration of the TM: $(q, \lambda, \rho) \in Q \times \Gamma^* \times \Gamma^*$



$\lambda \rho$ = non-blank part of tape
 on empty tape, all positions of the head are the same configuration

Df: A configuration (q, λ, ρ) , $q \neq q_+, q_-$ has a successor defined in this way:

- ① extend λ by a leading \sqcup , ρ by a trailing \sqcup
 - ② now, $\lambda = \lambda'x, \rho = x\rho'$ for some x, λ', ρ'
 - ③ evaluate $\delta(q, x)$... get (q', x', dir)
 - ④ execute instruction:
 - if $dir = \circ$... $(q', \lambda, x'\rho')$
 - if $dir = \leftarrow$... $(q', \lambda', yx'\rho')$
 - if $dir = \rightarrow$... $(q', \lambda x', \rho')$
- } new config.
- ⑤ normalize new λ, ρ by removing blanks

Df: A computation for input $\alpha \in \Sigma^*$ is a (potentially infinite) sequence K_0, K_1, \dots of configurations such that:

- ① $K_0 = (q_0, \epsilon, \alpha)$
 - ② $\forall i: K_{i+1}$ is a successor of K_i (if K_{i+1} exists)
- ← therefore state q_+ or q_- never occurs except for the last config of a finite seq.

- ③ if seq. is infinite: the computation diverges (doesn't terminate)
 - if K_n is last: K_n contains $K_n = (q_n, \lambda_n, \rho_n)$... machine stops
- \swarrow comp. accepts or rejects the input q_+ or q_- $\lambda_n \rho_n$ is the output of computation

Def: Computability: ↗ also general recursive
 Function $f: \Sigma^* \rightarrow \Sigma^*$ is computable
 $\equiv \exists$ M Turing machine s.t.
 $\forall x \in \Sigma^* M(x)$ halts and outputs $f(x)$
 ↑
 M on input x
 (& its computation)

↗ also partially recursive divergence ↓
 Function $f: \Sigma^* \rightarrow \Sigma^* \cup \{\uparrow\}$ ($\uparrow \notin \Sigma$)
 is partially computable $\equiv \exists$ M T.M.
 s.t. $\forall x \in \Sigma^*$: if $f(x) = \uparrow$: $M(x)$ diverges
 else $M(x)$ halts & outputs $f(x)$

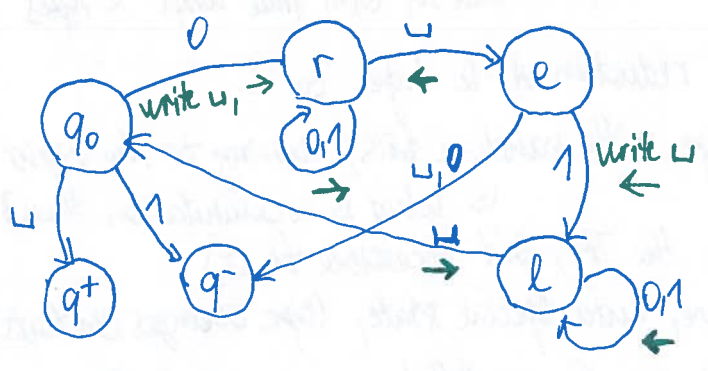
Language $L \subseteq \Sigma^*$ is computable
 $\equiv \exists$ M T.M. s.t. ↘ also recursive
 $\forall x \in \Sigma^* M(x)$ always halts
 & (accepts $x \iff x \in L$)
 ends in q^+
 ↪ equivalent to char. fn of L computable

Language $L \subseteq \Sigma^*$ is partially computable
 $\equiv \exists$ M T.M. s.t.
 $\forall x \in \Sigma^* M(x)$ halts $\iff x \in L$. ↓ also recursively enumerable
 in state q^+
 ↪ equivalent to $C_L(x) = \begin{cases} 1 & \text{if } x \in L \\ \uparrow & \text{if } x \notin L \end{cases}$ partially computable

Idea: Time & Space spent by computation
 ↑
 # configurations visited \neq # instructions executed
 # cells visited by the head

} will serve as basis for complexity theory (later)

Example: Recognizing $\{0^n 1^n\} \subseteq \{0,1\}^*$ by accepting/rejecting.
 Idea: erase first 0 & final 1, repeat.



$\Sigma = \{0,1\}$
 $\Gamma = \{0,1,\sqcup\}$
 $Q = \{q^0, q^+, q^-, r, e, l\}$

doesn't matter ↗

Q	0	1	⊔
q_0	(r, u, \rightarrow)	$(q^-, ?)$	$(q^+, ?)$
r	$(r, 0, \rightarrow)$	$(r, 1, \rightarrow)$	(e, \sqcup, \leftarrow)
e	$(q^-, ?)$	(l, \sqcup, \leftarrow)	$(q^-, ?)$
l	$(l, 0, \leftarrow)$	$(l, 1, \leftarrow)$	$(q_0, \sqcup, \rightarrow)$

Ideas:
 • Encode multiple variables with finite domains in the state - state is a tuple
 • Multi-track tape

} k tracks head reads/writes k-tuples
 but all tracks share head position
 ↑ each cell contains k-tuple
 Remember to en/decode tape at start/end of computation.

Variants of the TM (robustness of definition)

① One-way infinite tape ... 1-way \rightarrow 2-way trivial
 2-way \rightarrow 1-way "fold tape in half" simulation
 ↳ equally powerful (set of computable functions remains unchanged...)

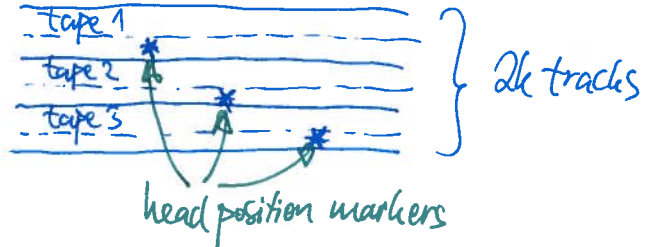
$\dots -2 \ -1 \ 0 \ 1 \ 2 \ \dots \rightarrow \begin{array}{|c|c|c|c|c|} \hline 0 & +1 & +2 & +3 & \dots \\ \hline -1 & -2 & -3 & -4 & \dots \\ \hline * & & & & \\ \hline \end{array} \left. \vphantom{\begin{array}{|c|c|c|c|c|}} \right\} 3 \text{ tracks}$
 * mark end of tape in track #3

state contains "positive half-tape" switch.

② k tapes with independent heads (but sharing a common work alphabet Σ)

- transition function: $(Q \setminus \{q^+, q^-\}) \times \Gamma^k \rightarrow Q \times \Gamma^k \times \{\leftarrow, \rightarrow, \bullet\}^k$
- configuration, successor, computation easy to extend
 - start: tape #1 contains input, other tapes empty
 - end: tape #1 contains output

• Equally powerful ... k-tape \rightarrow 1-tape:



1 step of orig. machine can be simulated by scanning the whole tape 2 times.

- find all heads, record symbols they see in state
 - write symbols back & move heads
- actually, you can do it in $O(n \log n)$ time (exercise)

• But complexity changes: $\{0^n 1^n\}$ requires super-linear time with 1 tape but can be solved in $O(n)$ time with 2 tapes

• Later: there is a more efficient reduction of k tapes to 2.

③ Randomized TM ... read-only tape with random bits, moving to the right only
 ↳ what is a computation then? ...

④ Oracles (functions defined outside the TM, but accessible to it)

- oracle tape: write query there, enter special state, tape changes contents to the answer

⑤ Interactive TM ... outside world can be modelled as an oracle \mathcal{O}

⑥ Exercise: 2-Dimensional tape ... head moves $\leftarrow, \rightarrow, \uparrow, \downarrow, \bullet$

... in some sense, the most powerful physically feasible computer is a TM with 3-D tape (or maybe 2-D only to allow heat spreading...)

- Exercises:
- accept strings in $\{0,1\}^*$ with even #1
 - reverse a string from $\{0,1\}^*$
 - add / multiply numbers written in binary ... what's the complexity?
 ↑ non-negative integers

Random-Access Machine

- formal model, but much closer to real hardware than the TM
- in fact, it's a family of related models, we will show the simplest of them
- RAM works with numbers (our version: the whole of \mathbb{Z})
- memory: seq. of numbers, indexed by numbers (negative indices allowed)
- addressing of operands:
 - literal constant (embedded in an instruction)
 - $[n]$ - directly addressed memory cell
 - $[[n]]$ - indirectly — " — (read $[n]$ to obtain another cell address)

- instructions:
 - ① movement of data
 $X \leftarrow Y$
Y = any, X = any except literal
 - ② arithmetics
 $X \leftarrow Y \oplus Z$
+, -, *, %, bitwise &, or, xor, bitwise shift
 - ③ control
 - halt
 - jump PLACE
 - if $X < Y$ jump PLACE
↑ $<, >, =, \neq, \leq, \geq$

- input is stored at agreed-upon locations in memory when the program starts
- output is found " " " " when the program stops

Example: sum of N numbers

In: $[0] = N, [1] = x_1, \dots, [N] = x_N$

Out: $[0] = \text{sum}$

Temporary: $[-1] = \text{copy of } N, [-2] = \text{current index}$

Program:

- $[-1] \leftarrow [0]$ copy N
- $[0] \leftarrow 0$ initialize sum
- $[-2] \leftarrow 1$ start with x_1

LOOP:

- if $[-2] > [-1]$ jump END
- $[0] \leftarrow [0] + [[-2]]$
- $[-2] \leftarrow [-2] + 1$
- jump LOOP

END: halt

Complexity: time = # executed instructions
 space = max (cell address used) - min (---)

this varies between RAM versions,
 e.g. we could define cost of an instruction as
 $\max \log(1+x)$
 $x \in \{\text{operands, addresses, result}\}$

"TM is equivalent to RAM"

- what can this mean?

- ... they can simulate each other: for each RAM program there is an equivalent TM & vice versa
- ... but RAM crunches numbers, while TM crunches strings

↓
 or, keep cost constant,
 but restrict size of cells
 somehow ...

We will assume that the RAM gets a string $\in \Sigma^*$ as input:

$[0]$ = length, $[1], [2], \dots$ = symbols of the string.

(this is WLOG since both TM and RAM can convert between all reasonable input formats)

(10)

TM to RAM - WLOG 1-tape TM with 1-way-infinite tape

- store the contents of the written-to part of the tape in $[1], [2], \dots$
- $[0]$ will specify how far the $\text{---} \cup \text{---}$ stretches.
- $[-1]$ = current position of head
- position in program represents machine state
- can simulate 1 step of the TM in constant time.

using some numbering of the work alphabet

RAM to TM

- representation of numbers: binary + sign symbol
- TM subroutines for arithmetics (inputs/output on special tapes)
- tape M: memory of the RAM $\text{---} \# \text{cell } -1 \# \text{cell } 0 \# \text{cell } 1 \# \text{---}$
- tape A: address of memory cell in which the head on tape M is
... can move 1 cell left/right, possibly extending M by empty cells at both ends
- memory read: given address on tape R, copy number read to tape D (data)
... compare R with A, move across cells until ~~data~~ $R=A$, copy data from M to D
- memory write: similar, but need to expand cells if they are too small for new data
- every instruction can be composed of read/write/arithmetics
- keep position in RAM program inside state of the TM
- simulation works, but with significant slowdown (inevitable?)

Computability

We will study it only for languages (decision problems), generalization to functions is straight-forward.

Df: Turing machine M accepts word $\alpha \in \Sigma^*$ \equiv computation on α ends in state q^+
 ☹️ rejects $\alpha \Leftrightarrow$ stops in q^- or runs forever (diverges)

- Language $L(M)$ accepted by $M \equiv \{ \alpha \in \Sigma^* \mid M \text{ accepts } \alpha \}$
- Language L is decided by $M \equiv M$ always stops & $L = L(M)$.

Df: Language L is computable (a.k.a. decidable/recursive) $\equiv \exists$ TM M : L is decided by M .
 \uparrow refers to Church's formalism of recursive functions (equivalent to TM)

• Language L is partially computable (a.k.a. partially decidable/recursively enumerable) $\equiv \exists$ TM M : L is accepted by M (i.e., $L(M) = L$).

Df: $R := \{ L \mid L \text{ is computable} \}$
 $RE := \{ L \mid L \text{ is partially computable} \}$

since elements of Σ can be arbitrary, these are proper classes.
 WLOG we can fix $\Sigma = \{0,1\}$ to make R and RE sets.

☹️ $R \subseteq RE \subseteq 2^{\{0,1\}^*}$ ← all languages over $\{0,1\}$
 $\uparrow \uparrow$ are these strict? Watch out...

Enumeration (or: why "recursively enumerable"?)

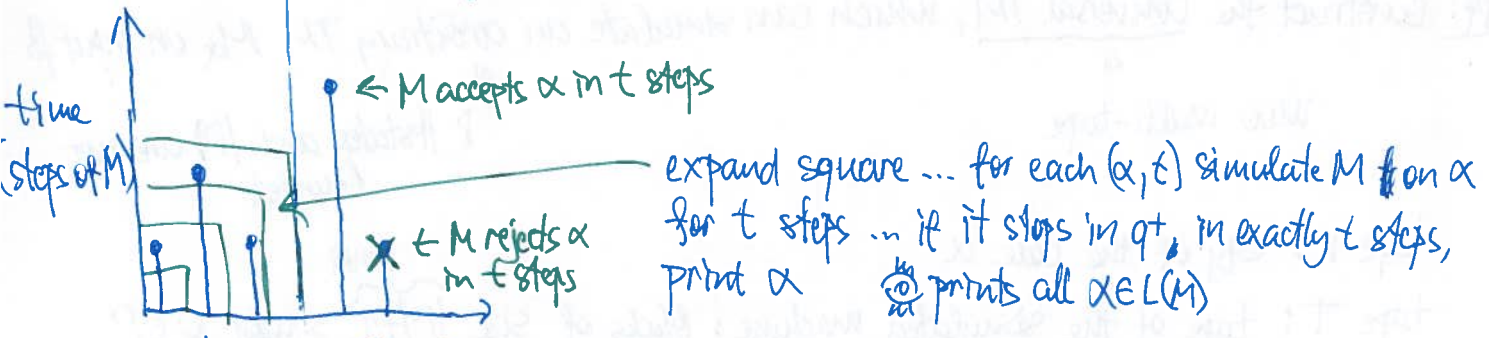
Df: Enumerator \equiv TM with no input, potentially running forever, printing strings (formally: printer is an oracle) } → language enumerated by M

L is enumerable $\equiv \exists$ enumerator which prints exactly the words of L

Thm: $L \in RE \Leftrightarrow L$ is enumerable

Pf: \Leftarrow we want to accept $\alpha \in L$... run enumerator, compare printed strings with α
 YES \Rightarrow stop in q^+ , NO \Rightarrow continue
 enumerator stops \Rightarrow stop in q^-

\Rightarrow we have TM M accepting L , let's build enumerator for L using M :



Strings in length-lexicographic order ($\alpha \leq_L \beta \equiv |\alpha| < |\beta| \vee |\alpha| = |\beta| \ \& \ \alpha \leq_{lex} \beta$)

Homework: $L \in RE \Leftrightarrow L$ is enumerable in \leq_{lex} order. [binary numbers with leading 1 removed]

Universal TM (why we don't need TM program in modifiable memory)

Df: Encoding of TMs (a.k.a. Gödel numbering) ← but in our case, the codes are actually strings, not numbers

we define it for 1-tape machines with $\Sigma = \{0,1\}$

alphabet: $\Gamma = \{x_0, x_1, x_2, \dots, x_m\}$
 0 1 2 other symbols in arbitrary order

states: $Q = \{q_0, q_1, q_2, \dots, q_n\}$
 initial q^+ q^- other states

directions: $\{d_0, d_1, d_2\}$
 ← → •

start code with $1^m 0 1^n 0$ to preserve $|\Gamma|$ and $|Q|$ even if symbols/states unused

transitions: $\delta(q_i, x_j) = (q_k, x_e, d_e) \rightarrow$ encode as $1^{i+1} 0 1^{j+1} 0 1^{k+1} 0 1^{e+1} 0 1^{t+1} 0$
 ↳ concatenate codes of all transitions → code of machine $\langle M \rangle$

Df: $M_\alpha :=$ machine with code α (if α not a valid code \Rightarrow machine which immediately halts in q^-)

⊙ $\forall TM M \exists \alpha : M \cong M_\alpha$

↳ isomorphism of TMs (defined in the obvious way)
 ↳ in fact, there are multiple such codes (we numbered Q, Γ arbitrarily etc.)

⊙ $L_\alpha := L(M_\alpha)$

⊙ $\forall L \in RE \exists \alpha : L = L_\alpha$

codes is countable \Rightarrow RE is countable ... but $2^{\{0,1\}^*}$ uncountable
 $\Rightarrow \exists L \notin RE$ (non-constructively)

Tool: Encoding of pairs $\langle \alpha, \beta \rangle : \langle x_1 - x_n, y_1 - y_m \rangle = x_1 0 x_2 0 \dots x_n 1 1 y_1 0 \dots y_m 0$
 ⊙ encoding & decoding is computable (& well-defined)

Df: Universal language $L_u := \{ \langle \alpha, \beta \rangle \mid \alpha, \beta \in \{0,1\}^* \text{ \& } \beta \in L_\alpha \}$
 "contains all partially computable languages" (in a sense)

Lemma: $L_u \in RE$

Df: Construct the universal TM, which can simulate an arbitrary TM M_α on input β
 ↳ #states and $|\Gamma|$ are not bounded
 ↳ M_u multi-tape

tape K: copy of the code α

tape T: tape of the simulated machine: blocks of size $|\Gamma|+1$, symbol $x_i \in \Gamma$
 stored as $1^i 0^{m-i}$

tape M: 1^m
 ↳ head on T encodes position of M_α 's head

tape S: current state of M_x stored as $1^j 0^{19-j}$

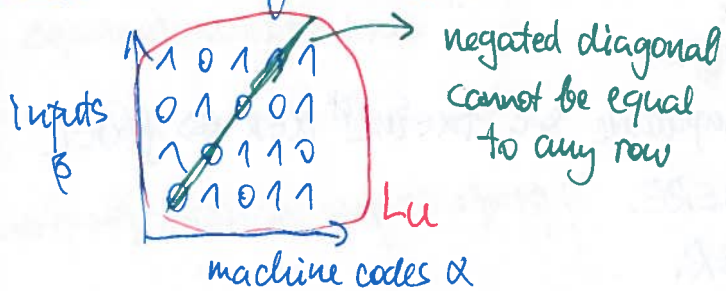
(13)

Init: Split $\langle \alpha, \beta \rangle$, copy α to tape K, encode β on tape T, initialize tape S
↳ & set tape M

Step: Read current symbol x from T, find entry for state s and symbol x on K, write new symbol & state, move head on T.

Lemmas: $L_u \notin RE$

Proof: Use diagonalization



diagonal language

$$L_d := \{ \alpha \in \{0,1\}^* \mid \alpha \notin L_\alpha \}$$

$L_d \notin RE \dots$ assume $L_d \in RE$

Then $\exists \alpha: L_d = L_\alpha$

but: $\alpha \in L_d \Leftrightarrow \alpha \notin L_\alpha \Leftrightarrow \alpha \notin L_d \quad \Downarrow$

If L_u were partially decidable, we could modify the machine accepting L_u to a machine accepting $L_d \quad \Downarrow$

Corollaries: • $L_u \notin R$ (R is closed under complement, so $L_u \in R$ would imply $\overline{L_u} \in R \subseteq RE$)

• $R \subsetneq RE \subsetneq 2^{\{0,1\}^*}$

↑ ↑
witnessed by L_u witnessed by L_u

Exercise: Are R and RE closed under \cap or \cup ?

• RE is not closed under complement

Thm (Post's): $L \in R \Leftrightarrow L \in RE \ \& \ \overline{L} \in RE$.

Pf: \Rightarrow trivial, because $R \subseteq RE$ & R closed under complement.

\Leftarrow "run machines accepting L and \overline{L} in parallel" (one step of each at a time)
One of them certainly stops.

Operations on machines codes

- swap q^+ with q^- : given M_x ^{deciding} ~~accepting~~ L_1 , find M_y deciding $\overline{L_1}$
- compose two machines: find M_y , which runs first M_x and then M_y on its output
- substitute M_x for an oracle in M_y

} all these are computable functions

More decision problems regarding machine codes

$L_{halt} := \{ \langle \alpha, \beta \rangle \mid M_{\alpha} \text{ halts on input } \beta \}$

$L_{empty} := \{ \alpha \mid L_{\alpha} = \emptyset \}$ $L_{total} := \{ \alpha \mid L_{\alpha} = \{0,1\}^* \}$

$L_{eq} := \{ \langle \alpha, \beta \rangle \mid L_{\alpha} = L_{\beta} \}$

Exercises: which of these (& their complements) are in R and/or RE?

Return to proof of $L_{halt} \notin RE$ via $L_d \notin RE$: "if we find a machine accepting L_d , we can use it to accept L_{halt} "

↳ let's generalize this.

Def: Many-to-one reduction between languages:

$K \leq_m L \equiv \exists f: \{0,1\}^* \rightarrow \{0,1\}^* \text{ computable s.t. } \forall x \in \{0,1\}^* x \in K \Leftrightarrow f(x) \in L$

\leq_m is a partial quasi-order on languages

Lemma: If $K \leq_m L$ and $L \in RE$, then $K \in RE$.

If $K \leq_m L$ and $L \in R$, then $K \in R$.

proof: compose machines for f and L

Corollary: If $K \leq_m L$ and $K \notin RE$, then $L \notin RE$. (Similarly $K \notin R \Rightarrow L \notin R$.)

Our original proof used $L_d \leq_m L_{halt}$ & $L_d \notin RE$ to show $L_{halt} \notin RE$.

Exercise: Find reductions between L_{halt} , L_{empty} , L_{eq} & their complements.

Examples: $L_{halt} \xrightarrow{\leq_m} L_{empty}$: given $\langle \alpha, \beta \rangle$, construct TM M_x which ignores its input & runs M_{α} on input β
↳ $L_y = \emptyset$ if $M_{\alpha}(\beta)$ diverges
↳ $L_y = \{0,1\}^*$ otherwise

$\Rightarrow (y \in L_{empty} \Leftrightarrow \langle \alpha, \beta \rangle \in \overline{L_{halt}})$

$L_{empty} \xrightarrow{\leq_m} \overline{L_{halt}}$: for given α , construct M_{β} which ignores its input, simulates M_{α} on all inputs in parallel & stops if $M_{\alpha}(\beta)$ stops on some β ...

$L \xrightarrow{\leq_m} M \Leftrightarrow \overline{L} \xrightarrow{\leq_m} \overline{M}$

Also, $L_{halt} \leq_m L_{total} \leq_m \overline{L_{halt}}$

$\langle \beta, \epsilon \rangle \in \overline{L_{halt}} \Leftrightarrow L_{\beta} = \emptyset \Leftrightarrow \beta \in L_{empty}$

Semantic properties of machines

Def: Property of languages: $P \subseteq RE \dots P$ is non-trivial $\equiv P \neq \emptyset$ & $P \neq RE$.
(semantic)

$L_P := \{ \alpha \in \{0,1\}^* \mid L_{\alpha} \in P \}$... all machines whose languages have the property P

Thm (Rice's): For every non-trivial property P , the language L_P is undecidable.

Proof idea: Show that $L_{halt} \rightarrow L_P$ for every non-trivial P .

Proof: Assume that $L_P \in R$ for some P .

WLOG $\emptyset \notin P$... otherwise use \bar{P} ... $L_{\bar{P}} = \bar{L}_P$, so it's also in R .

Find $L_w \in P$... exists as P is non-trivial

Reduction: if we want to answer $\langle \alpha, \beta \rangle \in L_{halt}$, i.e. if $M_x(\beta)$ halts

→ construct M_y which does on input δ :

this is computable

- run M_x on β (1)
- run M_w on δ (2)

👁 if $\langle \alpha, \beta \rangle \in L_{halt}$: (1) halts, (2) halts if $\delta \in M_w \Rightarrow L_y = L_w \in P$
 $\notin L_{halt}$: (1) diverges, (2) doesn't run $\Rightarrow L_y = \emptyset \notin P$

So this shows $L_{halt} \leq_m L_P$... but $L_{halt} \notin R$, so $L_P \notin R$.

What is the "hardest" language in a class?

- Let C be a set of languages.
- L is C -hard $\equiv \forall K \in C: K \leq_m L$
- L is C -complete $\equiv L$ is C -hard & $L \in C$

more precisely, it's C - m -hard (complete wrt. \leq_m)

Thm: L_u is RE -complete.

Pf: ① $L_u \in RE$

② for $K \in RE$, we find $\alpha: L_\alpha = K$

Then f reducing K to L_u is $\beta \mapsto \langle \alpha, \beta \rangle$

"Natural" undecidable problems (not directly involving machines)

- given a set of axioms and a formula φ , is φ provable?
- given a system of multi-variate polynomial equations over \mathbb{Z} , does it have a solution in \mathbb{R} ? → Matijasević theorem

both in $RE \setminus R$ (in suitable encoding)

& many more (e.g., plane tiling)

Relative computability

- given any language $A \subseteq \{0,1\}^*$, we can define ~~oracle~~ TM with an oracle giving access to A (see section on TM extensions)
- we can define relative language classes $R[A]$ and $RE[A]$
- we also have $M_x[A], L_x[A], L_u[A]$

👁 If $A \in R$, then $R[A] = R$ and $RE[A] = RE$ (in particular for $A = \emptyset$)

👁 Previous arguments about plain TM can be trivially relativized,

so in particular: $R[A] \neq RE[A] \neq \mathbb{Q}^{20,13^*}$

$R[A] = RE[A] \cap co-RE[A] \leftarrow co-T = \{L \mid \bar{L} \in T\}$

→ And also $L_u[A]$ is $RE[A]$ -complete

Def: Arithmetical hierarchy: classes $\Sigma_n, \Pi_n, \Delta_n$ for $n \in \mathbb{N}$

- $\Sigma_0 = \Pi_0 = \Delta_0 = R$
- $\Sigma_{n+1} = RE[\Sigma_n]$
- $\Pi_{n+1} = co-RE[\Pi_n]$
- $\Delta_{n+1} = R[\Sigma_n]$

We have:

$$\left. \begin{aligned} \Sigma_n &= RE \\ \Pi_n &= co-RE \\ \Delta_n &= R[R] = R \end{aligned} \right\} \begin{array}{l} \text{oracles from } R \\ \text{do not add} \\ \text{power to TM} \end{array}$$

$$\Sigma_n \subseteq \Sigma_{n+1}$$

this means:

$$RE[C] = \bigcup_{L \in C} RE[L]$$

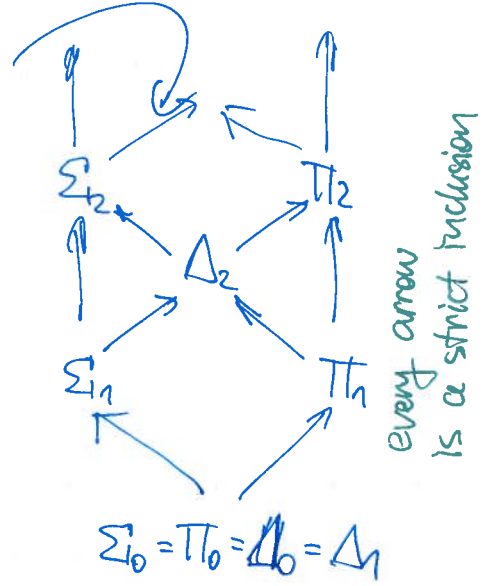
this inclusion is strict &

$$\begin{aligned} L_u^1 &= L_u \\ L_u^{n+1} &= L_u[L_u^n] \end{aligned}$$

L_u^n is Σ_n -complete
(by induction: $RE[\Sigma_n] = RE[L_u^n]$,
so $L_u[L_u^n]$ is Σ_{n+1} -complete)

~~$RE[L_u^n] \in \Sigma_n$~~
otherwise we would have
 $\Sigma_n \subseteq R[\Sigma_n] = R[L_u^n] \subsetneq RE[L_u^n] = \Sigma_{n+1}$

Also: $\Sigma_n \subsetneq \Delta_{n+1} \dots$ and this is strict as Σ_n is not closed under complement, while Δ_{n+1} is
 $\Pi_n \subseteq \Delta_{n+1} \dots$ we can negate oracle's answer
 $\Delta_{n+1} = \Sigma_{n+1} \cap \Pi_{n+1} \dots$ relative Post's thm.
 $\Delta_{n+1} \subseteq \Sigma_{n+1} \dots$ this is $R[L_u^n] \neq RE[L_u^n]$
 $\Delta_{n+1} \subseteq \Pi_{n+1} \dots$ analogous for co-RE



Quantified formulas

- every language in R can be interpreted as a predicate $\varphi(x)$ with string parameter φ - decidable predicates
- $\forall \beta \varphi(\beta) \equiv \exists \alpha \varphi(\alpha, \beta)$ lies in RE
 ... and every $L \in RE$ can be written in this way
 $[\alpha = \# \text{ steps after which a machine stops}]$
- $\varphi(\beta) \equiv \forall \alpha \varphi(\alpha, \beta) \dots$ this is co-RE ($\neg \forall \alpha \varphi(\alpha, \beta) \Leftrightarrow \exists \alpha \neg \varphi(\alpha, \beta)$)
 ... $\exists \alpha_1 \exists \alpha_2 \varphi(\alpha_1, \alpha_2, \beta)$ is again RE ... we can say $\exists \alpha$ s.f. $\alpha = \langle \alpha_1, \alpha_2 \rangle$
 & decode α inside φ
- $\exists \alpha_1 \forall \alpha_2 \varphi(\alpha_1, \alpha_2, \beta)$ is Σ_2

in general: $\exists \alpha_1 \forall \alpha_2 \exists \alpha_3 \dots Q_n \alpha_n \varphi(\alpha_1 - \alpha_n, \beta)$ is Σ_n
 $\forall \alpha_1 \exists \alpha_2 \forall \alpha_3 \dots Q_n \alpha_n \varphi(\alpha_1 - \alpha_n, \beta)$ is Π_n

$\in \Sigma_2$: $\exists \alpha_1 \forall \alpha_2 \varphi(\dots) \Leftrightarrow \exists \alpha_1 \neg (\exists \alpha_2 \neg \varphi(\dots))$ ← so this is in $RE[\Sigma_1] = \Sigma_2$
 can be answered by oracle $L_u \in \Sigma_1$

$\in \Sigma_2$ consider $L \in \Sigma_2 = RE[\Sigma_1]$:
 $\exists \gamma$ computation of TM M_γ on L_u $\exists \delta$ queries for L_u (γ, δ consistent, & answers for δ true)

see next page

Dependence of complexity on #tapes

Df: $L_{PAL} = \{ \alpha \alpha^R \mid \alpha \in \{0,1\}^* \}$ "even palindromes"
reversed

- trivial 2-tape TM deciding L_{PAL} in $\Theta(n)$ time.
- but all machines we found with 1 tape run in $\Theta(n^2)$ time!

Thm: Every 1-tape machine deciding L_{PAL} runs in time $\Omega(n^2)$.

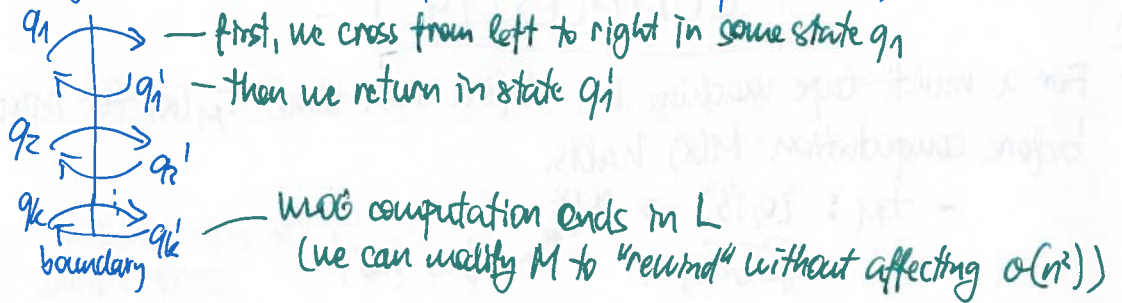
Proof: Assume there is M deciding L_{PAL} s.t. $T_M(n) \in o(n^2)$. \leftarrow that is: $\forall \epsilon > 0 \exists n : T_M(n) < \epsilon n^2$.

• Consider inputs of type:

part L	part Z	part R
α	$00 \dots 0$	α^R
$n/3$	$n/3$	$n/3$

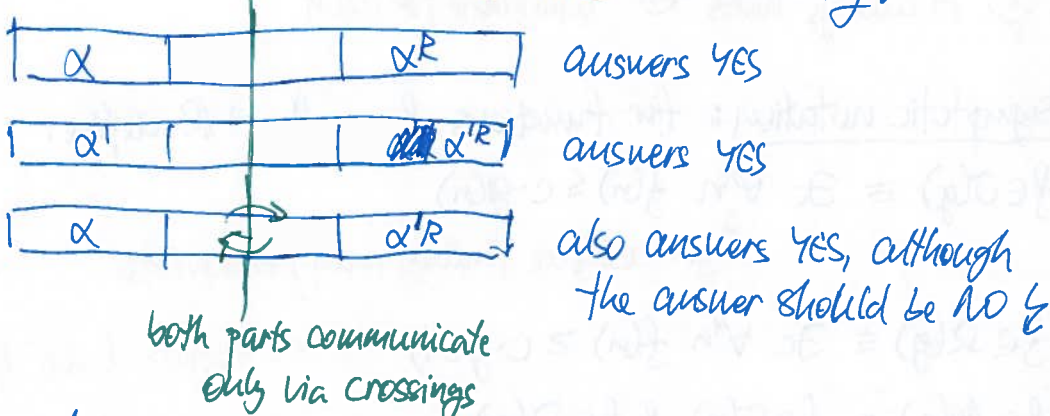
 \leftarrow answer is YES for every such string for $6/n$

• Consider boundary between 2 zeros in part Z & how computation of M crosses it:



\hookrightarrow crossing sequence $(q_1, q_1'), \dots, (q_k, q_k')$

• If two inputs with α, α' have the same C.S. for the same boundary:



• Similarly if they have same C.S. for different boundaries.
 (part Z can have odd length, but that implies NO anyway)

• Let's use P.M.P. (Pigeon-hole principle) to show that such α, α' exist:

c.s. of length $k = |Q|^{2k}$

c.s. of length at most $k \leq c \cdot |Q|^{2k}$ for some constant c
 (via sum of geom. series)

If #c.s. $<$ #inputs, then \exists two inputs with the same C.S.

\hookrightarrow so we want $c \cdot |Q|^{2k} < 2^{n/3} \dots 2^{\frac{\log c + 2k \log |Q|}{\leq 3k \log |Q|}} < \frac{n}{2^3} \dots k < \frac{n}{9 \log |Q|}$

(continued) We want to show that $\forall L \in \Sigma_2$ there is an equivalent formula $\exists \forall \dots$ (17)

- let M be a TM $[\Sigma_1]$... that is TM $[L_1]$ accepting L
- formula: $\exists \delta$ (check that δ is consistent with input α , with oracle L_1 , and with itself)

↑
computation of M
including all oracle
queries & answers

↑
decidable

↑
decidable

together:
 $\exists \delta (\exists \gamma \forall \epsilon \varphi(\alpha, \delta, \gamma, \epsilon))$

↑ code of $\gamma_1 - \gamma_k$ ↑ code of $\epsilon_1 - \epsilon_k$

for positive answers: $\exists \gamma_1 - \gamma_k$

$\varphi(\gamma_1, \text{question 1})$
 \vdots
 $\varphi(\gamma_k, \text{question k})$

for negative answers:
 $\forall \epsilon_1 - \epsilon_k \neg \varphi(\epsilon_1, \dots)$

Σ_1 -formula for L_1

$\& \dots$

... but this is $\exists \langle \delta, \gamma \rangle \forall \epsilon \varphi(\dots)$ as we need.

COMPLEXITY

- For a multi-tape machine M , define run time $t_M(x)$ for input x as #steps before computation $M(x)$ halts.

- $t_M: \underbrace{\{0,1\}^*}_{\text{generally } \Sigma^*} \rightarrow \mathbb{N}^*$ $\leftarrow \mathbb{N} \cup \{\infty\}$ for divergent computations

- Time complexity of machine M : $T_M: \mathbb{N} \rightarrow \mathbb{N}^*$ s.t. $T_M(n) = \max_{\alpha \in \Sigma^n} t_M(\alpha)$.
 \odot M always halts $\Leftrightarrow T_M(n)$ finite for all n .

Asymptotic notation: for functions $f, g: \mathbb{N} \rightarrow \mathbb{R}$ define:

① $f \in O(g) \equiv \exists c \forall^* n \ f(n) \leq c \cdot g(n)$ ← asymptotic upper bound
 \uparrow for all but finitely many exceptions

② $f \in \Omega(g) \equiv \exists c \forall^* n \ f(n) \geq c \cdot g(n)$ ← asymp. lower bound

③ $f \in \Theta(g) \equiv f \in O(g) \ \& \ f \in \Omega(g)$ ← both at once
 ... that is, $\Theta(g) = O(g) \cap \Omega(g)$

④ $f \in o(g) \equiv \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$ ← strict upper bound

⑤ $f \in \omega(g) \equiv g \in o(f)$ ← strict lower bound

Examples: $f: n \mapsto 5n^3 - 7n^2 + 18 \in O(n^3), O(n^4), O(2^n) \in o(n^4)$
 $\in \Omega(n^3), \Omega(n^2), \Omega(1) \in \omega(n^2)$

$O(1)$ = "bounded by constant"

$\in \Theta(n^3)$ "drop lower-order terms & multiplicative constant"

P. H. P. once again (we can find boundary with a small # crossings):

- we have $n/3$ boundaries
- \sum of lengths of their C.S. $\leq T_M(n) < \epsilon n^2$
- $\Rightarrow \exists$ boundary with at most $\frac{\epsilon n^2}{n/3} = 3\epsilon n$ crossings

now set ϵ such that we have:

$\text{min. \# crossings} \leq 3\epsilon n < \frac{n}{9 \log(n)}$ such ϵ exists & inequality satisfied for n large enough

so there are 2 inputs with the same C.S. \Rightarrow mixing produces contradiction.

Thm: For every multi-tape TM M there is 1-tape TM M' s.t. $L(M) = L(M')$ & $T_{M'}(n) \in O(T_M(n)^2)$.

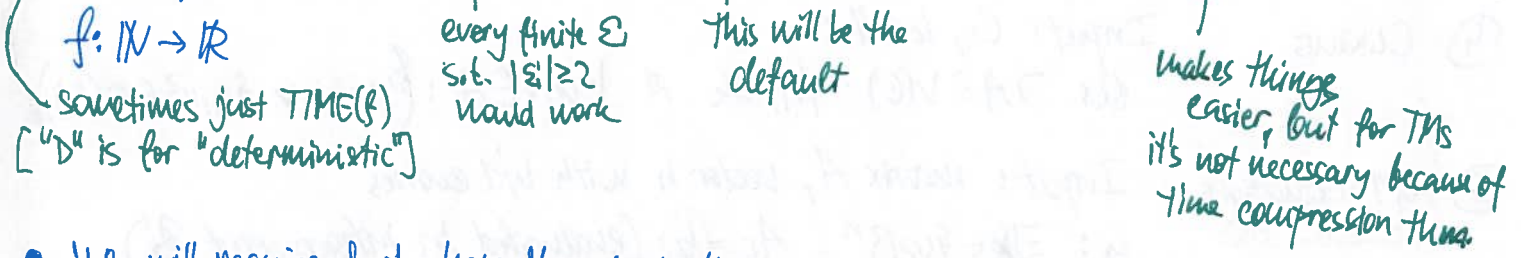
Proof: Analyze reduction from previous lectures. [hint: at most $T_M(n)$ tape cells used on each tape]

Thm: For every multi-tape TM M there is 2-tape TM M' s.t. $L(M) = L(M')$ & $T_{M'}(n) \in O(T_M(n) \cdot \log T_M(n))$.

(proof omitted)

Time complexity classes

$\text{DTIME}(f) := \{ L \subseteq \{0,1\}^* \mid \exists M \text{ multi-tape TM deciding } L \text{ in time } O(f) \}$



We will require f to have these properties:

- 1) non-decreasing
 - 2) $\forall n, f(n) \geq n$
 - 3) time-constructible $\equiv \exists \text{ TM } M_f \text{ which for input } 1^n \text{ produces output } 1^{f(n)} \text{ in time } O(f(n))$
- "proper time complexity function"

$P := \bigcup_{i \geq 1} \text{DTIME}(n^i)$ \Leftarrow polynomial-time decidable languages

Why we like P :

Examples:

- reachability in graphs
- evaluation of Boolean formulas
- corresponds (roughly) to "efficiently solvable"
- independent of model of computation (RAM gives the same P as TM)
- ~~polynomials~~ polynomials are the smallest set of functions containing constants & identity and closed under addition, multiplication and composition.

Classes of functions

- $DTIME(f) := \{g : \Sigma^* \rightarrow \Sigma^* \mid \exists TM \text{ computing } g \text{ in time } O(f)\}$
- $PF := \bigcup_{n \geq 1} DTIME(n^1)$

if $f(n) \in poly(n) : L_g := \{ \langle x, i \rangle \mid i\text{-th bit of } g(x) \text{ is } 1 \}$
 $L_g \in P \iff g \in PF$

$O(n^k)$ for some fixed k

encoded in binary

So studying only languages in P is WLOG.

Consider these problems:

as usually: suitably encoded

path vs. walk \leftarrow can repeat
 doesn't repeat vertices

① HAMILTON PATH

Input: undirected graph G , vertices u, v
 Question: \exists path in G with endpoints u, v containing all vertices (exactly) once.

② 3-COLORING

Input: undirected graph G
 $Q: \exists f: V(G) \rightarrow \{1, 2, 3\}$ s.t. $\forall \{u, v\} \in E(G): f(u) \neq f(v)$
 coloring of G with 3 colors

③ INDEPENDENT SET

Input: undirected graph $G, k \in \mathbb{N}$
 $Q: \exists A \subseteq V(G): |A| \geq k \ \& \ \forall u, v \in A: \{u, v\} \notin E(G)$

④ CLIQUE

Input: $G, k \in \mathbb{N}$
 $Q: \exists A \subseteq V(G): |A| \geq k \ \& \ \forall u, v \in A: \{u=v \vee \{u, v\} \in E(G)$

⑤ 0,1-Equations
 a.k.a. 2SE

Input: matrix A , vector b with 0/1 entries
 $Q: \exists x \in \{0, 1\}^n: Ax = b$ (evaluated in integers, not \mathbb{Z}_2)
 WLOG $b = 1$

⑥ SAT (boolean satisfiability): Input: Boolean formula $\varphi(x_1 \dots x_m)$ in CNF

$Q: \exists x_1 \dots x_m \in \{0, 1\}$ s.t. $\varphi(\vec{x})$ is true.

For all these: we are looking for something we can easily recognize (poly-time), but we don't know how to find it in poly time

Clause
 $\varphi \equiv (x_1 \vee x_2 \vee x_3) \wedge (\dots) \wedge (\dots)$

Reductions will again prove themselves useful:

Def: For languages $K, L: K \leq_m^P L \iff \exists f \in PF$ s.t.
 $\forall x \in \Sigma^* \ x \in K \iff f(x) \in L$

polynomial-time many-to-one reduction

Literals: either x_i or $\neg x_i$
 (restriction to CNF is WLOG, see later)
 (we cannot use thm. from Logic about equivalent formulas in CNF, because it blows up size exponentially)

Properties of \leq_P :

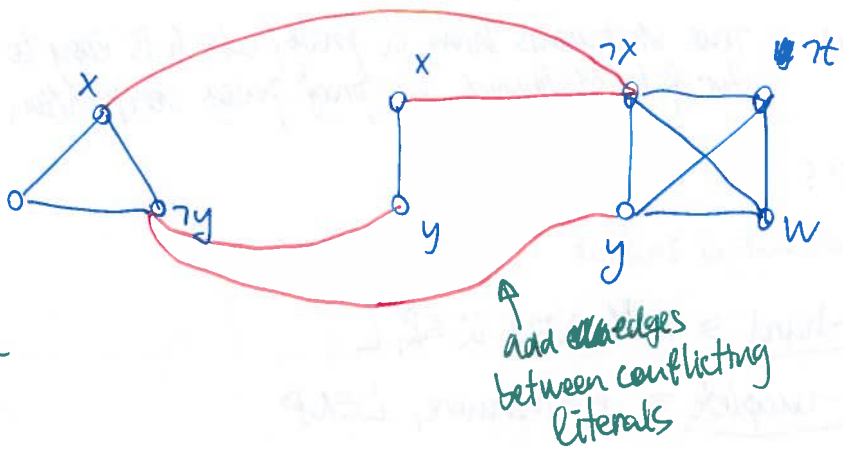
- reflexive & transitive (quasi-order)
- \exists incomparable languages (exercise)
- $K \leq_P L$ and $L \in P \Rightarrow K \in P$ [composition of 2 algorithms running in poly. time is again poly-time]

size of input for ARG_2 is at most time complexity of ARG_1

Example: SAT \leq_P INDEPSET

$$((x) \vee (\neg y) \vee (z)) \wedge (x \vee y) \wedge (\neg x \vee y \vee \neg z \vee w)$$

each clause gets complete subgraph labelled with literals of the clause



add edges between conflicting literals

given a formula

produce a graph, $k := \# \text{clauses}$

from each subgraph we must select exactly 1 vertex

- \exists satisfying assignment: for each clause, pick 1 satisfied literal, put its vertex to the indep. set \rightarrow got 1.s. of size k
- \exists indep. set of size k: each vertex selected in I.S. selects a variable which will be set to 0/1 to satisfy the corresponding clause, red edges guarantee that we won't set var to both 0 and 1 - remaining variables set arbitrarily \rightarrow get satisfying assignment
- the reduction runs in poly. time

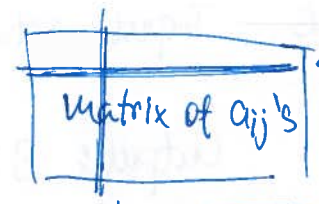
Example: INDEPSET \leq_P SAT ... given G, k , construct formula φ s.t. φ is satisfiable $\Leftrightarrow G$ has ind. set of size k

Variables: $x_1 \dots x_n$: vertex v_i selected to ind. set

a_{ij} for $1 \leq i \leq k, 1 \leq j \leq n$: vertex j is i -th in the ind. set \leftarrow order on vertices of the set

Clauses: $\forall \{v_i, v_j\} \in E(G): \neg x_i \vee \neg x_j$

$\forall i, j \ a_{ij} \Rightarrow x_j$ \leftarrow order describes the set (we allow unordered elements of set, which doesn't break anything)



$a_{ij} \Rightarrow \neg a_{ij}$ no number used multiple times

$a_{ij} \Rightarrow a_{ij}$ no vertex used twice or more

$a_{i1} \vee \dots \vee a_{in}$ each number used at least once

so we get CNF

implication $x \Rightarrow y$ is a clause $\neg x \vee y$

Exercises: ① $IMPSET \leq_m^P CLIQUE$

② 3-COLORING $\leq_m^P SAT$

Formalization of "search problems":

Def: Class of languages NP:

$$LENP \equiv \exists VEP \text{ (verifier)}$$

$$\forall \alpha \in \Sigma^* : \alpha \in L \Leftrightarrow (\exists \beta \in \Sigma^* : |\beta| \leq poly(|\alpha|) \ \& \ V(\alpha, \beta))$$

\exists certificate of polynomial size which is accepted by the verifier

👁️ $P \subseteq NP$... verifier does all the work & ignores β

👁️ resembles proofs in logic: true statements have a proof, which is easy to verify for false statements, no proof passes verification

Big question: Is $P=NP$?

⤴ 1MB price by Clay Mathematical Institute (waits for you lol)

Def: Language L is NP-hard $\equiv \forall K \in NP : K \leq_m^P L$

L is NP-complete \equiv furthermore, $L \in NP$

Lemma: Let $K \leq_m^P L$. Then:

① if $L \in NP$, then $K \in NP$ (just compose verifier with reduction)

② if K is NP-hard, then L is NP-hard. ($\forall M \in NP : M \leq K \leq L \Rightarrow M \leq L$)

③ if K is NP-complete & $L \in NP$, then L is NP-complete.

} makes it easy to prove NP-completeness once we have one NP-comp. problem

Lemma: If $LENP$ is NP-complete, then $P=NP$.

Proof: $P \subseteq NP$ is trivial, will prove $NP \subseteq P$:

Let $K \in NP$. Then $K \leq L$, which implies $K \in P$.

Thm (Cook-Levin): SAT is NP-complete.

↳ will be proven later

MORE REDUCTIONS

3D MATCHING

Input: sets B (boys), G (girls), C (cats)

$T \subseteq B \times G \times C$ (triples)

Output: $\exists T' \subseteq T$ s.t. each element of $B \cup G \cup C$ is contained in exactly 1 triple in T'

(generalizes bipartite matching, which is in P)

3-SAT: SAT, but all clauses contain at most 3 literals (generally: k-SAT)
3,3-SAT: Furthermore, every variable occurs in at most 3 clauses. (generally: k,l-SAT)
 [Extension: every literal occurs at most 2 times - i.e., the 3 occurs of a variable aren't all positive nor all negative.]

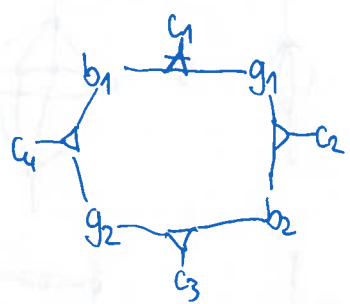
Reduction: $SAT \leq_P 3-SAT$
 $(l_1 \vee l_2 \vee \dots \vee l_k)$ → $(l_1 \vee l_2 \vee t) \wedge (l_3 \vee \dots \vee l_k \vee \neg t)$
 "long" clause with $k > 3$ literals → replace by 3 literals & $k-1$ literals
 ↑
 new variable

☞ New formula is satisfiable \Leftrightarrow the old one was.
 Iterate until all long clauses are broken.

Reduction: $3-SAT \leq_P 3,3-SAT$
 Replace variable x with $k > 3$ occurrences by new variables $x_1 \dots x_k$.
 Add clauses $(x_1 \Rightarrow x_2), (x_2 \Rightarrow x_3), \dots, (x_{k-1} \Rightarrow x_k), (x_k \Rightarrow x_1)$

☞ Preserves satisfiability. (this is $\neg x_2 \vee x_3$)
 ☞ Each new variable has at most 2 positive & at most 2 negative occurrences.
 Can apply the transform for $k=3$, too.

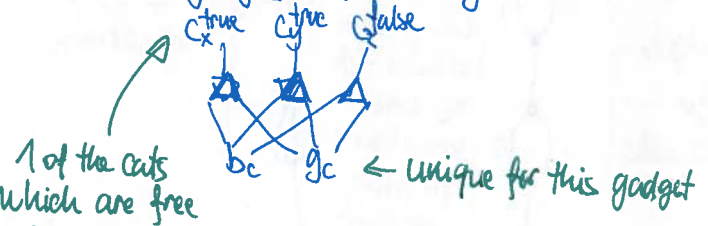
Reduction: $3,3-SAT \leq_P 3D-MATCHING$
Choice gadget (for each variable)



2 states: $\begin{matrix} \uparrow \\ \cdot \\ \downarrow \end{matrix}$ c_1, c_3 free ← logical 0
 and $\begin{matrix} \cdot \\ \uparrow \\ \downarrow \end{matrix}$ c_2, c_4 free ← logical 1

b_1, b_2, g_1, g_2 unique for this gadget
 $c_1 \dots c_4$ shared with clause gadgets

(also called consistency gadget)
Clause gadget for $x \vee y \vee \neg z$



1 of the cats which are free if x is true (that is c_2 or c_4) ← unique for this gadget

we have $(4 \cdot \#variables - \sum \text{of clause sizes})$ free cats \Rightarrow add this many pairs of "universal cat lovers" which have triples for every cat

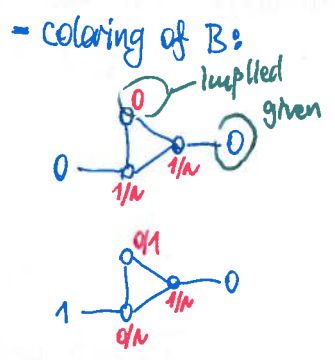
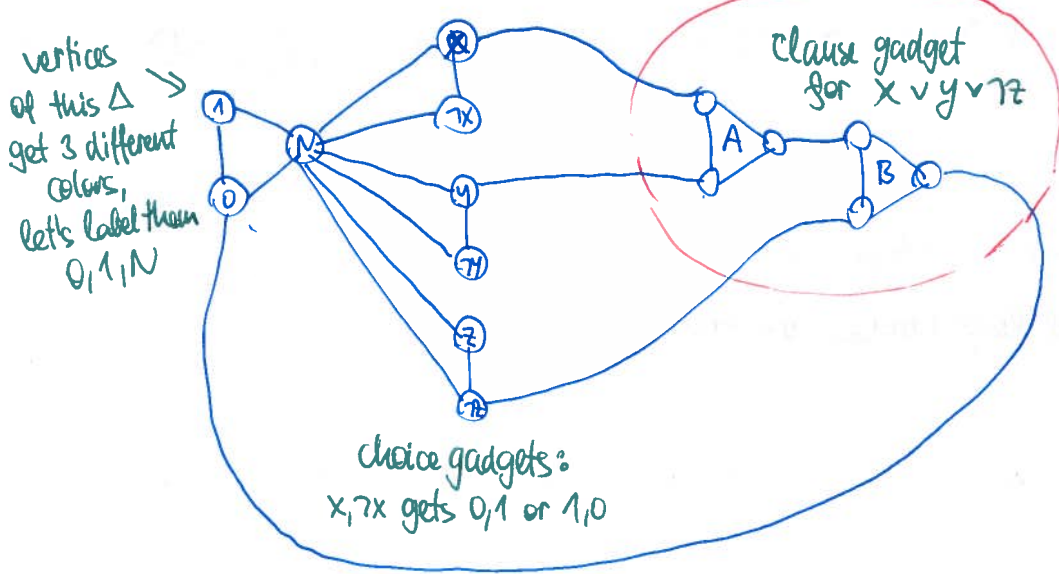
← each literal occurs at most 2 times in 3,3-SAT formulas, so we have enough cats for all literals

☞ \exists matching $\Leftrightarrow \exists$ satisfying assignment

Exercise 8 3D-MATCHING \leq_P ZOE (zero-one equations)

↳ & show that restriction of ZOE to equations with exactly 3 variables stays NP-complete.
 [This is sometimes called 1-in-3-SAT: exactly 1 literal must be true ... no negations are needed. There also exists a direct reduction from 3-SAT to this problem.]

Reduction: 3-SAT \leq_P 3-COLORING

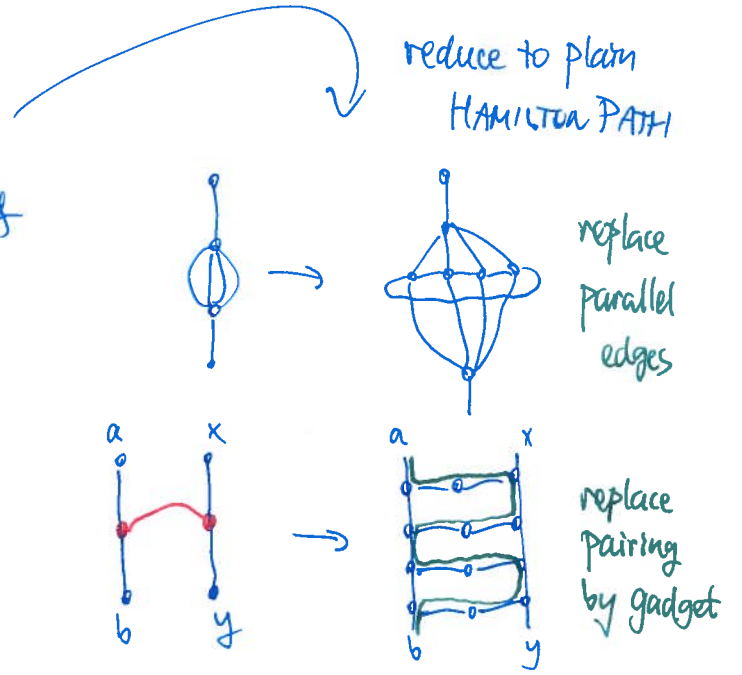
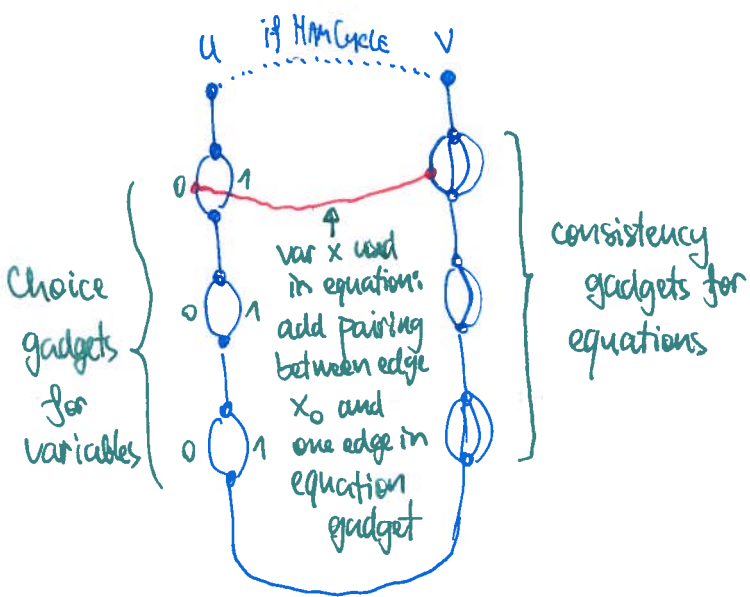


= coloring of B:
 - if inputs contain 1: use coloring of B which passes 0
 - if both are 0:

 need 1 here, so the 3rd literal must be 1

Reduction: ZOE \leq_P HAMILTON ~~CYCLE~~ ^{or PATH}

First consider problem HAMPATH* which allows:
 • parallel edges
 • pairing: $e \text{ --- } f \equiv$ must use exactly 1 edge of e, f



Exercises: • SUBSET SUM problem is NP-complete: Given a finite set $X \subseteq \mathbb{N}$, $s \in \mathbb{N}$, is there $X' \subseteq X$ s.t. $\sum_{a \in X'} a = s$? (hint: reduce from ZOE)

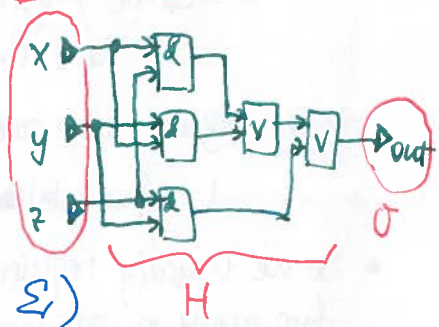
• 2 BANDITS: given finite $X \subseteq \mathbb{N}$, is there $X' \subseteq X$ s.t. $\sum X' = \sum(X \setminus X')$? Also NP-complete.

Brief detour: From formulas to Boolean circuits

Df: A Combinatorial Circuit consists of:

- a finite alphabet Σ
- finite sets I (input terminals) = $\{i_1, \dots, i_{|I|}\}$
 O (output terminals) = $\{o_1, \dots, o_{|O|}\}$
 H (gates) = $\{h_1, \dots, h_{|H|}\}$ } pairwise disjoint
- directed acyclic multigraph $(I \cup O \cup H, E)$
- arity $\cdot a: H \rightarrow \mathbb{N}$
- assignment of functions to gates $F: h \mapsto (f_h: \Sigma^{a(h)} \rightarrow \Sigma)$
- assignment of gate inputs to incoming edges $\alpha: (u,v) \in E \mapsto i \in \{1, \dots, a(v)\}$

Example: Majority of x, y, z



Where:

- $\forall i \in I \text{ deg}^{\text{in}}(i) = 0$
- $\forall o \in O \text{ deg}^{\text{in}}(o) = 1, \text{ deg}^{\text{out}}(o) = 0$
- $\forall h \in H \text{ deg}^{\text{in}}(h) = a(h) \ \& \ \forall i \in \{1, \dots, a(h)\} \exists! (x,h) \in E: \alpha((x,h)) = i$

Df: Boolean Circuit: Comb. circuit with $\Sigma = \{0,1\}$

Df: Computation of a circuit proceeds in steps.

- Step 0: input terminals and arity-0 gates (constants) have defined values.
- Step $i+1$: gates whose input is defined in step at most i produce output.
- As the graph is acyclic, gate outputs never change and every gate/terminal is defined within finite # steps.

⇒ the circuit computes a function from $\Sigma^{|I|}$ to $\Sigma^{|O|}$

Bounding arity: Since a single gate of high arity can compute anything in 1 step, we will bound arity by 2. (Actually, any fixed constant > 1 would work.)

Circuit complexity: Time \approx # layers (# steps of computation)
 Space \approx # gates

STW circuits are an interesting model of parallel computing

Boolean formulas \approx circuits with tree structure (except for inputs)

Lemmas: Every function $f: \{0,1\}^k \rightarrow \{0,1\}^l$ can be computed by a Boolean circuit consisting only of AND, OR and NOT gates. (OR can be replaced by $\overline{x \& \bar{y}}$)

in fact it's a formula in DNF

Proof: ① n-input AND/OR can be computed by a tree of 2-input ANDs/ORs.

② Function with multiple-bit output: replace by l single-bit functions.

③ Function whose truth table contains exactly one 1:
 e.g. $\neg x_1 \& \neg x_2 \& \neg x_3 \& x_4 \dots$ 1 at position 0001

④ Truth table with multiple 1s: OR functions for each 1.

⑤ Otherwise it's constant 0.

produces circuits of exponential size, but good enough for k, l constant

- Corollaries:
- ① can simulate arbitrary gates of fixed arity with $O(1)$ space/time overhead. (26)
 - ② can simulate arbitrary comb. circuit by a Boolean circuit (binary-encoded Σ^1)

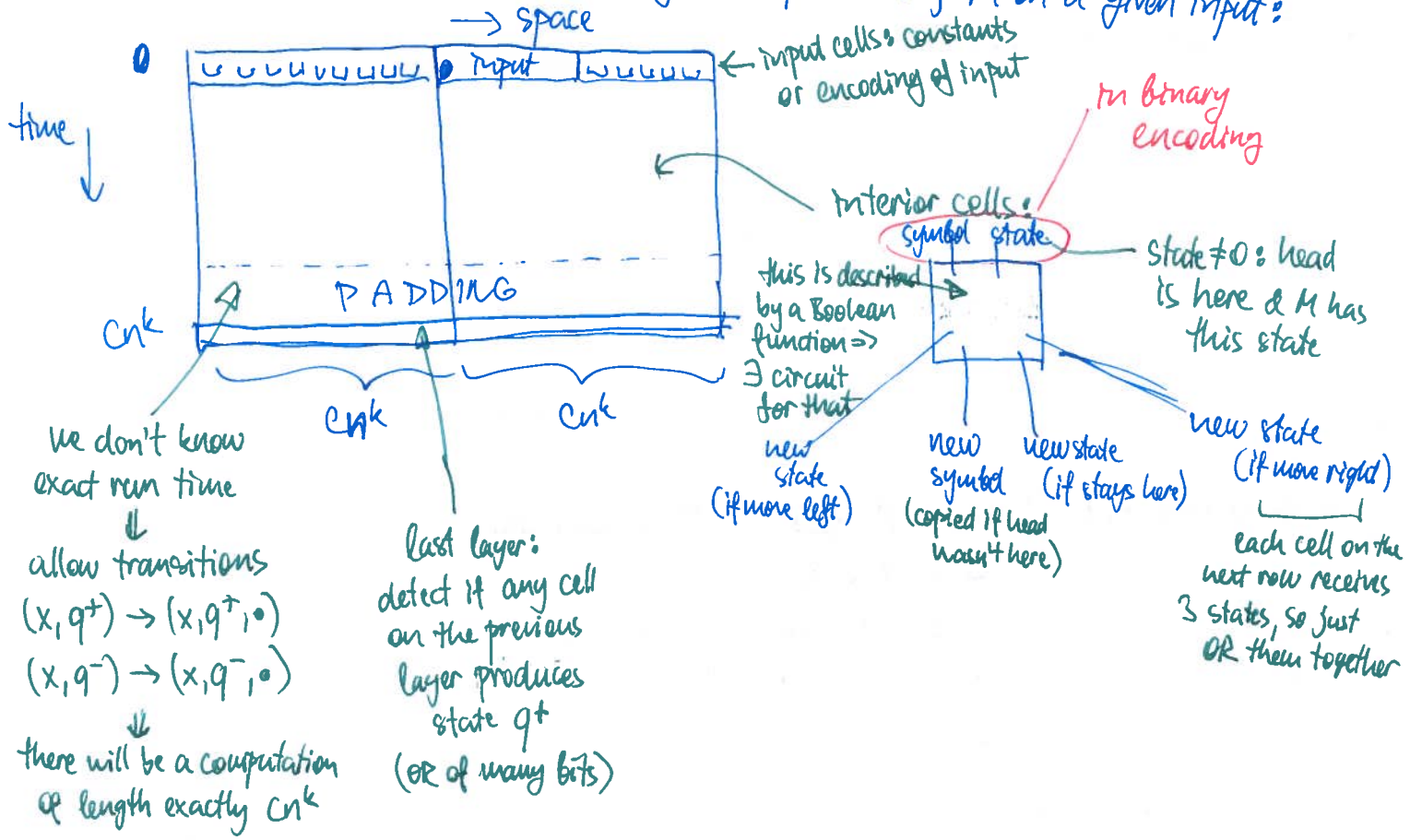
Problem: A circuit handles inputs of constant size only.
 ↳ a "program" is a family of circuits C_0, C_1, \dots
 where C_n solves the problem for inputs of size n .

But: If we allow arbitrary sequences, we can compute undecidable problems:
 $L = \{ \alpha \mid |\alpha| \text{ written in binary} \in L_u \}$

- So we usually require the family to be uniform: there is an algorithm which for every n produces C_n in time $\text{poly}(n)$.
- So languages decidable by uniform circuit families = P

Theorem: For every LEP there is $f \in PF$ s.t. for every n , $f(n)$ is an (encoding of) Boolean circuit with n inputs and 1 output which decides L for strings of length n .

Proof: Let M be a 1-tape TM deciding L in time at most $c \cdot n^k$ for some $c, k \in \mathbb{N}$. We will build a circuit producing a computation of M on a given input:
 ↳ with obvious meaning (computes char. function of L)

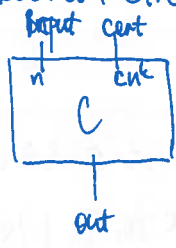


Def: CIRCUIT-SAT: given a Boolean circuit with 1 output, is there an input for which the output is true?
 ↳ Obviously, this is in NP.

Thm: CIRCUIT-SAT is NP-complete.

Proof: When reducing from LEMP to C-SAT: consider verifier $V \in P$ & upper bound cn^k for certificate size.

- Adapt verifier to accept certificates of size exactly cn^k (using reversible padding like 10^*)
- Find Boolean circuit for V on inputs of size $n + cn^k$:

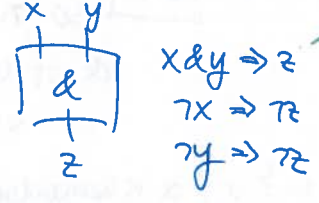
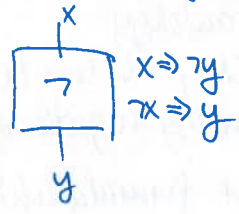


& fix input terminals to input α
 \downarrow
 SAT for $C_\alpha(\text{cert})$ computes $\alpha \in L$

done inside the reduction when receiving input α of size n

Lemma: CIRCUIT-SAT \leq_m^P SAT.

Proof: Assume WLOG that all gates are AND and NOT. Introduce new variables for gate outputs. Add consistency-checking clauses:

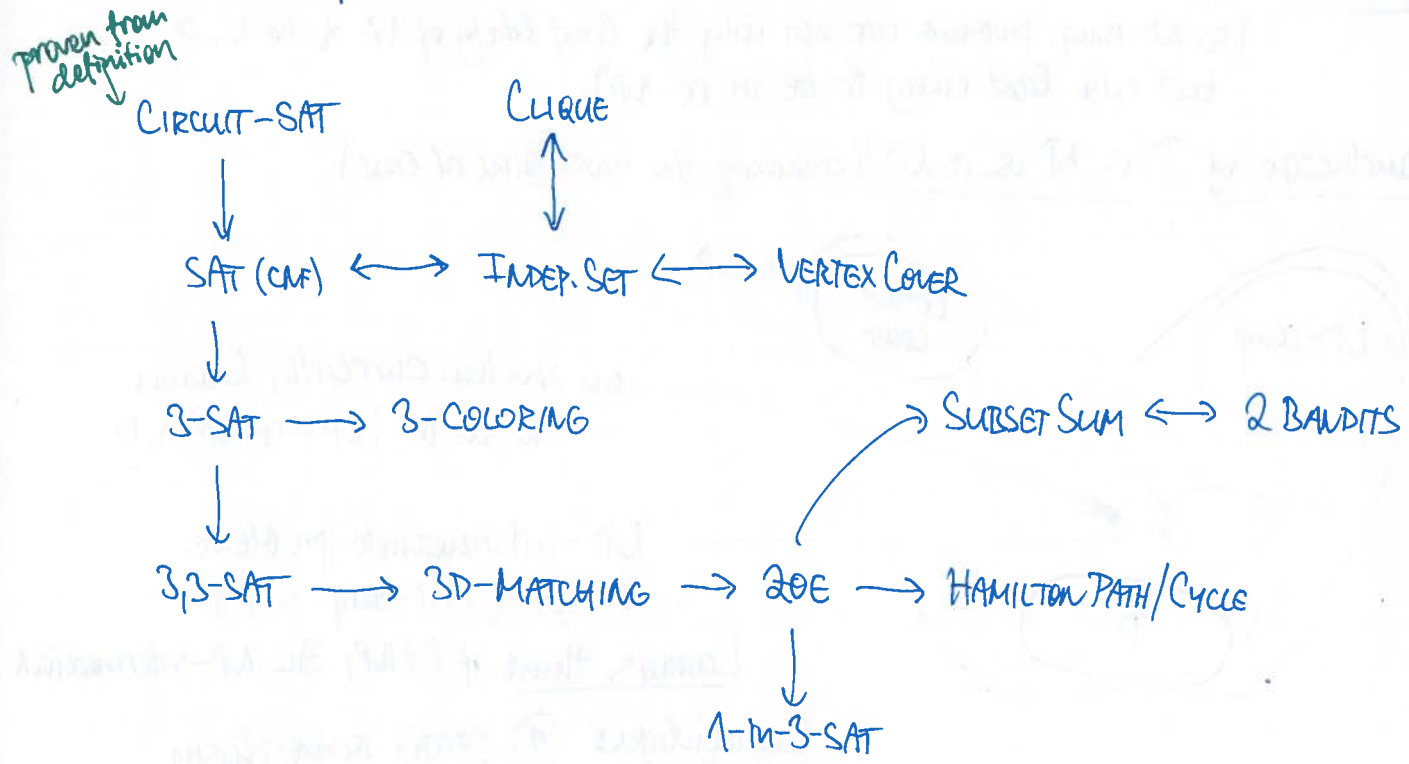


this is $\neg x \vee \neg y \vee z$ in CNF

this also means that SAT for general (non-CNF) formulas is reducible to SAT
 BTW we produced an instance of 3-SAT :)

Corollary: SAT is NP-complete. [This is Cook-Levin theorem!]

Map of NP-complete problems we encountered until now:



Further SAT variants: 2-SAT is in P

E_3, E_3 -SAT (clauses of size exactly 3, vars have exactly 3 occurrences)
 surprise: all instances satisfiable P

The class co-NP

Df: For a language $L \subseteq \{0,1\}^*$ we define its complement $\bar{L} := \{0,1\}^* \setminus L$

Df: For a class \mathcal{C} of languages: $\text{co-}\mathcal{C} := \{\bar{L} \mid L \in \mathcal{C}\} \rightarrow \text{co-P} = \text{P}$

Let's study co-NP...

- $P \subseteq NP \cap \text{co-NP}$... open if the inclusion is strict
- if $P = NP$, then $NP = \text{co-NP}$
- if $NP \neq \text{co-NP}$, then $P \neq NP$ (because $P = \text{co-P}$)
- as $K \leq_P L \Leftrightarrow \bar{K} \leq_P \bar{L}$, we have: L is NP-complete $\Leftrightarrow \bar{L}$ is co-NP-complete

certificate-based def.: $L \in \text{co-NP} \equiv \exists V \in P: (\alpha \in L \Leftrightarrow \forall \beta \in \{0,1\}, |\beta| \in \text{poly}(|\alpha|) \vee (\beta \notin L))$

so SAT is co-NP-complete

\uparrow this is not UNSAT (unsatisfiability), because for strings which do not encode a formula, we still have to answer 1 in SAT but 0 in UNSAT

\hookrightarrow but $\bar{\text{SAT}} \leq_P \text{UNSAT}$, so UNSAT is co-NP-comp.

$\hookrightarrow \exists \bar{x} \varphi(\bar{x}) \Leftrightarrow \forall x \neg \varphi(x)$

\hookrightarrow so $\neg \varphi$ is a tautology & if φ is in CNF, $\neg \varphi$ can be written in DNF by propagating negation

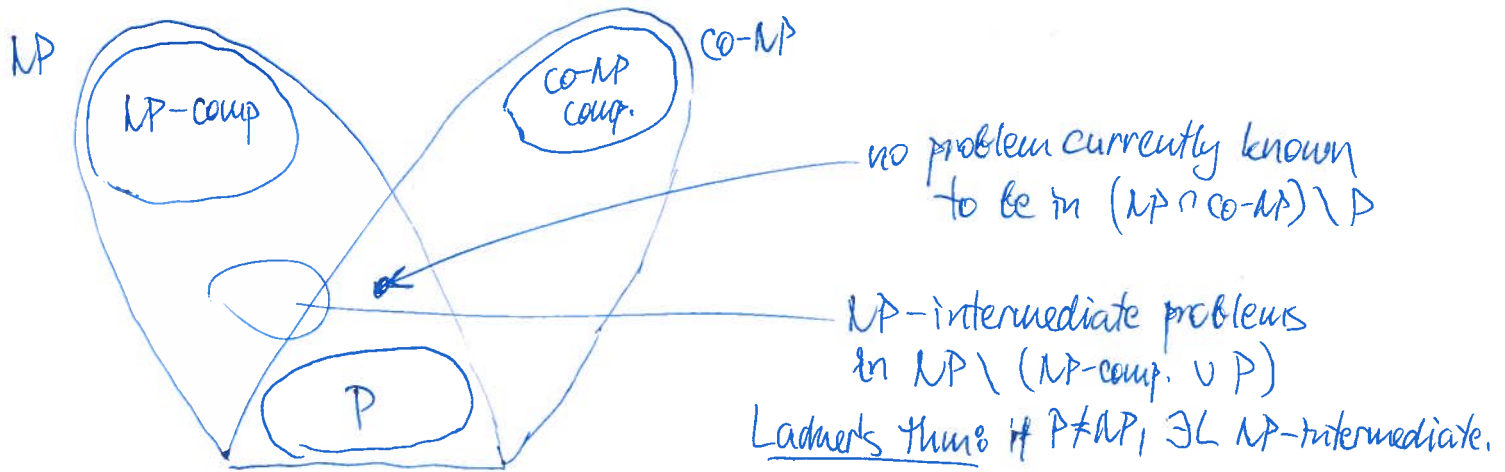
So: TAUTOLOGY := $\{\alpha \mid \alpha \text{ is (encoding of) DNF formula which is tautological}\}$ is also co-NP-complete (this is the most standard co-NP-c. problem)

Formally: TAUTOLOGY \leq_P UNSAT \leq_P SAT

Exercise: If $L \in \text{co-NP}$ is NP-complete, then $NP = \text{co-NP}$.

(so NP-comp. problems are not only the least likely of NP to be in P, but also least likely to be in co-NP).

Landscape of P vs NP vs. co-NP (assuming the most general case)



Candidates: ① graph isomorphism (known to be in $\text{DTIME}(n^{\log n})$)

② factoring (given x, a, b : does x have a factor $m \in [a, b]$?)

Non-deterministic TM (NTM)

extend $\delta: Q \times \Gamma^k \rightarrow \mathcal{P}(Q \times \Gamma^k \times \{\leftarrow, \rightarrow, \cdot\}^k)$... choice of instruction from the set

- successor relation is not a function \rightarrow multiple computations for a given output
- if $\delta(q,x) = \emptyset$, we assume rejection.
- input α accepted $\equiv \exists$ at least one accepting computation
- halting \equiv all computations halt
- time & space: maximum over all computations

\downarrow

\exists MLOG $|\delta(q,x)| \leq 2$

👁 enumeration works again ($NTM_\alpha = NTM$ with code α), we have an universal NTM d.c.

Exercise: k-tape \rightarrow 2-tape NTM with only constant-factor slowdown

Df: Non-deterministic complexity classes: $NTIME(f)$, $NTIMEF(f)$ for functions,

Theorems: $NP = NTIME(poly(n)) = \bigcup_{k \geq 0} NTIME(n^k)$

↳ all accepting computations must agree on result, \exists at least one accepting comp.

- Proof:
- \subseteq the NTM guesses the certificate using non-determinism & then it runs the verifier
 - \supseteq the certificate encodes the non-deterministic choices

Example: The following problem is NP-complete:

$$\{ \langle \alpha, \beta, t \rangle \mid NTM_\alpha \text{ accepts input } \beta \text{ within } t \text{ steps} \}$$

\rightarrow GAP: simulate $NTM_\alpha(\beta)$ using universal NTM with an "alarm clock" (reject after t steps)

reduction: calculate $t \in poly(n)$, $\alpha :=$ code of NTM solving source problem

Space Complexity

We want to count only "work space" of the TM.

3 types of tapes:

- input tape: read-only, head doesn't move more than 1 cell before/after input string
- k work tapes: read-write
- output tape: write-only, head cannot move left

space used by computation \equiv # visited cells on the work tapes (for NTM: max over computations)

👁 This doesn't change time complexity classes: we can copy input \rightarrow work \rightarrow output with constant slowdown

- 👁 We can encode information in position of head on input tape.
- this makes a difference if work space $\in o(\log n)$
 - otherwise we can keep track of the head position in binary

Space classes

- $DSPACE(f)$ } decision problems
- $NSPACE(f)$ }
- & $DSPACEF(f)$ } functions
- $NSPACEF(f)$ }

$PSPACE = DSPACE(poly(n))$
 $NPSPACE = NSPACE(poly(n))$

We want f to be:

- 1 non-decreasing
- 2 space-constructible
 $\& f(n)$ can be computed from 1^n , result in binary, in space $O(f(n))$
- 3 usually $f(n) \geq \log n$

proper space-complexity function

Basic Inclusions: $DTIME(f) \subseteq NTIME(f) \subseteq DSPACE(f) \subseteq NSPACE(f)$

↑ can try all certificates in space $O(f)$

So: $P \subseteq NP \subseteq PSPACE \subseteq NPSPACE$

Thm: $DSPACE(f) \subseteq DTIME(2^{O(f)})$ for every $f \geq \log n$.

Proof: First, let's bound # reachable configurations: $|Q| \cdot (n+2) \cdot (|M|+1)^{f(n)} \cdot f(n)^k$
↑ state ↑ pos. of head on input tape ↑ contents of work tapes, extra character for "end of tape" ↑ pos. of heads on work tapes
... this is $O(2^{O(f(n))})$

If a configuration repeats, the whole computation loops. (this requires deterministic TM)
⇒ add a binary counter of $O(f(n))$ bits, use it as alarm clock (increment in every step of the original TM). Alarm expires → reject.

in space-bounded computation with space $\geq \log n$, we can always make sure that the machine halts.

Corollary: $NPSPACE \subseteq EXPTIME := DTIME(2^{O(n)})$.

We want to prove the same for $NSPACE(f)$, but * makes it more complicated.

Reachability method

within space $f(n)$

Df: Configuration graph of a given NTM on a given input is a directed graph with:
V := set of configurations (for input tape, consider only head position)
↑ limited by available space

E := successor relation

start ∈ V ... initial config

accept ∈ V ... modify the TM to clean up before accepting
← clear working tapes } unique accepting config
← rewind input tape

⊙ $|V| \in O(2^{O(f)})$, $|E| \in O(|V|)$, graph can be generated in $O(\text{poly}(|V|))$ time & $O(f)$ space.

⊙ Machine accepts ⇔ graph contains a (directed) path from start to accept.

Thm: $NSPACE(f) \subseteq DTIME(2^{O(f)})$ for every $f \geq \log n$. [therefore $NPSPACE \subseteq EXPTIME$]

Proof: Construct the reachability graph & run BFS on it.
↑ time $O(\text{poly}(|V|, |E|))$ ↑ also time $O(\text{poly}(|V|, |E|))$
↑ which is $O(2^{O(f)})$

Generally: Time-/space-efficient algorithms for REACH translate to inclusions of complexity classes.
↑ $\{G, s, t\} \mid \exists \text{ path from } s \text{ to } t \text{ in } G$

Thm (Savitch's): $NSPACE(f) \subseteq DSPACE(f^2)$ for every $f \geq \log n$.

Corollary: $NPSPACE \subseteq PSPACE$, so $NPSPACE = PSPACE$.
→ will be proven soon...

Lemma: REACH $\in O(\log^2 n)$

Proof: Use "middle-first search".

Recursive function $D_k(x,y)$ computing " \exists walk from x to y with at most 2^k edges".

We have: $D_0(x,y) = (x=y) \vee ((x,y) \in E)$

$D_k(x,y) = \bigvee_{z \in V} (D_{k-1}(x,z) \wedge D_{k-1}(z,y))$... FOR loop & recursion

Every level of recursion requires $O(\log n)$ space for local variables, $\log n$ levels suffice to find a path.

Now, we want to combine generator of config graph with this algorithm, but we don't have space to store the graph.

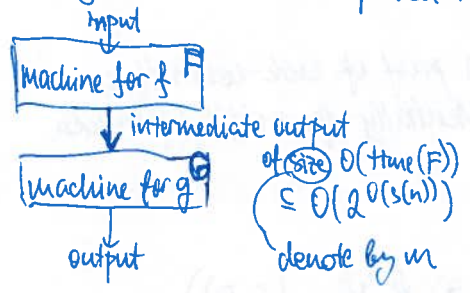
Lemma: If f can be computed in space $s(n)$ and $g \equiv \Pi \equiv t(m)$,

$\left. \begin{matrix} s(t) \geq \log \\ t(2^{O(s(n))}) \end{matrix} \right\}$

also applies to composition of a language with a function (that is, a reduction) $O(\log m)$ space

Then $g(f(\dots))$ can be computed in space $O(s(n) + t(2^{O(s(n))}))$

Proof:



Start G & keep track of (position) of head on G 's input tape. Whenever G moves its input head (\neq at the start of computation), re-run F to get the corresponding symbol of its output.

\hookrightarrow modify F : reset work tapes on startup, reset input head $\equiv \Pi$, keep track of output head position $] O(\log m)$ space

write to output tape: compare with G 's input head pos $\begin{cases} = & \text{remember char in state} \\ \neq & \text{discard written char (won't be read by } F) \end{cases}$

Total space:

- $s(n)$ for F
- $O(\log m)$ for head positions ... this is $O(t(m))$
- $t(m)$ for G

Corollary: Savitch's thm.

\hookrightarrow If $L \in \text{NSPACE}(f)$: graph generation requires $O(f)$ space, reachability needs $O((2^{O(f)})^2) = O(2^{O(f)})$

combined by the lemma to $O(2^{O(f)})$

Remark: REACH $\in O(\log n)$ would imply $\text{NSPACE}(f) = \text{DSPACE}(f)$... but this is long open.

It's known that undirected UREACH $\in O(\log n)$ [Reingold 2004, non-trivial]

\hookrightarrow this implies only $\text{SSPACE}(f) = \text{DSPACE}(f)$

\uparrow symmetric non-determinism (successor relation symmetric)

So we have: $\text{DTIME}(f) \subseteq \text{NTIME}(f) \subseteq \text{DSPACE}(f) \subseteq \text{NSPACE}(f) \subseteq \text{DTIME}(2^{O(f)}) \subseteq \text{DSPACE}(f^2)$

and: $\text{NSPACE}(\log n) \subseteq P \subseteq NP \subseteq \text{NPSPACE} = \text{NPSPACE} \subseteq \text{EXPTIME}$

\uparrow this is also known as NL

\uparrow also = co-NPSPACE as PSPACE is closed under complements

More about PSPACE

Thus QBF is PSPACE-complete.

with respect to \leq_m

the language of all true quantified Boolean formulas (all variables bound by quantifiers)

Proof: 1) QBF \in PSPACE by the following recursive algorithm:

- $QBF(\forall x \varphi(x)) = QBF(\varphi(0)) \& QBF(\varphi(1))$
- $QBF(\exists x \varphi(x)) = QBF(\varphi(0)) \vee QBF(\varphi(1))$
- $QBF(\varphi \vee \psi) = QBF(\varphi) \vee QBF(\psi)$
- $QBF(\varphi \& \psi) = QBF(\varphi) \& QBF(\psi)$
- $QBF(\neg \varphi) = \neg QBF(\varphi)$

$O(n)$ levels of recursion, $O(n)$ space per level. } $O(n^2)$ space

2) QBF is PSPACE-hard: consider $L \in$ PSPACE, TM M deciding L and its config. graph G .

- vectors of variables \bar{x} encoding vertices ... $O(\text{poly}(n))$ bits [log of $|G|$]
- formula $\varphi(\bar{x}, \bar{y}) \equiv (\bar{x} = \bar{y}) \vee (x, y) \in E(G)$
- can construct poly-sized circuit as in proof of Cook-Levin thm. & then reduce the circuit to an existentially-quantified formula as in Circuit-SAT \leq_m SAT.

• mimic proof of Savitch's thm.

Failed attempt: $\varphi_k(\bar{x}, \bar{y}) \equiv \exists \bar{z} (\varphi_{k-1}(\bar{x}, \bar{z}) \& \varphi_{k-1}(\bar{z}, \bar{y}))$

Double recursion \Rightarrow formula size grows exponentially!

Better: $\varphi_k(\bar{x}, \bar{y}) \equiv \exists \bar{z} \forall \bar{a} \forall \bar{b} ((\bar{a} = \bar{x} \& \bar{b} = \bar{z}) \vee (\bar{a} = \bar{z} \& \bar{b} = \bar{y})) \Rightarrow \varphi_{k-1}(\bar{a}, \bar{b})$

• G has size $O(2^{\text{poly}(n)}) \Rightarrow \log(\text{path len}) \in \text{poly}(n) \Rightarrow$ recursion has $\text{poly}(n)$ levels, formula size grows to $\text{poly}(n)$.

Intuition: PSPACE is the class of strategies for 2-player games with perfect information:

(\exists player 1 move) (\forall player 2's response) (\exists player 1's counter-response) ($\forall \dots$) ...

Examples

these are known to be PSPACE-complete

- graph coloring game: undirected graph, finite set of k colors each player colors an uncolored vertex, ~~every~~ no edge must have both ends of the same color
- graph path game: building path edge by edge, ~~each~~ ^{first} player has his target, ~~and player~~ vertices must not repeat
- variants of Go, checkers &c. (generalized to $L \times N$ boards)
- Sokoban (1-player, but enough internal state, which restricts future moves, but can be modified)

Alternating Turing Machine (ATM)

- 3 kinds of states:
- deterministic: config is accepting \Leftrightarrow next config is accepting
 - existential: ^{config is accepting} ~~accepts~~ $\Leftrightarrow \exists$ non-det. choice ~~leading to the next~~ leading to accepting config.
 - universal: is accepting $\Leftrightarrow \forall$ non-det. choice leads to accepting config.
- We will require all computations to halt.

ATM \rightarrow classes $ATIME(f), AP \dots$ [we won't define space-bounded classes as we require all computations to halt & alarm clocks don't help]

Theorem: $AP = PSPACE$.

Proof: \supseteq is easy: $QBF \in AP$ since we can execute quantifiers using corresponding state types. So if $L \leq_m QBF$, we can first compute the reduction and then solve QBF .

\subseteq : simulate the ATM recursively as in $QBF \in PSPACE$.

- configurations take $O(\text{poly}(n))$ space - all computations are poly-time, so they are poly-space, too.
- recursion depth is bounded by time of the ATM.

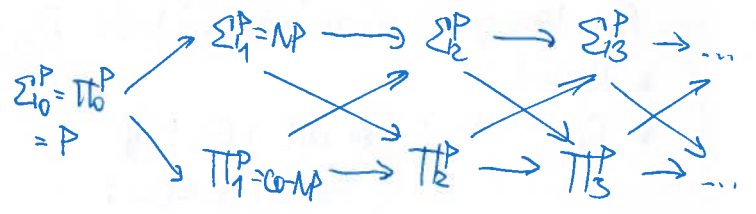
Polynomial Hierarchy

Consider following restrictions of QBF :

- Σ_k -formulas: $(\exists x_1 \dots \exists x_k)(\forall \dots)(\exists \dots) \dots \psi(\dots)$
 - k groups of quantifiers, starting with \exists
 - quantifier-free formula
 - Π_k -formulas: similar, but starting with \forall
 - Σ_k -SAT := $\{ \langle \varphi \rangle \mid \varphi \text{ is a true } \Sigma_k\text{-formula} \}$... similarly Π_k -SAT.
 - Σ_1 -SAT is SAT (for general formulas, not only CNF), Π_1 -SAT is TAUT.
- negation of a Σ_k -formula can be written as a Π_k -formula & vice versa

We can use this to define new classes which generalize NP:

- $\Sigma_k^P := \{ L \mid L \leq_m^P \Sigma_k\text{-SAT} \}$... $\Sigma_1^P = NP, \Sigma_0^P = P$
- $\Pi_k^P := \{ L \mid L \leq_m^P \Pi_k\text{-SAT} \}$... $\Pi_1^P = co-NP, \Pi_0^P = P, \Pi_k^P = co-\Sigma_k^P$
- generally, the following inclusions hold:



we defined classes using a problem complete for them

This is akin to the arithmetical hierarchy, but the inclusions are not known to be strict.

$PH := \bigcup_k \Sigma_k^P = \bigcup_k \Pi_k^P$

since every Σ_k/Π_k -SAT reduces trivially to QBF , we have $PH \subseteq PSPACE$ (not known to be strict)

Example: "given Bool. formula φ , find the shortest ψ s.t. $\forall \bar{x} \varphi(\bar{x}) \Leftrightarrow \psi(\bar{x})" \in \Sigma_2^P$ (in fact, it's Σ_2^P -complete)

Remark: Definition using oracle machines also works:

$\Sigma_{k+1}^P := NP[\Sigma_k^P] = NP[\Pi_k^P]$

$\Pi_{k+1}^P := co-NP[\Sigma_k^P] = co-NP[\Pi_k^P]$

we can add $\Delta_{k+1}^P := P[\Sigma_k^P] = P[\Pi_k^P] \dots$



this leads to the same hierarchy

Remark: We can also define Σ_k^P & Π_k^P using alternative TMs:

- $\Sigma_k\text{-TIME}(f) := \{ L \mid L \text{ can be decided by an ATM running in time } \leq f(\text{input}) \text{ which performs at most } k \text{ quantifier changes, starting with } \exists \}$
- $\Sigma_k^P = \Sigma_k\text{-TIME}(poly(n))$

Collapse of PH

- $P = NP \Leftrightarrow P = PH$
- $NP = co-NP \Leftrightarrow NP = PH$
- If $\Sigma_j^P = \Sigma_{j+1}^P$, then $\Sigma_k^P = \Sigma_j^P$ for all $k > j$] We say that PH collapsed to the j-th level
- ... so $PH = \Sigma_1^P$
- If $\Sigma_j^P = \Pi_j^P$, then $\Sigma_{j+1}^P = \Sigma_j^P$ (we can reduce $\exists \bar{x} \forall \bar{y} \psi(\dots)$ to $\exists \bar{x} \exists \bar{y} \psi(\dots)$)

this is weaker than $P=NP$, but still open

Note: If graph isomorphism is in P, then $PH = \Sigma_1^P$. [proof non-trivial]

SPACE CO-CLASSES

Unlike non-deterministic time classes, non-det space classes are known to be closed under complement.

Theorem (Immerman-Szelepcsényi): $NSPACE(s(n)) = co-NSPACE(s(n))$

for all space-constructible functions $s(n) \geq \log n$.

Proof: We design a non-deterministic algorithm for non-reachability in config. graphs. More generally, we'll calculate R_i s = #vertices reachable from source by walk of len $\leq i$.

Then modify graph by adding edges from target to all vertices, so (not reachable from src) $\Leftrightarrow R_n = n$ for $n = \#vertices$.

$V_i :=$ set of those vertices

- $R_0 = 1$
- $R_{i-1} \rightarrow R_i$: For all $v \in V$:
 For all $w \in V_{i-1}$:
 if $(w,v) \in E$ or $v=w$: $R_i \leftarrow R_i + 1$

• Enumeration of V_i s

$t \leftarrow 0$
 For all $u \in V$:

If guess that $u \in V_i$:

If \nexists walk $src \rightarrow u$ of length $\leq i$: REJECT
 $t \leftarrow t + 1$

If $t \neq R_{i-1}$: REJECT

if we don't guess correctly, either the path doesn't exist or $t < R_{i-1}$ at the end

guess the path using non-determinism & check that it's valid

Space needed: • $O(1)$ variables for vertices & counters } $O(\log |V|)$ space,
 • R_i and R_{i-1} } where $|V| = 2^{O(s(n))}$,
 so this is in $NSPACE(s(n))$.

INSIDE P

☀️ this is transitive

We need to use log-space reductions \leq_{log}^{log} (when using \leq_{in}^P , all problems in P except \emptyset and Σ_0^P are equivalent)

Important classes: $L := DSPACE(\log n)$, $NL := NSPACE(\log n)$

We know: $L \subseteq NL = co-NL \subseteq P$
↑ trivial ↑ Imm.-Sz. ↑ reachability & $2^{c \log n} = n^c$ } inclusions not known to be strict

Theorem: CIRCUIT-EVAL is P-complete wrt. \leq_{log}^{log} .
☀️ given Boolean circuit & input, is the output true?

Proof: Verify that the circuit construction we used when proving Cook's thm can be carried out in log. space. } open: CIRCUIT-EVAL $\in NL$

Theorem: REACH is NL-complete wrt. \leq_{log}^{log} .

Proof: Configuration graph of a NTM can be constructed in log space. } open: REACH $\in L$

2-SAT (CNF formulas, all clauses have ≤ 2 literals)

☀️ $(\alpha \vee \beta)$ is an implication $\neg \alpha \Rightarrow \beta$, which is also $\neg \beta \Rightarrow \alpha$
☀️ exactly 2 if we replace (x) by (xvx)

For a 2-CNF formula φ , construct its implication graph: vertices = variables & their negations
edges = implications (clause produces two)

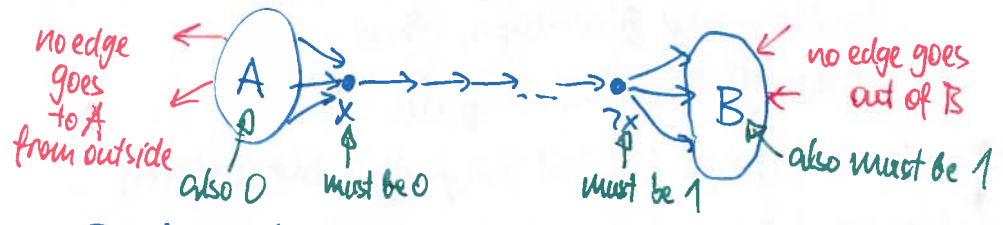
Lemma: φ is unsatisfiable $\Leftrightarrow \exists$ var. v s.t. G_φ contains both a path $v \rightarrow \neg v$ and a path $\neg v \rightarrow v$ } "contradictory cycle"

Proof: First observe: if literal α is set to 1, all literals reachable from α must be also 1.
... if v set to 0, all literals from which v is reachable must be also 0.

Hence \Leftarrow is true.

\Rightarrow : prove contra-positive: If there \exists a contradictory cycle, we construct a satisfying assignment.

① If there exists a path $x \rightarrow \neg x$:



Also, A is the mirror image of B:
- literals negated
- edge directions flipped

If $A \cap B \neq \emptyset$: \exists contradictory cycle.

Otherwise: remove $A, B, x, \neg x$ & continue (because of red note, the removed part cannot affect SAT-ability of the rest)

② \exists path $\neg x \rightarrow x$: symmetrically.

③ no such paths exist: add edge $x \rightarrow \neg x$ for some remaining variable x and continue (this couldn't have created a new contradictory cycle) } effectively setting $x=0$

Corollary: 2-SAT $\in P$ (In fact, there is an $O(n)$ -time alg. on the RAM)

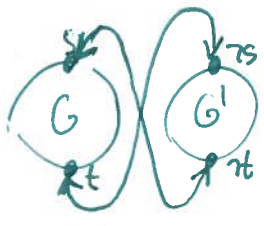
Thm: 2-SAT is NL-complete.

Proof: REACH \in NL, which is also co-NL, so $\overline{\text{REACH}} \in \text{NL}$.

We can decide 2-SAT using a subroutine for $\overline{\text{REACH}}$, so 2-SAT \in NL.

NL-hardness: REACH is NL-complete, NL=co-NL, so $\overline{\text{REACH}}$ is also NL-complete.

Let's reduce $\overline{\text{REACH}}$ to 2-SAT in log space:



- given graph G and vertices s,t (if s=t, REJECT) ↖ by producing a constant non-SATable formula
- build a formula with implication graph G, containing only positive literals (a disjoint copy of G gets created with the mirror image...)
- add implications $t \Rightarrow \neg s, \neg t \Rightarrow s$ (this creates $s \Rightarrow \neg t, \neg s \Rightarrow t$)
- resulting formula is SATable $\Leftrightarrow \exists$ path $s \rightarrow t$ (no other contradictory cycle is possible)

HIERARCHY THEOREMS

Goal: Show that some classes are different ∞

- Tools
- time/space-constructibility of functions
 - enumeration of machines M_x (we can also use integer codes instead of strings)
 - Universal Turing Machine (UTM): given $\langle x, \beta \rangle$, simulates M_x on input β .

- complexity: if M_x runs in time T and space S, on input β ,
 UTM(x, β) runs in:

- space $\in O(S)$ constants dependent on α (e.g., size of work alphabet)
- time $\in O(T^2)$ or $O(T \log T)$

- can extend the UTM to count space/time used

↙ by the simulated machine
 ↘ by the UTM itself
 & stop simulation if limit exceeded

↑ because of reduction
 k tapes \rightarrow 1 tape

↑ can use a better reduction
 $k \rightarrow 2$ tapes
 (we haven't proven that)

Theorem (space hierarchy): If f, g are non-decreasing space-constructible functions, $f \in o(g)$ and $f(n) \geq \log n$, then $DSPACE(f(n)) \subsetneq DSPACE(g(n))$.

Proof: \subseteq trivial, will construct a language $L \in DSPACE(g(n)) \setminus DSPACE(f(n))$.

Define machine M : Given input β :

↓
 then $L := L(M)$

1. Check that β has the form $\alpha 10^l$ for some α, l .
2. Write $g(|\beta|)$ 1s on a work tape X .
3. Simulate M_x on input β using an UTM.
 Stop if more than $g(|\beta|)$ cells are used by the UTM (assume M_x rejected then)
4. If M_x accepted, reject.
 If rejected, accept.

We check that M runs in space $O(g(n))$, so $L \in DSPACE(g(n))$.

Let's show that $L \notin DSPACE(f(n))$... if it were true, there $\exists M_\alpha$ deciding L in space $f'(n) \in O(f(n))$.

So the UTM can simulate M_α in space $c \cdot f'(n)$ for some c (depending on α).

\uparrow this is in $o(g(n))$, so $c \cdot f'(n) < g(n)$ for n large enough

Construct input $\beta := \alpha 10^l$ for l large enough.

Then the UTM fits in the time bound $g(|\beta|) \Rightarrow$ on this input, M_α doesn't agree with M \downarrow

Notes The trick with padding α by 10^l is actually not necessary, because for every machine, there are infinitely many equivalent codes \rightarrow just pick code α large enough.

Corollaries $DSPACE(n) \neq DSPACE(n^2) \neq DSPACE(n^3) \neq \dots$, so $PSPACE \neq DSPACE(n^k)$ for every k .
 $DSPACE(n) \neq DSPACE(n \log \log n) \neq DSPACE(n \log n) \neq DSPACE(n^2)$
 $NL \subseteq DSPACE(\log^2 n) \neq DSPACE(n) \neq PSPACE$ } so $NL \neq PSPACE$ and $QBF \notin NL$
 \uparrow Savitch's thm.
 $PSPACE \subseteq DSPACE(2^n) \neq DSPACE(2^{n^2}) \subseteq EXSPACE$ } so $PSPACE \neq EXSPACE$

Theorem (time hierarchy): If f, g are time-constructible non-decreasing functions such that $f \cdot \log f \in o(g)$, then $DTIME(f(n)) \neq DTIME(g(n))$.

Proof: Almost identical, modify step 3 to stop the UTM after $g(n)$ steps.

If M_α decides M in time $f' \in O(f)$, then UTM simulates M_α in time at most $f'(n) \log f'(n) \in o(g(n))$

So for large enough equivalent code α , UTM completes simulation of $M_\alpha(\alpha)$ in time $g(|\alpha|)$.

Therefore M_α disagrees with M on input α \downarrow

Corollaries $DTIME(n) \neq DTIME(n^2) \neq \dots$ (but we cannot separate $DTIME(n \log n)$ from $DTIME(n)$ this way)
 $DTIME(n^k) \neq DTIME(n \log n) \neq EXP \dots$ so $P \neq EXP$.
 $P \neq DTIME(n^k)$ for every k .

So we have: $L \subseteq NL \subseteq P \subseteq NP \subseteq PSPACE = NPSPACE \subseteq EXP \subseteq NEXP \subseteq EXSPACE$.

Note: What about non-deterministic classes?

We have $NTIME(f(n)) \neq NTIME(g(n))$
and $NSPACE(f(n)) \neq NSPACE(g(n))$ } whenever $f \in o(g)$

- non-deterministic reduction of tapes can be done with constant overhead in both time & space
- we have non-deterministic UTM
- but how do we negate its output ???
 - for space-bounded classes, use Immerman-Szelepcsenyi thm.
 - for time-bounded, a more involved proof is needed (not covered here)

RELATIVE CLASSES

We can define complexity classes for machines with an oracle.

For example $P[A]$ a.k.a. P^A and $NP[A]$ a.k.a. NP^A . ← called "relative" classes wrt. A
Many proofs apply to relativized statements of theorems, too.

But P vs. NP cannot be relativized: ← this limits proof techniques which could separate P from NP (e.g., diagonalization as in proofs in hierarchy that's doesn't work)

Theorem: There exist languages A, B s.t. $P[A] = NP[A]$, but $P[B] \neq NP[B]$.

Proof: (A) Let $A = QBF$. Then $P[A] = PSPACE[A] = PSPACE$
 $NP[A] \subseteq PSPACE[A] = PSPACE$.

(B) For every language B , define $U_B := \{1^n \mid \exists \beta \in B \text{ with } |\beta| = n\}$. ← "shadow cast by the language B "

We have $U_B \in NP[B]$: just guess β and check it's in B .

Construct B s.t. $U_B \notin P[B]$: in step i , we make sure that $M_i[B]$ doesn't decide U_B within $2^n/10$ steps for inputs of size n . To achieve that, we put finitely many strings inside or forever outside B we "decide their fate"

Step i : Choose $n >$ lengths of all strings whose fate we already decided.

Run $M_i[B]$ on 1^n for $2^n/10$ steps.

- when it queries B for a string β :
 - if fate of β was already decided, answer consistently
 - if not, put β outside B and answer NO
- if it accepted 1^n , arrange $1^n \notin U_B$: so far, no string of length n is in B , put the remaining ones outside B
- if it rejected 1^n , add one string of length n to B (so $1^n \in U_B$)
 - ↳ so far, we met at most $2^n/10$ such strings, so some undecided strings must remain.

Now if some machine $M[B]$ decides U_B in time $P^{poly}(n)$,

we have $f(n) < 2^n/10$ for n large enough.

For large enough i s.t. M_i is equivalent to M , n is also large enough
⇒ M disagrees with U_B on input 1^n .

So $U_B \notin P[B]$.

CIRCUIT COMPLEXITY

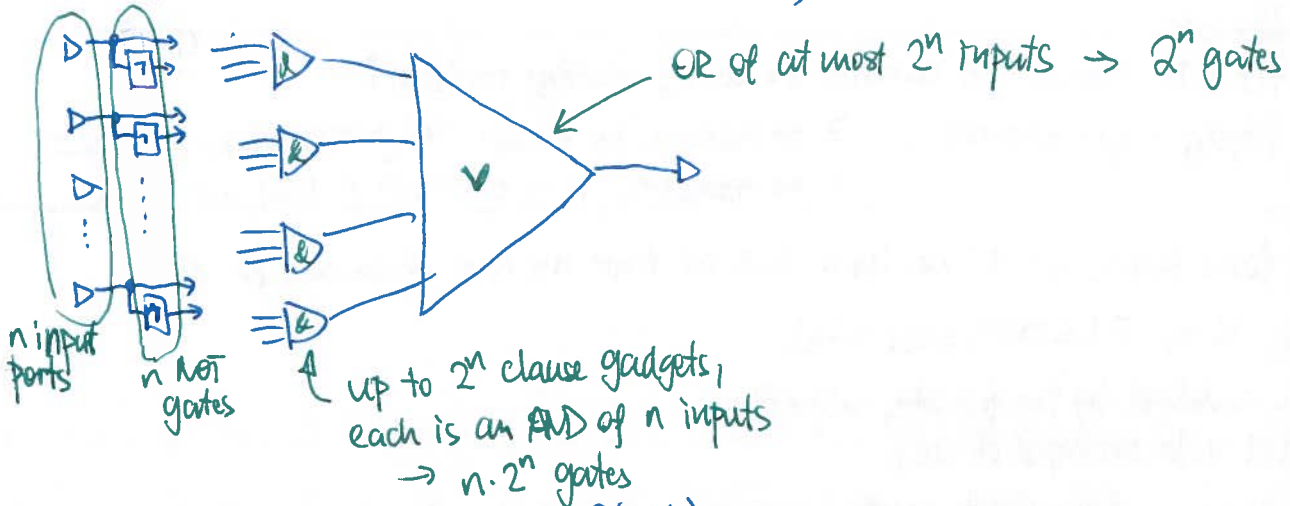
Back to (non-uniform) Boolean circuits.

We will use only AND, OR, NOT gates (other gates can be simulated with constant overhead).

Circuit size = #gates + #ports (input & output) ← size $\leq S$ is equivalent with size = S as we can pad circuits with redundant gates

Thm: For every function $f: \{0,1\}^n \rightarrow \{0,1\}$ there exists a circuit of size $\leq 10n \cdot 2^n$ computing f .

Proof: Use DNF formula for f (see lecture on Cook-Levin thm.)



Note: There is a better construction with $O(2^n/n)$ gates. ← surprisingly, this is asymptotically optimal

Exercise: Achieve $O(2^n)$ gates.

Thm: For all sufficiently large n , there is $f: \{0,1\}^n \rightarrow \{0,1\}$ which is computed by no circuit of size at most $2^n/10n$.

Proof: There are 2^{2^n} functions from $\{0,1\}^n$ to $\{0,1\}$.

Let's count circuits of a given size s :

$$\# \text{circuits} \leq 3^s \cdot s^{2s} = 2^{s \cdot \log_2 3} \cdot 2^{2s \log_2 s} \leq 2^{3s \log_2 s}$$

↑ except for ports, each gate can be AND, OR, NOT
 ↑ #interconnections: each gate has at most 2 inputs, which are connected to a port or output of another gate

Now for $s = 2^n/10n$:

$$\# \text{circuits} \leq 2^{\frac{3 \cdot 2^n}{10n} \cdot n} < 2^{2^n} \text{ for } n \text{ large enough.} \leftarrow \text{in fact, the majority of functions has no small circuits}$$

Notes: All problems in P have polynomially large circuits.

Hypothesis (Kolmogorov): $O(n)$ is enough.

Surprisingly, the best lower bound so far is $5n$.

Idea: If we found LEMP with super-polynomial lower bound for circuit size, then $P \neq NP$. But so far, we failed completely...

Def: For $S: \mathbb{N} \rightarrow \mathbb{N}$ we define $SIZE(S(n))$ as the class of languages, which are computable by a (non-uniform) family $\{C_n\}_{n=0}^{\infty}$ of circuits s.t. size of $C_n \leq S(n)$. beware, no σ here

We know: $P \subseteq \bigcup_k SIZE(n^k + k)$ ← this is to overcome finitely many exceptions in σ

Df: Computation with advice: the TM gets an extra input, (advice), which depends only on the size of the main input.

$DTIME(f(n))/g(n)$: the class of languages L s.t. \exists Turing Machine and $\exists \alpha: \mathbb{N} \rightarrow \{0,1\}^*$ where $\forall x \in \{0,1\}^*$ with $n=|x|$ $M(\langle \alpha, x \rangle)$ halts within $O(f(n))$ steps, accepts iff $x \in L$ and $|\alpha(n)| \leq g(n)$.

Then:
 $P/g(n) :=$
 $DTIME(\text{poly}(n))/g(n)$

- $P/0 = P$
- $P/1 \not\equiv P$, because $P/1$ contains the unary Halting problem?
- $P/\text{poly} = \bigcup_k \text{SIZE}(n^k + k) \dots \geq$ the advice is the circuit, TM just evaluates the circuit
 \leq we translate the TM to a circuit & hard-wire the advice in it

Circuit lower bounds for NP are hard, but at least we can make one for EXP:

Theorem: $\forall k \geq 0 \exists L \in \text{EXP} \setminus \text{SIZE}(n^k + k)$.

Proof: L is defined by the following algorithm:

1. Let α be an input of size n .
2. Let $\beta_0, \dots, \beta_{2^n-1}$ denote possible inputs for an n -input circuit: $\beta_j := j$ written in binary.
3. $C_0 \leftarrow \{ \text{all circuits of size } n^k + k \text{ with } n \text{ inputs} \}$
4. $i \leftarrow 0$
5. While $C_i \neq \emptyset$ & $i < 2^n$:
6. Simulate all circuits in C_i on input β_i , let t_i be the minority answer.
7. $C_{i+1} \leftarrow$ those circuits from C_i which gave output t_i
8. $i \leftarrow i+1$
9. If $\alpha = \beta_j$ for some $j < i$: answer t_j
 Else: answer N or arbitrary

Now: $C_{i+1} \leq \frac{1}{2} |C_i| \Rightarrow C_i \leq |C_0| / 2^i$
 $|C_0| \leq 2^{n^{k+1}} \Rightarrow$ after less than 2^n steps, we get $C_i = \emptyset$ (i.e., we don't run out of β_j 's)
 ↑ see calculations in circuit lower bound thm. } works for n large enough

So the whole algorithm runs in time $O(2^{n^{k+2}})$, so $L \in \text{EXP}$.

But no circuit in $\text{SIZE}(n^k + k)$ can agree with L on n large enough.

Improvement: Choose $k := \lfloor \log n \rfloor \dots$ then run time is in $O(2^{n^{\log n + 2}}) \in O(2^{2^n})$

So L is ~~not~~ in EEXP , but not in $\text{SIZE}(n^k + k)$ for any k . This is called EEXP or 2-EXP

So $L \in \text{EEXP} \setminus P/\text{poly}$.

Therefore $\text{EEXP} \not\subseteq P/\text{poly}$.

Theorem: If $NP \subseteq P/poly$, then $PH = \Sigma_1^P$. ← generally, it's believed that the PH class does not collapse, so there should be languages in NP with no poly-size circuits

Proof: (not shown at the lecture)

- we want to show that $\Sigma_1^P = \Pi_1^P$
- so $\Pi_1^P \subseteq \Sigma_1^P$ suffices (the other inclusion by taking complements)
- so $T_2\text{-SAT} \in \Sigma_1^P$ suffices

↑ this is $\{ \langle \varphi \rangle \mid \varphi \text{ is a true formula of the form } \forall \alpha \in \{0,1\}^n \exists \beta \in \{0,1\}^m \varphi(\alpha, \beta) \}$

↑ unquantified formula of size $O(n)$

- If $NP \subseteq P/poly$, there is a family of poly-size circuits $\{C_n\}_{n=0}^{\infty}$

Solving: given $\langle \varphi \rangle$ and α , is there β s.t. $\varphi(\alpha, \beta)$ is true?

↳ we can convert this to $\langle \varphi \rangle, \alpha \mapsto$ find that β ... still within polynomial size
 ↑ the exercise with SAT oracle earlier ... → circuits C_n

- We don't know how C_n looks, but we can guess it and verify:
 $\exists \langle C_n \rangle \forall \alpha \varphi(\alpha, C_n(\alpha))$... this solves $T_2\text{-SAT}$, but it is in Σ_1^P .

PROBABILISTIC ALGORITHMS (a.k.a. randomized)

Define the Probabilistic TM (PTM): random states & exactly 2 possible instructions, the TM decides by flipping a fair coin (i.e., generates an uniformly random bit, independent of all the other random bits)

← another form of non-determinism like \exists & \forall states

We have a probability distribution on computations
 ↳ PTM accepts α

Df: • $BPTIME(f(n)) :=$ class of all languages L s.t. \exists PTM: all computations halt within $O(f(n))$ steps and $P(M(\alpha)) = L(\alpha) \geq 2/3$.
 ↳ indicator of $\alpha \in L$ 2-sided error

• $RTIME(f(n)) :=$ class of all languages L s.t. \exists PTM: all computations halt within $O(f(n))$ steps, if $\alpha \in L: P(M(\alpha) \text{ accepts}) \geq 2/3$ if $\alpha \notin L: P(M(\alpha) \text{ accepts}) = 0$ 1-sided error

RP := $RTIME(poly(n))$, co-RP (error at the opposite side)

Amplification of probability of success:

① For RP: Let LRP, M the corresponding PTM. Run $M(\alpha)$ t times independently, accept if at least one run accepted. → this is still poly-time...
 $\alpha \notin L$: always rejected
 $\alpha \in L$: rejected with pr. $\leq (1-2/3)^t$

↓ symmetrically for co-RP

Corollary: Iterating decreases pr. of error exponentially with #tries. This works whenever the original $P(\text{accepts}) \geq c$ for any $c > 0$. So the definition is robust w.r.t. change of the (arbitrary) $2/3$.

② For BPP: Run t times independently, use majority answer.

assume $= 2/3$, more is obviously better

Analysis: Let random variable $X_i :=$ indicator of correct answer in try $\#i$, $E[X_i] = 2/3$

Let $X := \sum_i X_i$ (# successful tries), then $\mathbb{E}[X] = \frac{2}{3} \cdot t$

$P(\text{majority is wrong}) = P(X < \frac{1}{2}t)$

Tool: Chernoff's bound for the left tail:

Let $X_1 - X_k$ be independent random variables with domain $\{0,1\}$, $X = \sum_i X_i$, $\mu = \mathbb{E}[X]$, $\delta \in (0,1)$. Then:

this is Chernoff with $\mu = \frac{2}{3}t$, $(1-\delta) \cdot \frac{2}{3} = \frac{1}{2}$, so $\delta = \frac{1}{6}$.

Hence $P \leq e^{-\frac{(\frac{1}{6})^2 \cdot \frac{2}{3} \cdot t}{2}} = e^{-\frac{2 \cdot t}{4^2 \cdot 3 \cdot 2}} = e^{-R(t)}$

$P(X < (1-\delta)\mu) \leq e^{-\frac{\delta^2 \mu}{2}}$

So Pr. of error again decreases exponentially with #tries.

This works whenever the original machine answers correctly with $P \geq c$, where $c > \frac{1}{2}$.

Certificate-based definition: BPP is the class of all languages L s.t. $\exists V \in P$ and

$\forall \alpha \in \{0,1\}^*$ $P_{\beta \in \{0,1\}^{\text{poly}(n)}} (V(\langle \alpha, \beta \rangle) = L(\alpha)) \geq \frac{2}{3}$.

Why this is the same BPP: old \Rightarrow this: the certificate is a sequence of all random bits generated by the TM, the rest can be simulated deterministically.
 this \Rightarrow old: first generate random β , then run V .
 padded to the same size for all computations

• for RP: $LERP \Leftrightarrow \exists V \in P \forall \alpha \in \{0,1\}^* : P_{\beta \in \{0,1\}^{\text{poly}(n)}} (V(\langle \alpha, \beta \rangle) = 1) \begin{cases} \geq \frac{2}{3} & \text{if } \alpha \in L \\ = 0 & \text{if } \alpha \notin L \end{cases}$

\hookrightarrow this implies $RP \subseteq NP$

"Zero-based errors": Two definitions: ① TM runs in expected time $O(t(n))$, always answers correctly $ZTIME(t(n))$
 ② TM runs in worst-case time $O(t(n))$, can answer MAYBE. If answer is not MAYBE, it's correct. $P(\text{answers MAYBE}) \leq 1/3$.
 $ZPP := ZTIME(\text{poly}(n))$

① \Rightarrow ② Run machine for $3 \cdot t(n)$ steps, if it times out, answer MAYBE.
 $P(\text{MAYBE}) = P(\text{time} \geq 3 \cdot \mathbb{E}(\text{time})) \leq \frac{1}{3}$
 by Markov's inequality

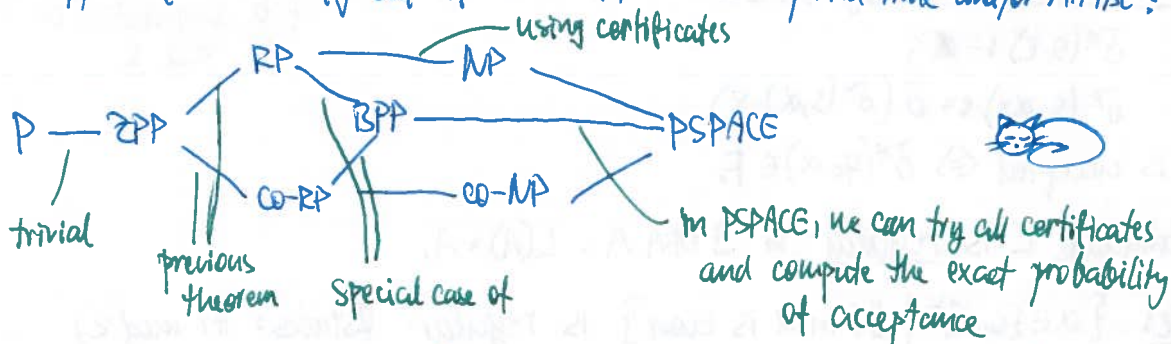
② \Rightarrow ① Run machine repeatedly as long as it returns MAYBE.
 Tool: "Water jug lemma" (a.k.a. geometric distribution)
 If $P(\text{1 try succeeds}) = p$, then $\mathbb{E}(\text{\#tries until the 1st success}) = 1/p$.
 $\mathbb{E}(\text{\#tries}) \leq 3$, so $\mathbb{E}(\text{time}) \in O(t(n))$.

Theorem: $ZPP = RP \cap \text{co-RP}$.

Proof: ① $ZPP \subseteq RP$: use worst-case def. of ZPP, translate MAYBE to NO.
 ② $ZPP \subseteq \text{co-RP}$: the same, but MAYBE \rightarrow YES.
 ③ $RP \cap \text{co-RP} \subseteq ZPP$: let M_1 be the machine witnessing LERP, M_2 for LE co-RP.
 Run both. If they agree, use their answer. Otherwise return MAYBE.
 ☹ both cannot be wrong simultaneously \rightarrow if they agree, the answer is right.
 $P(\text{MAYBE}) \leq \frac{1}{3}$ (if $\alpha \in L$, M_2 cannot fail, if $\alpha \notin L$, M_1 cannot fail)

Exercise: What happens if we modify def. of BPP or RP to use expected time and/or MAYBE?

Inclusions:



Exercise: Define class PP: $LEPP \equiv \exists M$ PTM running in w.c. time $O(\text{poly}(n))$ s.t. $P(L(\alpha) = M(\alpha)) > 1/2$ for all α .

beware: amplification doesn't work for PP (not exponentially!)

Show that: $NP \subseteq PP$, $co-NP \subseteq PP$, $BPP \subseteq PP$, $PP \subseteq PSPACE$.

Theorem: $BPP \subseteq P/poly$.

← amplify

Proof: For inputs of size n : iterate $O(n)$ times to get $P(\text{error}) \leq \frac{1}{2} \cdot 2^{-n}$

Let $r := \#$ random bits used by the machine (certificate size)

For a fixed input α : $P_{\beta \in \{0,1\}^r} (V(\langle \alpha, \beta \rangle) \neq L(\alpha)) \leq \frac{1}{2} \cdot 2^{-n} \Rightarrow \#$ "bad" certs for which this happens $\leq 2^r \cdot \frac{1}{2} \cdot 2^{-n}$

→ taking union over all α : $\#$ bad certs $\leq 2^r \cdot \frac{1}{2} \cdot 2^{-n} \cdot 2^n = \frac{1}{2} \cdot 2^r < 2^r$

\Rightarrow there exists a certificate which is good for all inputs: this will be the advice.

So our algorithm just calls V on $\langle \text{input}, \text{advice} \rangle$. This implies $LEPP \subseteq P/poly$.

Notes It is known that $BPP \subseteq \Sigma_1^P \cap \Pi_1^P$ (Sipser-Gács theorem) ← this is stronger than $BPP \subseteq PSPACE$.

There are no known BPP-complete problems nor hierarchy theorems. ← BPP is a "semantic" class, so diagonalization doesn't work. It's believed that $BPP = P$ (otherwise hard-to-believe things happen)

REGULAR LANGUAGES

Df: Deterministic Finite-state Automaton (DFA) consists of:

- Q - a finite non-empty set of states
- Σ - a finite non-empty alphabet
- $\delta: Q \times \Sigma \rightarrow Q$ - transition function
- $q_0 \in Q$ - initial state
- $F \subseteq Q$ - a set of accepting states

Df: Computation of a DFA over an input string $\alpha \in \Sigma^*$

is a sequence of states $s_0, s_1, \dots, s_{|\alpha|}$ such that $s_0 = q_0$ and $\forall i \ s_{i+1} = \delta(s_i, \alpha[i])$.

← uniquely determined

alternatively:

- DFA is a multi-graph with labelled edges (by Σ)
- computation is a walk in the graph starting in q_0 and labelled by the input α .

• The input is accepted $\equiv s_{|\alpha|} \in F$.

• $L(A) :=$ the language of all words accepted by the automaton A .

Def: Extended transition function $\delta^*: Q \times \Sigma^* \rightarrow Q \leftarrow \delta^*(s, \alpha)$ is the final state of a computation on α starting in state s . (44)

s.t. $\delta^*(s, \epsilon) := s$
 $\delta^*(s, \alpha x) := \delta(\delta^*(s, \alpha), x)$

α is accepted $\Leftrightarrow \delta^*(q_0, \alpha) \in F$.

Def: Language L is regular $\equiv \exists$ DFA $A: L(A) = L$.

Example: $\{\alpha \in \{0,1\}^* \mid \#1 \text{ in } \alpha \text{ is even}\}$ is regular (states: $\#1 \pmod 2$)

Example: Every finite language is regular (states: prefixes of words in the language)

Example: $\{0^n 1^n \mid n \geq 0\}$ is not regular. If there existed a DFA accepting it: set $t := |Q|$, consider $s_0 \dots s_t$, where $s_i := \delta^*(q_0, 0^i)$. By Pigeon-hole principle, there is $i < j$ s.t. $s_i = s_j$. Now $\delta^*(q_0, 0^i 1^i) = \delta^*(q_0, 0^j 1^i)$, so $0^i 1^i$ is accepted $\Leftrightarrow 0^j 1^i$ is. \downarrow

Lemma (Pumping lemma for regular languages):

For every regular language L , there exists $n \geq 0$ such that:

Every $w \in L$, $|w| \geq n$ can be decomposed as $w = \alpha\beta\gamma$, where:

- ① $\forall t \geq 0 \alpha\beta^t\gamma \in L$ (including $t=0$)
- ② $\beta \neq \epsilon$
- ③ $|\alpha\beta| \leq n$.

Proof: Consider an automaton accepting L . Set $n := |Q|$.

Given $w \in L$, $|w| \geq n$, define $s_0 \dots s_m: s_i := \delta^*(q_0, w[1:i])$

let $m := |w|$

Since $m \geq n$, there is $i < j \leq n$ s.t. $s_i = s_j$.

Now set $\alpha := w[1:i]$, $\beta := w[i:j]$, $\gamma := w[j:m]$ this implies ② and ③

① $\delta^*(q_0, \alpha) = s_i = s_j = \delta^*(q_0, \alpha\beta) \dots$ so $\delta^*(s_i, \beta) = s_j$, hence $\forall t \geq 0 \delta^*(q_0, \alpha\beta^t) = s_i$, so $\forall t \delta^*(q_0, \alpha\beta^t\gamma)$ is always the same. For $t=1$, $\alpha\beta\gamma \in L$, so all $\alpha\beta^t\gamma \in L$.

Example: $0^n 1^n$ again ... If it were regular, use $0^n 1^n$ with n from the lemma.

Both α, β must consist purely from 0s, so $\alpha\beta^t\gamma$ is $0^i 1^j$ and we can increase i , while staying inside the language. \downarrow over a common alphabet

Lemma: Intersection of two regular languages is regular.

Proof: Let L_1, L_2 be regular, DFA $A_1 = (Q_1, \Sigma, q_{01}, F_1)$ accepting L_1 and DFA $A_2 = (Q_2, \Sigma, q_{02}, F_2)$ accepting L_2 .

Construct a product of A_1 and A_2 :

$Q := Q_1 \times Q_2$

$\delta((s_1, s_2), x) := (\delta_1(s_1, x), \delta_2(s_2, x))$

$q_0 := (q_{01}, q_{02})$

$F := F_1 \times F_2$

we have $\delta^*((s_1, s_2), \alpha) = (\delta_1^*(s_1, \alpha), \delta_2^*(s_2, \alpha))$
 so $\alpha \in L(A) \Leftrightarrow \alpha \in L_1 \cap L_2$.

Intuition: Run A_1, A_2 in parallel, accept iff both accepted.

Exercise: Regular languages are also closed under complement and ~~the~~ union.

Df: Non-deterministic Finite-state Automaton (NFA)

Like DFA, but $\delta: Q \times \Sigma \rightarrow \mathcal{P}(Q)$ - we have multiple possible instructions to execute
and $Q_0 \subseteq Q$ replaces q_0 - multiple initial states

What changes: Computation requires $S_{i+1} \in \delta(S_i, \alpha[i])$, $S_0 \in Q_0$

There can be multiple computations for a given input, or perhaps none.

α is accepted \equiv there exists a computation ending in an accepting state.

Df: $\delta^*: \mathcal{P}(Q) \times \Sigma^* \rightarrow \mathcal{P}(Q)$ defined as:
 $\delta^*(S, \epsilon) := S$, $\delta^*(S, \alpha x) := \bigcup_{t \in \delta^*(S, \alpha)} \delta(t, x)$.

} again: α is accepted $\Leftrightarrow \delta^*(Q_0, \alpha) \cap F \neq \emptyset$.

so non-determinism doesn't increase computing power of FAs

Thm: If L is accepted by an NFA, then it is regular.

Proof: Construct a DFA $(Q', \Sigma, \delta', q'_0, F')$ which simulates δ^* of the original NFA $(Q, \Sigma, \delta, Q_0, F)$.

Let $Q' := \mathcal{P}(Q)$

$\delta'(s, x) := \delta^*(s, x)$

$q'_0 := Q_0$

$F' := \{s \in Q \mid s \cap F \neq \emptyset\}$

\rightarrow

Then $\delta'^*(q'_0, \alpha) = \delta^*(Q_0, \alpha)$,

so $\alpha \in L(A') \Leftrightarrow \alpha \in L(A)$.

Nicer generalization: ϵ -NFA, which adds ϵ -edges: these can be traversed without reading a symbol from the input

Df: Extend $\delta: Q \times (\Sigma \cup \{\epsilon\}) \rightarrow \mathcal{P}(Q)$.

Computation = walk from $q_0 \in Q_0$ s.t. concatenated edge labels yield input string.

Df: ϵ -closure $U_\epsilon(s)$ of a state s := set of all states reachable from s using only ϵ -edges.
 $U_\epsilon(S)$ of $S \subseteq Q$:= $\bigcup_{s \in S} U_\epsilon(s)$.

extending δ^* to ϵ -NFAs: $\delta^*(S, \epsilon) := U_\epsilon(S)$

$\delta^*(S, \alpha x) := U_\epsilon\left(\bigcup_{t \in \delta^*(S, \alpha)} \delta(t, x)\right)$

Thm: For every ϵ -NFA $(Q, \Sigma, \delta, Q_0, F)$, there is a NFA $A' = (Q', \Sigma, \delta', q'_0, F')$ accepting the same language.

Proof: Just add ϵ -closure: $Q' := Q$

$Q'_0 := U_\epsilon(Q_0)$

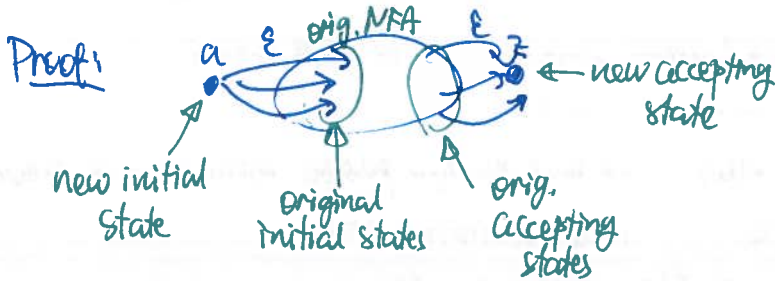
$\delta'(S, x) := U_\epsilon(\delta(S, x))$

$F' := F$

so $\delta'^*(S, x) = \delta^*(S, x)$,
hence $L(A) = L(A')$.

ϵ -NFAs accept still the same regular languages, but they are easier to construct.

Lemma: For every ϵ -NFA, there is an equivalent ϵ -NFA (accepting the same language) which has a unique initial state (with no incoming edges) and unique accepting state (with no outgoing edges)

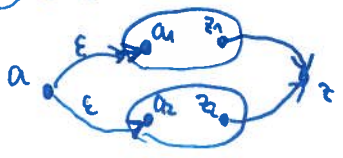


Theorem: The following operations with languages preserve regularity:

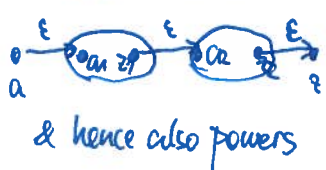
- \bar{L} complement
- $L_1 \cap L_2$ intersection
- $L_1 \cup L_2$ union
- $L_1 \cdot L_2 := \{ \alpha \cdot \beta \mid \alpha \in L_1, \beta \in L_2 \}$ concatenation (associative)
- L^k : $L^0 := \{ \epsilon \}$, $L^{t+1} := L^t \cdot L$ power
- $L^* := \bigcup_{t \geq 0} L^t$ iteration
- $L^+ := \bigcup_{t \geq 1} L^t$ positive iteration
- $L^R := \{ \alpha^R \mid \alpha \in L \}$ reversal ← word written backwards

Proof: For \bar{L} and $L_1 \cap L_2$, we already have the proof. Otherwise use ϵ -NFAs with unique init/acc. state.

① Union



② Concatenation



③ Positive iteration



④ Iteration: add union with $\{ \epsilon \}$

this is an equivalent definition of regularity which does not use automata

⑤ reversal: swap role of a, z , switch orientation of all edges.

Theorem (Kleene): L is regular $\Leftrightarrow L$ can be constructed from $\emptyset, \{ \epsilon \}, \{ x \}$ for $x \in \Sigma$, using finitely many unions, concatenations and iterations.

Proof: \Leftarrow follows from the previous thm.

Prove \Rightarrow using even more generalized NFAs, where each edge is labelled by a language and we can traverse the edge if we read a word in that language from the input.

Consider a DFA accepting L . We will transform it gradually to $a \xrightarrow{L} z$, ~~while~~ always preserving the accepted language & making sure that languages on the edges can be constructed in the required way (using $\cup, *, +$).

Steps:

① Initialization: add unique init. & acc. states:



repeat while there are parallel edges

② elimination of parallel edges: replace



by $x \xrightarrow{L_1 \cup L_2} y$ (we can have $x=y$ here)

③ elimination of states: ~~for~~ ^{remove} a state $s \neq a, z$, routing around it:

a) if s has no loops: replace all $x \xrightarrow{L_1} s \xrightarrow{L_2} y$ by $x \xrightarrow{L_1 \cdot L_2} y$

repeat until only a, z remain

b) if s has a loop: replace all $x \xrightarrow{L_1} s \xrightarrow{L_2} s \xrightarrow{L_3} y$ by $x \xrightarrow{L_1 \cdot L_2^* \cdot L_3} y$

Theorem: $DSPACE(1) = NSPACE(1) =$ class of all regular languages.

Building the proof: Deterministic machines first.

- ① TMs with just the input tape, which is read only & head doesn't move left ... this is equivalent to a DFA (technical detail: how do we accept/reject?)
- ② Allow moving left. Tech. detail: delimit the input as $\langle \alpha \rangle$. On \langle , the TM must move right, on \rangle , it must move left.

This is called the bi-directional DFA. We will prove that these accept just regular languages. (Infinite loop or divergence is interpreted as rejecting the input.)

- ③ Allow work tapes of constant size: their contents & head positions can be moved inside machine state \rightarrow this is equivalent to ②.
- ④ Non-deterministic TMs:
 - ① becomes NFA, so also regular
 - ② will need a generalized proof
 - ③ still reduces to ②.

Need to prove: If L is accepted by a bi-dir. DFA, then L is regular.

Consider computation of the bi-dir. DFA on suffixes of a given input α :

- we start on $\alpha[i]$ in some state s
- we let the computation run until
 - it stops in q^+ or q^-
 - it diverges (equivalent to q^-)
 - it leaves the suffix $\alpha[i:]$ by moving left from position i .
- we can describe this behavior by a function $f_i : Q \setminus \{q^+, q^-\} \rightarrow Q$

The f_i 's can be constructed backwards ...

- $f_{|\alpha|}$ is trivial (the TM must ~~move~~ not move right, so iterate $\bar{\sigma}$ until it moves left/stops)
- $f_{i+1} \rightarrow f_i$: for $f_i(s)$, construct a sequence of states:

$s_0 = s$

$s_j \rightarrow s_{j+1}$: if $s_j = q^+/q^-$, stop & define $f_i(s) := s_j$

otherwise evaluate $\bar{\sigma}(s_j, \alpha[i]) \rightarrow (s'_j, \text{movement})$

- if movement = \leftarrow : stop & define $f_i(s) := s'_j$
- if movement = \bullet : $s_{j+1} := s'_j$ & continue
- if movement = \rightarrow : $s_{j+1} := f_{i+1}(s'_j)$ & continue

If $s_{j+1} = s_i$ for $i \leq j$, the machine diverged, so $f_i(s) := q^-$ & stop.

$\Rightarrow f_i$ is a function of f_{i+1} and $\alpha[i]$.

So there is a DFA processing α^R , whose states are the f_i 's.

α^R is accepted $\Leftrightarrow f_{|\alpha|}(q_0) = q^+$

minor technicality:
we let the TM start on \leftarrow
instead of the first char. of α

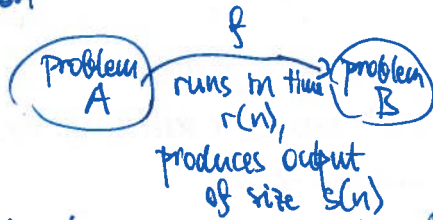
so L^R is regular, therefore L is also regular.

FINE-GRAINED COMPLEXITY

Goal: Finer results than "polynomial vs. exponential"
 E.g., prove that a $\Theta(n^2)$ -time alg. is optimal.

Caveats: This will be model-dependent. We will assume RAM here.

Tool: Fine-grained reduction



If B can be solved in time $T(n)$, then A can be solved in time $O(r(n) + T(s(n)))$
 ↑ covers copying of input/output of f

Upper bounds for B imply upper bounds for A.
 Lower bounds for A imply lower bounds for B.

Orthogonal Vectors Problem (OV):

Inputs: two sets of vectors $A, B \subseteq \{0,1\}^d$, $|A|, |B| \leq n$

Questions: are there $a \in A, b \in B$ s.t. $\langle a, b \rangle = 0$? ← i.e., bitwise AND is everywhere zero

Baseline algorithms: $O(n^2 d)$ trivial, $O(nd \cdot 2^d)$ ← for each $a \in A$, construct all orthogonal vectors and look them up in a suitable data structure for B (e.g., a trie)

Hypothesis (OVH): For no $\epsilon > 0$, there is an algorithm solving OV in time $O(n^{1-\epsilon} \cdot \text{poly}(d))$.

NFA Acceptance Problem (NFAA):

Input: Non-deterministic finite-state automaton M of size $|M| = \#states + \#transitions$, string α .

Query: Does M accept α ?

Baseline: $O(|M| \cdot |\alpha|)$ by computing δ^* (see previous lecture)
 $O(2^{|M|} + |\alpha|)$ by reducing to a DFA first.

Theorem: Assuming OVH, there is no $\epsilon > 0$ s.t. NFAA can be solved in time $O((|M| \cdot |\alpha|)^{1-\epsilon})$.

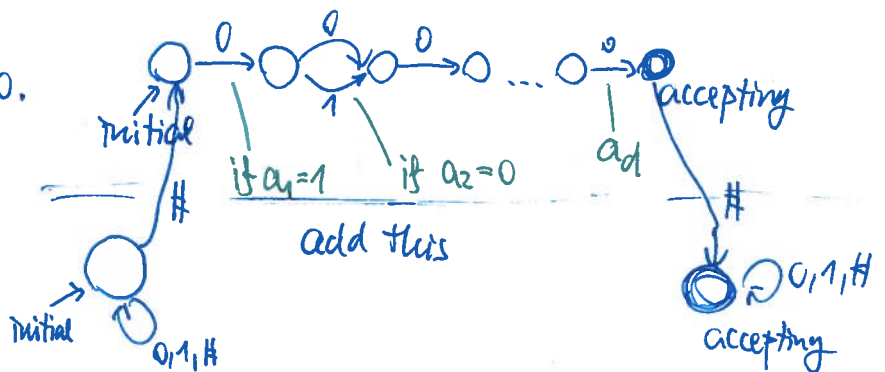
Proof: Will show reduction $OV \rightarrow NFAA$ running in time $O(nd)$, producing $|M| \in O(nd)$, $k \in O(nd)$.

If NFAA can be solved in $O((|M| \cdot |\alpha|)^{1-\epsilon})$ time for some $\epsilon > 0$, then OV can be solved in $O((nd)^{1-\epsilon}) = O(n^{2-2\epsilon} \cdot d^{2-2\epsilon})$ time, contradicting OVH.

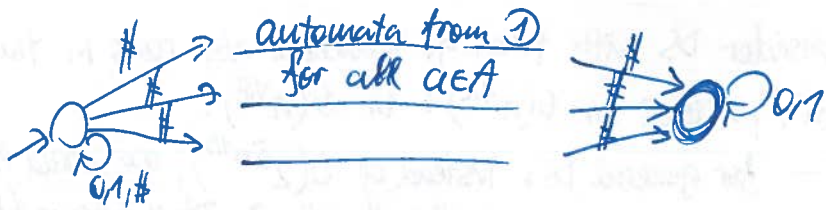
Now the reductions:

① given $a \in A$, construct NFA which accepts $b \Leftrightarrow \langle a, b \rangle = 0$.

② given a , construct NFA accepting $\#b^1\#b^2\#\dots\#b^n\#$
 $\Leftrightarrow \exists i: \langle a, b^i \rangle = 0$



③ Add a choice of $a \in A$:



Surprisingly, fine-grained bounds are connected with the "big world" of P vs. NP.

Exponential Time Hypothesis (ETH): ~~Formula~~ $\exists \epsilon > 0$ s.t. 3-SAT can't be solved in $O(2^{\epsilon N})$ time.

↳ justification: baseline alg. is $O(2^N \cdot \text{poly}(M,N))$ -time
 state-of-the-art alg. is $O(1.3280^N \cdot \text{poly}(M,N))$ -time.

for k-SAT:
 $N := \#$ variables,
 $M := \#$ clauses

Obviously, ETH implies $P \neq NP$.

improvements don't seem to converge towards 1

Strong ETH: $\forall \epsilon > 0 \exists k$ s.t. k-SAT cannot be solved in time $O(2^{\epsilon N})$.

dependence on M is not important as $M \leq \binom{N}{k} \cdot 2^k$, which is polynomial for fixed k

(SETH) ↳ justification: state-of-the-art kSAT algs are slower for higher k , converging towards 2^N .

It's known that SETH \Rightarrow ETH.

Dominating Set problem (DS)

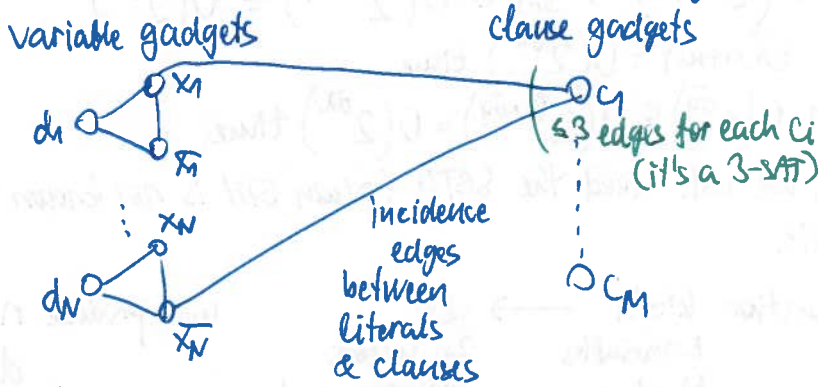
Input: undirected graph with n vertices, $q > 0$

Question: is there a dominating set D of size q ?

↳ for $G=(V,E)$: $D \subseteq V, \forall u \in V \exists v \in D. (u=v \text{ or } \{u,v\} \in E)$
 u dominated by v

Theorem: DS is NP-complete.

Proof: DS \in NP is trivial, will show NP-hardness by reducing 3-SAT to DS.



Will set $q=N$. This implies that a dom. set must use exactly 1 vertex from each var. gadget.

Formula satisfiable \Rightarrow choose dom. set according to assignment, check that all clause vertices are dominated.

Dom. set exists \Rightarrow choose assignment according to literals used in D ($d_i \in D \Rightarrow$ choose x_i arbitrarily), check that the formula is satisfied.

Theorem: ETH $\Rightarrow \exists \delta > 0$ s.t. DS cannot be solved in time $O(2^{\delta \cdot n^{1/3}})$.

Proof: Finer analysis of the same reduction.

Graph has n vertices, m edges for $n = 3N + M$
 $m \leq 3n$

we have $M \leq \binom{N}{3} \cdot 2^3 \leq 2N^3$,
 so $n \leq 3N^3$ for N large enough,
 $m \in O(n)$

Reduction runs in $O(n+m) = O(N^3)$ time.

If DS can be solved in $O(2^{\delta \cdot n^{1/3}})$ time, then 3SAT can in $O(2^{3^{1/3} \cdot \delta \cdot N})$.
 For δ small enough, this contradicts the ETH.

Now consider DS with fixed q . Baseline alg. runs in time $O(\binom{N}{q} \cdot qn) \leq O(n^{q+1})$.

Is $O(n^q)$ possible? Or $O(n^{q/2})$? Or $O(n^{\sqrt{q}})$?

For general DS: instead of $O(2^{\delta n^{1/3}})$, we would like $O(2^{\delta n})$... how to get rid of N^3 in the #vertices? It is known (but we won't prove it here) that 3-SAT is hard even for sparse formulas ($M \in O(N)$).

Theorem: If ETH holds, then $\exists \delta > 0 \forall^* q$: DS cannot be solved in $O(n^{\delta q})$ time.

Proof: We will show that a $O(n^{\delta q})$ -time alg. for DS with $q \geq \frac{2}{\delta}$ implies a $O(2^{\delta N})$ alg. for 3-SAT. So for δ small enough, this would contradict the ETH.

Modify the previous reduction of 3-SAT to DS:

- ~~divide~~ ^{partition} variables to q groups per M/q variables
- variable gadgets: for each group, create vertices for all partial assignments setting variables in the group $\rightarrow 2^{M/q}$ vertices + add an extra d_i vertex edges form a clique
- clause gadgets: for each clause, add vertex c_i connected to all partial assignments which satisfy this clause

Again, a DS of size q selects a 1 vertex from each var. gadget. This either selects one partial assignment to vars in the group or $d_i =$ "pick any".

Graph size: $n = q(2^{M/q} + 1) + M \in O(2^{M/q})$ for fixed q

$m = (2^{M/q} + 1)^2 + 3M \in O(2^{2M/q})$ ^{because of *} $\in O(2^{\delta N})$

Reduction takes $O(n+m) = O(2^{\delta N})$ time.
SAT is solved in $O(n^{\delta q}) \leq O(2^{\frac{M}{q} \cdot \delta q}) = O(2^{\delta N})$ time.

• For hardness of OV, we will need the SETH (plain ETH is not known to suffice)

Theorem: SETH \Rightarrow OVH.

Proof: Will show a reduction k -SAT \rightarrow OV will produce $n = 2^{M/2}$

N variables	$2n$ vectors	$d = M$
M clauses	dimension d	

in time $O(nd)$, assuming k fixed.

So a $O(n^{2-\epsilon} d)$ alg. for OV implies a $O(2^{\frac{2-\epsilon}{2} N \cdot M})$ -time alg. for k -SAT, contradicting SETH for k large enough.

Reduction: Split variables to 2 groups X, Y of size $N/2$.

Construct A using X :

- vectors correspond to partial assignments to X $\left. \vphantom{\begin{matrix} \bullet \\ \bullet \\ \bullet \end{matrix}} \right\} 2^{N/2}$ vectors
- coordinates correspond to clauses $\left. \vphantom{\begin{matrix} \bullet \\ \bullet \end{matrix}} \right\} M$ coordinates
- "0" means that the clause is satisfied by the partial assignmt.

Similarly, construct B using Y .

 $\langle a, b \rangle = 0 \Leftrightarrow$ all clauses are satisfied by a ~~exactly~~ union of the two part. assignmts.

Problem: Longest Common Subsequence (LCS) — define $L(\alpha, \beta) :=$ ~~the~~ max. length of a common sub-sequence of α, β (51)

Input: strings $\alpha, \beta \in \Sigma^*$, $|\alpha|, |\beta| \leq n$; $c > 0$

Output: Is $L(\alpha, \beta) > c$?

↑
Obtained by deleting elements while preserving order (i.e., not a subword)

Theorem: LCS can be solved in time $O(n^2)$ independent of $|\Sigma|$.

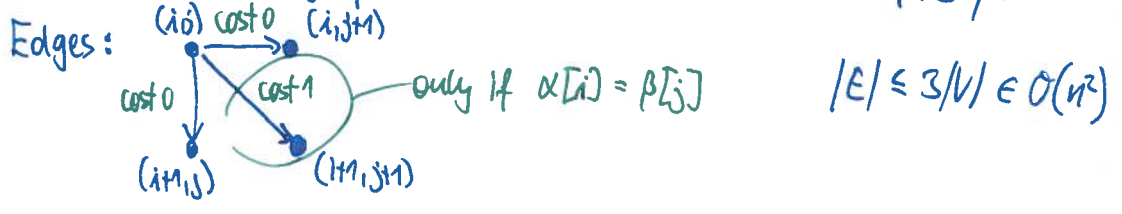
Proof: Let $A[i, j] := L(\alpha[1:i], \beta[1:j])$.

We have $A[0, j] = A[i, 0] = 0$,

$$A[i, j] = \min \begin{cases} A[i-1, j] \\ A[i, j-1] \\ A[i-1, j-1] + 1 \end{cases} \text{ only if } \alpha[i-1] = \beta[j-1]$$

Using this, we can fill in the table of $A[i, j]$'s row by row in time $O(n^2)$.
Then $A[n, n] = L(\alpha, \beta)$.

Alternative proof: Define a directed graph with $V = \{0\} \times \{0\} \cup \{1 \leq i \leq |\alpha|\} \times \{1 \leq j \leq |\beta|\}$, $|V| \in O(n^2)$



Path from $(0,0)$ to $(|\alpha|, |\beta|)$ of cost c corresponds to a common subseq. of length c .

↳ LCS = cost of the longest path -- but since the graph is acyclic, this can be computed using the same recurrence as in the previous proof. → the same $O(n^2)$ -time alg.

Theorem: Assuming ~~ETH~~ ETH, for no $\epsilon > 0$ there is a $O(n^{2-\epsilon})$ -time alg. for LCS.

Proof: Omitted, see the lecture notes by Karl Bringmann.

That's all, thanks for your attention! ☺