

Complexity Theory

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Chapter 1

Problems, Solutions, and Resources

1.1 Problems

Alphabet set is finite, call it Σ . Strings must be finite length.

Given some input, and a computer that produces some output, the description could be infinite — both input and output.

However, the machine's *description* (aka, the relationship between input and output) must be finite.

So, the *total input* can be infinite, but the input chunk must be finite, and the response of the machine per *input chunk* must be finite.

So, we can just use the language $L = \{0, 1\}$ for the machine.

Problems which have yes/no as answers are called decision problems. Inputs are from Σ^* , outputs are from $\{0, 1\}$. The problem is a mapping $f : \Sigma^* \rightarrow \{0, 1\}$. This is equivalent to providing the set $\text{ACCEPT} \subset \Sigma^* = f^{-1}(1)$. Note that $\text{REJECT} = \text{ACCEPT}^c$. The set ACCEPT is called a language.

Now, we can study languages by looking at their grammars (welcome, Chomsky).

What about fractional bit problems? Is this useful? Could we exploit some properties of fractional dimension?

Cantor set

take $S_0 = [0, 1]$ In each iteration, remove the middle one-third of each continuous interval. Therefore,

- $S_0 = [0, 1]$
- $S_1 = [0, \frac{1}{3}] \cup [\frac{2}{3}, 1]$

In S_∞ , *uncountably infinite* points remain (However, this set has *measure* 0).

So now, the question is, what is the dimension? We define Hausdorff dimension, and use this to exhibit fractional dimension of the Cantor set.

TODO: fill this up!

The total number of problems that can exist is $\text{powerset}(\Sigma^*)$. RE (recursively enumerable) Is a subset of $\text{powerset}(\Sigma^*)$ which computers can handle. The annoying thing is that there are *finite length problems* which computers cannot solve.

1.1.1 Kannan

If the universe is a machine, then it must have infinite description.

QM is the meeting point of universes?

1.2 Solutions

1.2.1 Kannan

Question: We study a lot of Science — why? What is the ultimate goal of science? Equivalently, what is the theory of everything we need to find to halt on the journey of Science?

Assuming Science = God, we need to ask Science a question. Which language will we use to query Science? Or, well, which language is *enough* to query Science? If the query alphabet is Σ , we can ask Σ^* questions. However, we can only reasonably pose questions of finite length (even though the Science oracle can answer questions of infinite length).

In this case, have we achieved the ultimate goal of science?

1.3 Resources

Chapter 2

Diagonalization

- Level 1: \mathbb{R} is uncountable.
- Level 2: $\exists L, L \notin \text{RE}$.
- Level 3: Halting problem is undecidable.
- Level 4: Time/Space hierarchy.
- Limitations: Exists oracles A, B such that — $P^A = NP^A, P^B \neq NP^B$
- Level 5: If $P \neq NP$, $\exists L, L \notin P, L \notin NPC$ (Ladner's theorem)

Diagonalization cannot separate P, NP — If it could, then it should also separate P with any oracle, and NP with the same oracle. We know that there exists an oracle such that we can separate $P^A = NP^A$, as well as $P^B \neq NP^B$.

chapter Review of the last 3 lectures, after add-drop

- Is it easier to *pose* problems than to *solve* them?
- Can every "solvable" problem have a solution that uses finite resources?
- What problems are *interesting*?
- Are all interesting problems solved in an *interesting* way? (P v/s NP)
- Can things get more interesting? (Quantum Mechanics, Approximation, Randomness, Interactivity, ...)

Are there problems with infinite length input / output but can still be posed in finite time? Eg. output π in decimal. However, we decided that both input/output should be finite. We decided this does not belong to problems we wish to solve it, since we cannot solve it in finite time. If we believe that nature is inherently noisy, or nature is quantized, or nature has finite precision, then we cannot consider problems that require infinite time as problems in this universe (since Nature / the universe itself cannot pose such a problem).

Quantum mechanics (which is a theory of quantization) is developed over infinite precision mathematics (\mathbb{C}). Does this really make sense? There is a way in which a quantized universe can be infinite precision: This is by using 'external help': There are infinite such quantized universes which intersect at some points, and at those points, precision will increase. (If we both have a resolution of 1 pixel but are at a gap of $1/2$, my least count is now $1/2$). If there are an infinite number of universes overlapping at a single point, then we can construct "infinite precision". (*I feel this is crazy. Is this really crazy?*)

Posing a question is creating a language $L \subset \Sigma^*$. (Sid: a solution is a classifier for L).

Kannan's view:

- Finite space \equiv finite information can be stored. (Turing: finite tape alphabet. Since a cell demarcates a finite volume, we want to have a finite amount of info in this cell)
- Information travels at finite speed. If we have cells, we should not be able to store and retrieve information "equally" (based on how far we are from it). Hence, all infinite memory must be sequential memory since information travels at finite speed.
- Finite program \equiv finite control.

Solution to these choices is a TM.

Do all languages have a TM recognizing it? No (**RE** = solvable by TM).

The class **R** = decidable by a TM (TM halts on all inputs). Diagonalization led us to Halting problem.

We have the class P , and we claimed that P is interesting. Given that P is considered interesting because of feasibility, it is possible that there are questions that are interesting even though **solving them** is not feasible. For example, if we can actually **understand** the solution, or the proof of non-existence of solutions, then we will care. $IP = PSPACE$ is one such magical case where if someone can solve with a lot more power than you have access to, you can learn things from them interactively in reasonable time.

Chapter 3

Hierarchy Theorems

$\exists L$, such that $\forall f : \mathbb{N} \rightarrow \mathbb{N}$, where f is space/time constructible,

$$\begin{aligned} \text{Space}(f) &\supsetneq \text{Space}(o(f)) \\ \text{Time}(f) &\supsetneq \text{Time}\left(\frac{o(f)}{\log f}\right) \end{aligned}$$

So, there is a Hierarchy of complexity classes in time and space.

3.0.1 Proof sketch

We exhibit a language A , such that $A \in \text{Space}(f(n))$, and $A \notin \text{Space}(o(f(n)))$.

Let D decide A . D 's definition:

- compute $f(n)$ and mark the end of $f(n)$ cells. If the read-write head ever crosses it, **REJECT**, **HALT**. We first need $f(n)$ to use $f(n)$ cells or less to compute. This is called as **space-constructibility**. ($f : \mathbb{N} \rightarrow \mathbb{N}$ is space-constructible iff given n , \exists TM which computes $f(n)$ using at most $f(n)$ cells). Also, we want $f(n)$ to be at least $\log(n)$. Clearly, this process is in space $f(n)$.
- We now need to "separate" A from the smaller classes. If A can be solved in a smaller space (ie, we cannot separate A), then there must be a TM (say, D') which decides A in space less than $f(n)$. So now, we need to choose some input such that D' is different from D . We can use diagonalization to construct such a function.
- let the input be x . Let $x = M10^*$ for some TM M . if not, **REJECT**, **HALT**.
- Let D simulate M on input M . If M takes less than $f(n)$ time to run on M , then M can decide A in time less than $f(n)$. So now, D knows how much space $M(\langle M \rangle)$ requires. if $M(\langle M \rangle)$ accepts, we reject. If $M(\langle M \rangle)$ rejects, we accept (diagonalization).
- To find out whether $M(\langle M \rangle)$ rejects, note that it is space-bounded, so we can just check how many states of the configuration space it visits. If it has not halted after visiting all states in the configuration space, we can conclude that $M(\langle M \rangle)$ does not halt. The configuration space is $O(2^{f(n)})$. So we need to run D for time $O(2^{f(n)})$, and then **REJECT** if it continues running.

Arjun Q: Are there examples of non-space constructible functions, which are non-trivial? Other than ones that are too-small?

Proofs of time are similar to the space separation theorem.

- compute $t(n)$ ($t(n)$ should be time-constructible). decrement a counter initialized to $t(n)$. if this hits 0, REJECT, HALT. We get a \log factor due to the slowdown of keeping time. (People are trying to speed this up).
- once again, repeat the same construction used for $SPACE$.

3.1 Savitch's Theorem: $NSPACE(f(n)) \subseteq PSPACE(f(n)^2)$

$NSPACE(f(n))$ – one branch of a NTM N decides L in space $O(f(n))$.

Configuration space is $O(\text{alphabet}^{f(n)}) = O(2^{f(n)})$ – otherwise, configurations are repeated.

Our branch depth is exponential in $f(n)$. So, we need to keep track of $O(2^{f(n)})$ data.

Given $\langle C_1 \in Config(N), C_2 \in Config(N), t \in \mathbb{N} \rangle$ if we can find whether C_1 goes to C_2 in t space, then we can solve our original problem.

This can be solved by recursion by asking if there exists a C_{mid} , such that $C_1 \rightarrow C_{mid}$ in $t/2$ steps, similarly $C_{mid} \rightarrow C_2$ in $t/2$ steps.

3.2 Cook Levin theorem

L is NP-complete, if

- $L \in NP$
- $\forall L' \in NP$, there exists a Karp reduction from L' to L : $L' \leq_p L$ (NP-hard)

$A \leq_p B$ if there exists a poly time computable function $f : \Sigma^* \rightarrow \Sigma^*$ such that

$$w \in A \Leftrightarrow f(w) \in B$$

Karp reduction = poly time mapping reduction.

Define SAT, and show that SAT is NP-complete.

We have boolean formulas ϕ , which is given in CNF.

$$\text{SAT} = \{\phi \mid \phi \text{ is in CNF (product of sums), } \phi \text{ is satisfiable}\}$$

This is clearly decidable since we can try all possible assignments.

It is in NP since a NDTM can try to guess assignments.

To show that this is NP-complete, take any language L' in NP. We provide a karp reduction to SAT. We take the poly-time checker for L' into a SAT problem ψ , such that **iff** a solution for ψ exists, then the poly time checker will accept the string, and vice versa (for reject).

3.3 EXPSPACE completeness - $EQ_{REG\uparrow}$

$r \uparrow \equiv \exists k \in \mathbb{N}. r \uparrow$ is regular if r is regular. We need this operator to control input size.

Question: Check if two regular expressions are the same – We show that this \notin PSPACE, and hence \notin PTIME. We show that this problem is EXPSPACE complete.

Chapter 4

NP

4.1 Cook Levin theorem

SAT is NP-hard.

4.1.1 Proof

. Unfold theorem statement into: $\forall L \in \text{NP}, L \leq_p \text{SAT}$. Since this should work for all things in NP, let's just write down the definition:

there exists an NDTM N such that N accepts w , $\forall w \in L$, in $|w|^k$ steps.

N is an NDTM, so N accepts w means that there exists a branch of N that accepts w in $|w|^k$ steps.

We should be able to construct a CNF such that $\phi(w)$ is SAT iff there exists an accepting branch for $N(w)$.

Caveats

1. the construction of ϕ from N should make sure that ϕ has $\text{poly}(|w|)$ clauses — otherwise, this is no longer a poly-time reduction. We know that $\langle \text{AND}, \text{OR}, \text{NOT} \rangle$ is universal, so we can clearly construct any TM into a circuit. The problem is that the CNF we construct from the truth-table of the TM will be polynomial.

Sid Q: Proof that boolean circuits are universal?

Proof sketch

Consider the NDTM $N(n)$, we will now argue about its configuration.

We can cut off the turing tape after the first polynomial number of cells — since the NDTM can only access those many cells.

We should start with the initial state q_{start} .

We should get the accept state q_{accept} in n^k steps.

If we can pose this in terms of a CNF formula of poly-length, we are done.

Setting up SAT

Variables - Cells of the tape The state of the turing tape on the i th step at the j th position of the turing tape for all $s \in \text{alphabet}(N)$ as $x_{i,j,s}$. $x_{i,j,s} = 1$ is interpreted as "at step i , on cell j , value s is written."

For this to be valid, we need each cell to have exactly one symbol.

Formula - Validity of cells For every (i, j) for at **least one** s must be 1: $\phi_{\text{celleat}} = \bigwedge_{i,j} (\bigvee_s x_{i,j,s})$

For every (i, j) for at **most one** s must be 1. This is equivalent to saying that for every (s, t) , one of them must be absent. $\phi_{\text{cellmost}} = \bigwedge_{s,t,s \neq t} (\neg x_{i,j,s} \vee \neg x_{i,j,t})$.

$$\phi_{\text{cell}} = \phi_{\text{celleat}} \wedge \phi_{\text{cellmost}}$$

Formula - Initial state The initial cells contain the correct letters, corresponding to the initial input $w = \langle w_1 w_2 \dots w_n \rangle$, and the other cells must be blank. $\phi_{\text{init}} = x_{1,0,\#} \wedge x_{1,0,q_{\text{start}}} \wedge (x_{1,1,w_1} \wedge x_{1,2,w_2} \wedge \dots \wedge x_{1,n,w_n}) \wedge (x_{1,n+1,\text{blank}} \wedge x_{1,n+2,\text{blank}} \wedge \dots \wedge x_{1,n^k,\text{blank}})$

We use the $x_{\text{STATE,LOC,STATE ALPHABET}}$ to encode the current state we are in. We adjoin this to the alphabet since we automatically get mutual exclusion. The size of the STATE ALPHABET is constant, so it doesn't matter. The LOC of this will correspond to the **location of the head**. So, by placing the state symbol at a point, we understand that the head is at $(\text{LOC} + 1)$ in the original tape.

Formula - Final state To encode a TM configuration, what we need to know is the state of the TM, the position of the head, and the letters on the tape.

$$\phi_{\text{accept}} = \bigvee_{i,k} x_{i,j,q_{\text{accept}}}$$

Formula - Transition Every valid move in the TM can only touch two adjacent cells, since the memory is not really 'random-access', it localizes computation.

So, every transition formula will view three adjacent cells at once. If we can view three are valid, we can "slide the window" to check the correctness of all cells.

$$\phi_{\text{move}} = \bigwedge_{\text{valid adjacent triplets}}$$

To identify valid adjacent triplets, let the original config be $\langle \mathbf{a} \mid \mathbf{b} \mid \mathbf{c} \rangle$, let the new config be $\langle \mathbf{d} \mid \mathbf{e} \mid \mathbf{f} \rangle$.

The total number of legal windows will be a subset of s^9 . So, for any legal window, we can create a formula of some constant size.

$$\phi_{\text{move}} = \bigwedge_{\text{legal window.}}$$

The number of legal windows is $O(n^k)$, since we only have n^k steps and each legal window adds some constant number of formulas.

Final ϕ : $\phi = \phi_{\text{cell}} \wedge \phi_{\text{init}} \wedge \phi_{\text{accept}} \wedge \phi_{\text{move}}$ If the NDTM accepts, then ϕ will be valid and vice versa.

Sid Q: Check that this reduction actually takes *poly-time in n* ! It is clear to me that this inflates the size in poly, but it's not clear to me that this takes only *poly-time in n* .

Hence, SAT is **NP-hard** (well, NP-complete).

4.2 3-SAT is NP-complete

3-SAT: Every clause has 3 literals.

We reduce 4-SAT into 3-SAT, and then we show how to reduce **n-SAT** to **(n-1)-SAT**

$$C_{4\text{-SAT}} = x_1 \vee x_2 \vee x_3 \vee x_4$$

$$C'_{3\text{-SAT}} = (x_1 \vee x_2 \vee z) \wedge (x_3 \vee x_4 \vee \neg z)$$

If C is true, We can always pick an assignment for z to make the **other clause true**. So, we can pick a correct z .

Similarly, if C is false, C' reduces to $C' = z \wedge \neg z$. This is **UNSAT**, so this is not possible.

Now, in general, we reduce **n-SAT** into **(n-1)-SAT**.

4.3 CLIQUE is NP-complete

Does the graph contain a k clique?

Proof.

Reduce 3-SAT to **CLIQUE**

l clauses, m variables.

For every clause and its complement, create a collection of vertices. Within a triplet, there are no edges. All triplets are across. Connect everything to everything, unless they are contradictory.

The number of edges will be high.

Now, ask for an l clique. If this exists, then we have l assignments which do not contradict each other.

Chapter 5

PSPACE

5.1 PSPACE-completeness

L is PSPACE-complete if:

- $L \in \text{PSPACE}$
- $\forall A \in \text{PSPACE}, A \leq_p L$ (polytime-reduction)

5.1.1 TQBF - Totally quantified boolean formula

A boolean formula ϕ . (For example: $\phi = (x_1 \vee x_2) \wedge (x_3 \vee \neg x_2)$). TQBF will be of the form $\exists x_1, \forall x_2, \exists x_3 \dots, \phi$. SAT was $\exists x_1, \exists x_2, \dots, \exists x_n, \phi$. TQBF allows "forall" quantification.

5.1.2 Winning strategies in games

Winning strategies are basically objects of the form:

WinningStrat $\equiv \exists$ a move x_1 , such that \forall moves x_2, \exists a move $x_3, \dots, \phi(x_1, x_2, \dots)$

WinningStrat looks like the specification of a winning path in a game tree! In the literature, this game is called the **FORMULA-GAME**.

So, we choose to work on TQBF now.

5.1.3 Proof of PSPACE-completeness of TQBF

Proof of $\text{TQBF} \in \text{PSPACE}$:

Let *problem* = $Q_1 x_1, Q_2 x_2, \dots \phi$. Let the solution procedure be called *S*.

$S(Q_1 x_1 Y)$. First check $x = 0, S(Y)$, next $x = 1, S(Y)$. When we recurse for $x = 1$ from doing $x = 0$, we try to reuse space. Once we get the answer, we need to **AND** (for \forall) or **OR** (for \exists) correctly.

$\text{Space}(n) = \text{Space}(n - 1) + \text{poly}(n)$ to store the answer bits. We get $\text{Space}(n) = \text{Space}(n - 1) + \dots$, because we can **reuse the space in the two recursive invocations**.

If we solve the recurrence, we find that this is in PSPACE.

Proof of $L \leq_p$ TQBF:

We want to solve the problem $\Phi_{c_{init}, c_{accept}, O(2^{f(n)})}$, where $\Phi_{x,y,t}$ is the predicate that links x to y in configuration space in t steps, and $f(n)$ is the time used by the turing machine for L .

We are trying to solve the reachability of the initial state to the final state of the turing machine for L for any given input w in TQBF. So, we create the configuration graph $G_{\langle L, w \rangle}$, and we encode the path problem from c_{init} to c_{accept} using Φ .

$$\Phi_{c_1, c_2, t} = \exists c_m, \Phi_{c_1, c_m, t/2} \wedge \Phi_{c_m, c_2, t/2}.$$

$$\text{But this is equivalent to: } \Phi_{c_1, c_2, t} = \exists c_m, \forall (c_3, c_4) \in \{(c_1, c_m), (c_m, c_2)\}, \Phi_{c_3, c_4, t/2}.$$

This does not cause a blow up in formula length - at each step, formula length increases by a constant amount! Also, this reduction is quadratic time, so the reduction happens in *poly*.

Hence, TQBF is PSPACE-complete.

5.2 Relativization – P versus NP (Baker Gill Soloway '75)

there exists oracles A, B such that — $P^A = NP^A, P^B \neq NP^B$. If our proof for $P \neq NP$ tries to relativize, then our proof will not work, because we have choice of oracles A and B which allow for both $P^A = NP^A$ and $P^B \neq NP^B$.

5.2.1 Intuition: program diagonalization proofs relativize

Assume that we are applying diagonalization of programs. What we are actually doing is we are talking about the execution of programs on a universal turing machine $U(M, x)$ where M is the programs we diagonalizing on.

Now, if we have M^A , we can simply give the universal TM access to A , and then $U^A(M^A, x)$ can be diagonalized the exact same way. So, in some sense, program diagonalization commutes with relativization.

5.2.2 Proof of BGS

$$P^{TQBF} = NP^{TQBF}$$

$NP^{TQBF} \subset NPSPACE$, since $NPTIME \subset NPSPACE$. (replace oracle call to TQBF with actual code, everything will live in NPSPACE).

This means that $NP^{TQBF} \subset NPSPACE = PSPACE$.

Next $PSPACE \subset P^{TQBF}$, by the same argument that we had used for NP .

Hence,

$$NP^{TQBF} \subset NPSPACE = PSPACE \subset P^{TQBF} \text{ (in fact, } PSPACE = P^{TQBF} \text{).}$$

$$P^B \neq NP^B:$$

Let us assume we have our relativizing oracle B .

$U_B = \{n \in \mathbb{N} \mid \exists x \in \Sigma^*, |x| = n, x \in B\}$. Informally, U_B is the set of all possible string lengths in the language B .

$U_B \in NP^B$, because for a given n , it will try to guess the string x which will be a certificate for n , and will verify it using the oracle-access to B .

Next, we wish to show that $U_B \notin P^B$. We need to build B in such a way that this happens. We build B in stages.

At stage i , some finite number of strings would be put in B . Take a TM M_i which runs in $O(n^i)$. We want $M_i^B \notin P$. Consider $M_i^B(k)$. Some oracle calls from M_i have been made before. If the oracle call was made before, then be consistent with the answer given before.

However, if we get a **new** query, we want a setup such that $k \notin U_B$ iff $M_i^B(k) = \text{YES}$, and similarly $n \in U_B$ iff $M_i^B(k) = \text{NO}$. The way we do this is by making k so large that it is not possible to check efficiently whether $k \in U_B$.

Make k large enough that all of it cannot be seen in polytime. Now, in this case, extend B such that checking if k belongs to U_B is outside the reach of $M_i^B(k)$.

TODO: read this proof, this makes no sense to me

Chapter 6

LogSpace

6.1 The languages L & NL

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

Open problem: L versus NL.

$L \subset P$ since configuration space is $2^{O(\log(n))} = O(n)$, hence we can brute force the configuration space.

6.1.1 NL-completeness

A language A is NL-complete if:

- $A \in NL$
- $B \in NL, B \leq_L A$

Note that $B \leq_L A$ uses a mapping reduction: $\exists f : \Sigma^* \rightarrow \Sigma^*$, f converts inputs of B to inputs of A in log-space.

f must be **implicitly computable**: That is, the problem of generating the i -th bit should be done in L.

6.2 The PATH problem

$PATH = \{\langle G, s, t \rangle \mid \exists s \rightarrow t \text{ path in } G\}$

Theorem 1 $PATH \in NL$

Proof. We perform the naive solution — We try to explore all possible path possibilities for $s \rightarrow t$.

Theorem 2 $A \in NL \implies A \leq_L PATH$

Proof. Since $A \in NL$, there exists an NDTM M which decides A by using $O(\log(n))$ space.

We try to walk through the configuration graph of M (henceforth called $Config_M$, from which we can try to find the path from the initial configuration to the final configuration. We can then try to solve $PATH(Config_M, c_{init}, c_{final})$, which will tell us whether an accepting path exists.