

# COMPLEXITY THEORY

## Lecture 16: Alternation

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### Baker, Gill, Solovay

**Theorem 14.18 (Baker, Gill, Solovay, 1975):** The answer to  $P \stackrel{?}{=} NP$  does not relativise: there are languages **A** and **B** such that  $P^A = NP^A$  and  $P^B \neq NP^B$ .

**In words:** The P vs. NP problem does not relativise, and therefore cannot be solved by any techniques that do.

- Equality was shown using **A = TRUE QBF**. It is so far not known that this oracle is not in P, so this might be the world we are living in.
- Inequality was shown using **B** that diagonalises against all polytime OTM to show that they cannot decide **L<sub>B</sub>**.

## Review

## Alternation

## Alternating Computations

Non-deterministic TMs:

- Accept if **there is** an accepting run.
- Used to define classes like NP

Complements of non-deterministic classes:

- Accept if **all** runs are accepting.
- Used to define classes like coNP

We have seen that existential and universal modes can also **alternate**:

- Players take turns in games
- Quantifiers may alternate in QBF

Is there a suitable Turing Machine model to capture this?

## Alternating Turing Machines: Acceptance

Acceptance is defined inductively:

**Definition 16.2:** The set of **accepting configurations** of an ATM  $\mathcal{M}$  is the least set of configurations  $C$  for which either of the following is true:

- $C$  is existential and some successor configuration of  $C$  is accepting.
- $C$  is universal and all successor configurations of  $C$  are accepting.

$\mathcal{M}$  accepts a word  $w$  if the start configuration on  $w$  is accepting.

**Note 1:** configurations with no successor are a base case, since we have:

- An existential configuration without any successor configurations is rejecting.
- A universal configuration without any successor configurations is accepting.

Hence we don't need to specify accepting or rejecting states explicitly.

**Note 2:** defining this to be the **least** set implies that infinite runs are never enough to declare a configuration to be accepting.

## Alternating Turing Machines

**Definition 16.1:** An **alternating Turing machine** (ATM)  $\mathcal{M} = (Q, \Sigma, \Gamma, \delta, q_0)$  is a Turing machine with a non-deterministic transition function  $\delta: Q \times \Gamma \rightarrow 2^{Q \times \Gamma \times \{L, R\}}$  whose set of states is partitioned into **existential** and **universal** states:

$Q_\exists$ : set of existential states

$Q_\forall$ : set of universal states

- Configurations of ATMs are the same as for (N)TMs:  
tape(s) + state + head position
- A configuration can be **universal** or **existential**, depending on whether its state is universal or existential
- Possible transitions between configurations are defined as for NTMs

## Nondeterminism and Parallelism

ATMs can be seen as a **generalisation of non-deterministic TMs**:

An NTM is an ATM where all states are existential (besides the single accepting state, which is always universal according to our definition).

ATMs can be seen as a **model of parallel computation**:

In every step, **fork** the current process to create sub-processes that explore each possible transition in parallel

- for universal states, combine the results of sub-processes with AND
- for existential states, combine the results of sub-processes with OR

**Alternative view:** an ATM accepts if its computation tree, considered as an AND-OR tree, evaluates to **true**

## Example: Alternating Algorithm for MinFormula

### MinFORMULA

Input: A propositional formula  $\varphi$ .  
Problem: Is  $\varphi$  the shortest formula that is satisfied by the same assignments as  $\varphi$ ?

MinFORMULA can be solved by an alternating algorithm:

```
01 MinFORMULA(formula  $\varphi$ ) :  
02 universally choose  $\psi :=$  formula shorter than  $\varphi$   
03 existentially guess  $\mathcal{I} :=$  assignment for variables in  $\varphi$   
04 if  $\varphi^{\mathcal{I}} = \psi^{\mathcal{I}}$  :  
05   return false  
06 else :  
07   return true
```

## Example: Alternating Algorithm for Geography

Recall the **GEOGRAPHY** game discussed in Lecture 10:

```
01 ALTGEOGRAPHY(directed graph  $G$ , start node  $s$ ) :  
02 Visited := { $s$ } // visited nodes  
03 cur :=  $s$  // current node  
04 while true :  
05   // existential move:  
06   if all successors of cur are in Visited:  
07     return false  
08   existentially guess cur := unvisited successor of cur  
09   Visited := Visited  $\cup$  {cur}  
10   // universal move:  
11   if all successors of cur are in Visited:  
12     return true  
13   universally choose cur := unvisited successor of cur  
14   Visited := Visited  $\cup$  {cur}
```

## Time and Space Bounded ATMs

As before, time and space bounds apply to any computation path in the computation tree.

**Definition 16.3:** Let  $\mathcal{M}$  be an alternating Turing machine and let  $f : \mathbb{N} \rightarrow \mathbb{R}^+$  be a function.

- (1)  $\mathcal{M}$  is  $f$ -time bounded if it halts on every input  $w \in \Sigma^*$  and on every computation path after  $\leq f(|w|)$  steps.
- (2)  $\mathcal{M}$  is  $f$ -space bounded if it halts on every input  $w \in \Sigma^*$  and on every computation path using  $\leq f(|w|)$  cells on its tapes.

(Here we typically assume that Turing machines have a separate input tape that we do not count in measuring space complexity.)

## Defining Alternating Complexity Classes

**Definition 16.4:** Let  $f : \mathbb{N} \rightarrow \mathbb{R}^+$  be a function.

- (1)  $\text{ATime}(f(n))$  is the class of all languages  $\mathbf{L}$  for which there is an  $O(f(n))$ -time bounded alternating Turing machine deciding  $\mathbf{L}$ .
- (2)  $\text{ASpace}(f(n))$  is the class of all languages  $\mathbf{L}$  for which there is an  $O(f(n))$ -space bounded alternating Turing machine deciding  $\mathbf{L}$ .

## Common Alternating Complexity Classes

$AP = APTIME = \bigcup_{d \geq 1} ATIME(n^d)$  alternating polynomial time

$AExp = AExpTime = \bigcup_{d \geq 1} ATIME(2^{n^d})$  alternating exponential time

$A2Exp = A2ExpTime = \bigcup_{d \geq 1} ATIME(2^{2^{n^d}})$  alt. double-exponential time

$AL = ALogSpace = ASpace(\log n)$  alternating logarithmic space

$APSpace = \bigcup_{d \geq 1} ASpace(n^d)$  alternating polynomial space

$AExpSpace = \bigcup_{d \geq 1} ASpace(2^{n^d})$  alternating exponential space

**Example 16.5:**  $\mathbf{GEOGRAPHY} \in APTIME$ .

## Alternating Complexity Classes: Basic Properties

### Nondeterminism:

ATMs can do everything that the corresponding NTMs can do, e.g.,  $NP \subseteq APTIME$

**Reductions:** Polynomial many-one reductions can be used to show membership in many alternating complexity classes, e.g., if  $L \in APTIME$  and  $L' \leq_p L$  then  $L' \in APTIME$ .

In particular:  $PSPACE \subseteq APTIME$  (since  $\mathbf{GEOGRAPHY} \in APTIME$ )

### Complementation: ATMs are easily complemented:

- Let  $\mathcal{M}$  be an ATM accepting language  $L(\mathcal{M})$
- Let  $\mathcal{M}'$  be obtained from  $\mathcal{M}$  by swapping existential and universal states
- Then  $L(\mathcal{M}') = \overline{L(\mathcal{M})}$

For alternating algorithms this means: (1) negate all return values, (2) swap universal and existential branching points

## Example: Complement of $\mathbf{MINFORMULA}$

Original algorithm:

```
01 MINFORMULA(formula  $\varphi$ ) :
02   universally choose  $\psi :=$  formula shorter than  $\varphi$ 
03   existentially guess  $I :=$  assignment for variables in  $\varphi$ 
04   if  $\varphi^I = \psi^I$  :
05     return false
06   else :
07     return true
```

Complemented algorithm:

```
01 COMPLMINFORMULA(formula  $\varphi$ ) :
02   existentially guess  $\psi :=$  formula shorter than  $\varphi$ 
03   universally choose  $I :=$  assignment for variables in  $\varphi$ 
04   if  $\varphi^I = \psi^I$  :
05     return true
06   else :
07     return false
```

## Alternating Time vs. Deterministic Space

## From Alternating Time to Deterministic Space

**Theorem 16.6:** For  $f(n) \geq n$ , we have  $\text{ATime}(f) \subseteq \text{DSpace}(f^2)$ .

**Proof:** We simulate an ATM  $\mathcal{M}$  using a TM  $\mathcal{S}$ :

- $\mathcal{S}$  performs a depth-first search of the configuration tree of  $\mathcal{M}$
- The acceptance status of each node is computed recursively (similar to typical PSpace algorithms we have seen before)
- $\mathcal{M}$  accepts exactly if the root of the configuration tree is accepting

The maximum recursion depth is  $f(n)$ . The maximum size of a configuration is  $O(f(n))$ . Hence the claim follows.  $\square$

**Note:** The result can be strengthened to  $\text{ATime}(f) \subseteq \text{DSpace}(f)$  by not storing the whole configuration. See [Sipser, Lemma 10.22].

## From Nondeterministic Space to Alternating Time

**Theorem 16.7:** For  $f(n) \geq n$ , we have  $\text{NSpace}(f) \subseteq \text{ATime}(f^2)$ .

**Proof:** We simulate an NTM  $\mathcal{M}$  using an ATM  $\mathcal{S}$ .

**Challenge:** the computing paths of  $\mathcal{M}$  might be up to  $2^{df(n)}$  in length.

**Solution:** recursively solve Yieldability problems, as in Savitch's Theorem:

- We want to check if  $\mathcal{M}$  can go from configuration  $C_1$  to  $C_2$  in at most  $k$  steps
- To do this, existentially guess an intermediate configuration  $C'$ .
- Universally check if  $\mathcal{M}$  can go from  $C_1$  to  $C'$  in  $k/2$  steps, and from  $C'$  to  $C_2$  in  $k/2$  steps.

Storing one intermediate configuration  $C'$  takes space  $O(f(n))$ . Maximal recursion depth is  $O(f(n))$ . Hence the result follows.  $\square$

## Harvest: Alternating Time = Deterministic Space

For  $f(n) \geq n$ , we have shown

$$\text{ATime}(f) \subseteq \text{DSpace}(f^2) \quad \text{and} \quad \text{DSpace}(f) \subseteq \text{NSpace}(f) \subseteq \text{ATime}(f^2).$$

The quadratic increase is swallowed by (super)polynomial bounds:

**Corollary 16.8 (“Alternating Time = Deterministic Space”):**  $\text{APTime} = \text{PSpace}$  and  $\text{AExpTime} = \text{ExpSpace}$ .

**Proof:**

- $\text{ATime}(n^d) \subseteq \text{DSpace}(n^{2d}) \subseteq \text{PSpace}$   
 $\text{DSpace}(n^d) \subseteq \text{NSpace}(n^d) \subseteq \text{ATime}(n^{2d}) \subseteq \text{APTime}$
- Second claim is left as an exercise  $\square$

One can also read this as “Parallel Time = Sequential Space.”

## Alternating Space vs. Deterministic Time

## From Alternating Space to Deterministic Time

In this direction, the increase is exponential:

**Theorem 16.9:** For  $f(n) \geq \log n$ , we have  $\text{ASpace}(f) \subseteq \text{DTime}(2^{O(f)})$ .

**Proof:** The proof is similar to the exponential deterministic simulation of space-bounded NTMs in Lecture 9 (Theorem 9.7):

- Construct configuration graph of ATM
- Iteratively compute acceptance status of each configuration
- Check if starting configuration is accepting

Each step can be done in exponential time (in particular, computing the acceptance condition in each step is no more difficult than for plain NTMs).  $\square$

## From Deterministic Time To Alternating Space

The exponential blow-up can be reversed when going back to ATMs:

**Theorem 16.10:**

If  $f(n) \geq \log n$  is space-constructible, then  $\text{DTime}(2^{O(f)}) \subseteq \text{ASpace}(f)$ .

**Proof:** We show: for any  $g(n) \geq n$ , we have  $\text{DTime}(g) \subseteq \text{ASpace}(\log g)$ .

We simulate a TM  $\mathcal{M}$  using an ATM  $\mathcal{S}$ . This is not so easy:

- A computation of  $\mathcal{M}$  is exponentially longer than the space available to  $\mathcal{S}$   
 $\leadsto$  we solved this before with Yieldability
- A configuration of  $\mathcal{M}$  is exponentially longer than the space available to  $\mathcal{S}$   
 $\leadsto$  this is more tricky . . .

There is a coarse proof sketch in [Sipser, Lemma 10.25]. We follow a more detailed proof from the lecture notes of Erich Grädel [Complexity Theory, WS 2009/10] (link).

## From Deterministic Time To Alternating Space (2)

**Notation:** The proof is easier if we write a configuration  $\sigma_1 \cdots \sigma_{i-1} q \sigma_i \sigma_{i+1} \cdots \sigma_m$  as a sequence

$$* \sigma_1 \cdots \sigma_{i-1} \langle q, \sigma_i \rangle \sigma_{i+1} \cdots \sigma_m *$$

of symbols from the set  $\Omega = \{*\} \cup \Gamma \cup (Q \times \Gamma)$ .

Then the  $\Omega$ -symbol (state and tape) at position  $i$  follows deterministically from the  $\Omega$ -symbols at positions  $i-1$ ,  $i$ , and  $i+1$  in the previous step.

We write  $\mathcal{M}(\omega_{i-1}, \omega_i, \omega_{i+1})$  for this symbol.

**Proof idea:**

- Only store a pointer to one cell in one configuration of  $\mathcal{M}$
- Verify the contents of current cell  $i$  in step  $j$  by guessing the previous cell contents  $\omega_{i-1}, \omega_i, \omega_{i+1}$  in step  $j$ .
- Check iteratively that the guessed symbols are correct

## From Deterministic Time To Alternating Space (3)

Let  $h : \mathbb{N} \rightarrow \mathbb{R}$  be a space-constructible function in  $O(g)$  that defines the exact time bound for  $\mathcal{M}$  (no  $O$ -notation).

```
01 ATMSIMULATETM(TM  $\mathcal{M}$ , input word  $w$ , time bound  $h$ ) :
02 existentially guess  $s \leq h(|w|)$  // halting step
03 existentially guess  $i \in \{0, \dots, s\}$  // halting position
04 existentially guess  $\omega \in Q \times \Gamma$  // halting cell + state
05 if  $\mathcal{M}$  would not halt in  $\omega$  :
06     return false
07 for  $j = s, \dots, 1$  do :
08     existentially guess  $\langle \omega_{-1}, \omega_0, \omega_1 \rangle \in \Omega^3$ 
09     if  $\mathcal{M}(\omega_{-1}, \omega_0, \omega_1) \neq \omega$  :
10         return false
11     universally choose  $\ell \in \{-1, 0, 1\}$ 
12      $\omega := \omega_\ell$ 
13      $i := i + \ell$ 
14 // after tracing back  $s$  steps, check input configuration:
15 return "input configuration of  $\mathcal{M}$  on  $w$  has  $\omega$  at position  $i$ "
```

## Summary and Outlook

For  $f(n) \geq \log n$ , we have shown  $\text{ASpace}(f) = \text{DTime}(2^{O(f)})$ .

**Corollary 16.11 (“Alternating Space = Exponential Deterministic Time”):**  
 $\text{AL} = \text{P}$  and  $\text{APSpace} = \text{ExpTime}$ .

We can sum up our findings as follows:

$$\begin{array}{ccccccccccc} \text{L} & \subseteq & \text{PTime} & \subseteq & \text{PSpace} & \subseteq & \text{ExpTime} & \subseteq & \text{ExpSpace} & & \\ & & \parallel & & \parallel & & \parallel & & \parallel & & \\ & & \text{ALogSpace} & \subseteq & \text{APTime} & \subseteq & \text{APSpace} & \subseteq & \text{AExpTime} & & \end{array}$$

### What's next?

- Alternation as a resource that can be bounded
- A hierarchy between NP and PSpace
- End-of-year consultation