

## 13. Alternation

Goal: Study a new resource: alternation

- Useful in understanding the relationship among complexity classes, more generally, between time and space.
- Useful to classify problems according to their complexity (alternation simplifies some proofs).

Idea: Alternation generalizes non-determinism.

computation tree ← • If non-deterministic algorithm branches a process into multiple sub-processes / child processes.

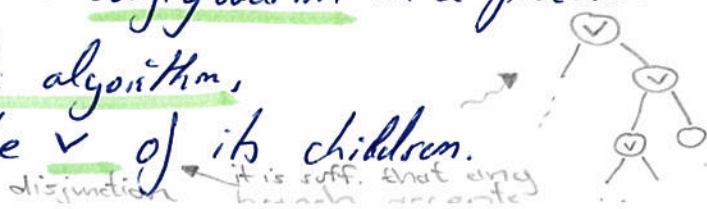
one is sufficient ← • It accepts, if any of the child processes accept.

• An alternating algorithm also branches a process into child processes, but additionally specifies the mode of acceptance:

- ↳ the current process should accept, if any of its children accept,
- ↳ or it should accept, if all the children accept.

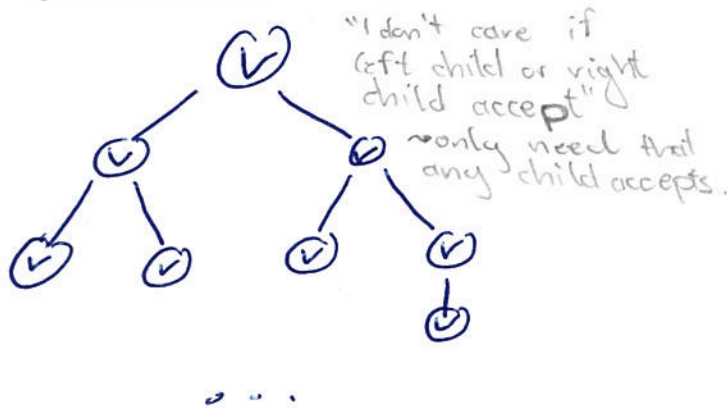
Illustration: Consider non-deterministic and alternating algorithms as (computation) trees

- that represent the branching structure of the spawned processes.
- Each node represents the configuration in a process.
- ↳ In a non-deterministic algorithm, each node computes the ∨ of its children.

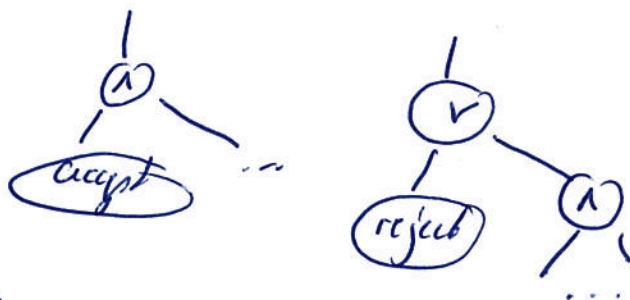
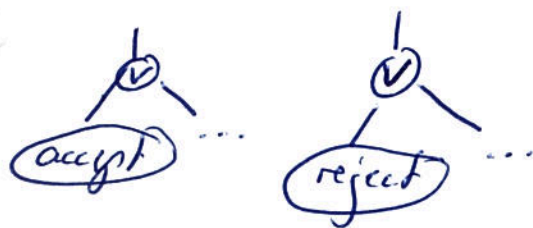
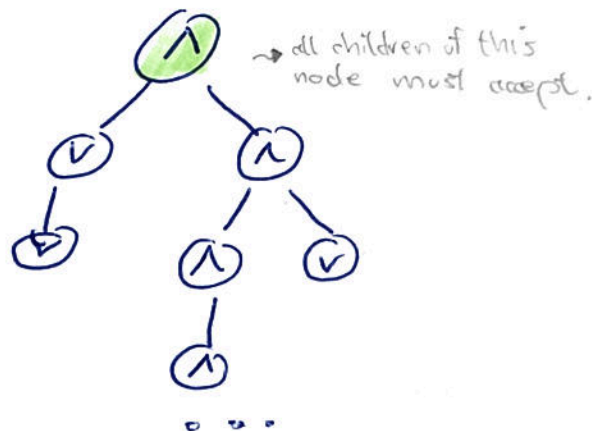


↳ In an alternating algorithm,  
the nodes compute the  $\wedge$  or the  $\vee$ ,  
as determined by the algorithm.

Non-determinism:



Alternation:



Goal: Study the relationship of alternating time and space classes  
to the known complexity classes.

## 13.1 Alternating Time and Space

Definition:

• An alternating Turing machine (ATM)

is a non-deterministic Turing machine

with an additional feature:

the set of states  $Q$  is divided

into universal states and existential states.

• The semantics of an ATM is defined via its computation tree  
(that makes explicit all branching).

to do so:  
need notion of an  
alternating TM

We label each configuration of the computation tree

by  $\wedge$  or  $\vee$ ,

depending on whether the ATM is  $\bullet$  in an universal state  $\rightarrow \wedge$   
 $\circ$  or in an existential state  $\rightarrow \vee$

a config. in an univ. state  
Then  $\wedge$ -configuration is accepting,  
if all its children are accepting.

a config. in an exist. state  
Then  $\vee$ -configuration is accepting,  
if some child is accepting.

The input is accepted, if the initial configuration  
is found to be accepting.

Note that an ATM does not need to make explicit  
accept and reject states:

$\hookrightarrow$  Then  $\vee$ -configuration without children is rejecting

$\hookrightarrow$  Then  $\wedge$ -configuration without children is accepting.

it has no child  
which could acc.

all children  
accept

Note:

There are alternative definitions of ATMs in the literature  
that do contain  $\rightarrow$  as a labelling of states.

$\rightarrow$  But this model  
does not have  
more power

( $\rightarrow$ -configurations are supposed to have a single successor).

Lemma:

Every ATM  $M$  with rejection can be converted  
into an ATM  $M'$  without rejection  
without overhead in time and space.

Proof (Sketch):

Remember from logic the concept of dualizing a formula  $F$

by turning  $v$  to  $w$  and  $w$  to  $v$ .

Then, if we also invert the valuation  $\ell$

to  $\ell(a) := 1 - \ell(a)$ , for any var.

we obtain

$$\ell(\text{dual}(F)) = 1 - \ell(F).$$

• With the same trick ( $v \rightarrow w, w \rightarrow v$ ), we can dualize the TTM  $M$  into  $M_d$ . also has negation

• We moreover replace every negation  $\neg$  in  $M$  by  $v$

• Then  $M'$  is the union of  $M$  and  $M_d$ . in  $M_d$  by  $v$ .

Upon every former negation,  $M$  moves to  $M_d$  (by  $v$ ) if we see a neg. and are in  $M \rightarrow$  goto  $M_d$  and  $M_d$  moves to  $M$  (by  $w$ ). □

The time and space requirements of TTMs

• defined like for NTMs.

↳ An TTM is  $t(n)$ -time-bounded,

if all paths in the computation tree are of length  $\leq t(n)$ .

↳ An TTM is  $s(n)$ -space-bounded,

if all paths in the computation tree

use at most  $s(n)$  work tape cells (per configuration).

Definition: Let  $s, t: \mathbb{N} \rightarrow \mathbb{N}$ .

↳  $\text{TTIME}(t(n)) := \{L(M) \mid M \text{ is a } t(n)\text{-time-bounded TTM (that is a decider)}\}$

• ASPACE( $s(n)$ ) :=  $\{L(M) \mid M \text{ is an } s(n)\text{-space-bounded NTM (that is a decider)}\}$  §

There are alternating analogues of the robot complexity domes:

$$\underline{AL} := \text{ASPACE}(\mathcal{O}(\log n))$$

$$\underline{AP} := \bigcup_{k \in \mathbb{N}} \text{ATIME}(\mathcal{O}(n^k))$$

$$\underline{APSPACE} := \bigcup_{k \in \mathbb{N}} \text{ASPACE}(\mathcal{O}(n^k))$$

$$\underline{AEXP} := \bigcup_{k \in \mathbb{N}} \text{ATIME}(2^{\mathcal{O}(n^k)})$$

### 13.2 From Alternating TIME to Deterministic Space

Goal: • Show that, up to a polynomial, alternating time coincides with deterministic space.

• We add  $\mathcal{O}$  for not having to reason about compression and speed-up for NTMs.

Theorem: Let  $f(n) \gg n$ .

$$\underline{\text{ATIME}(f(n))} \stackrel{(1)}{\subseteq} \underline{\text{DSPACE}(f(n))} \stackrel{(2)}{\subseteq} \underline{\text{ATIME}(\mathcal{O}(f(n)^2))}$$

In particular, AP = PSPACE.

$$\text{AP} = \bigcup_{f \text{ poly}} \text{ATIME}(\mathcal{O}(n^f)) \subseteq \underbrace{\bigcup_{f \text{ poly}} \text{DSPACE}(\mathcal{O}(n^f))}_{\text{PSPACE}} \subseteq \bigcup_{f \text{ poly}} \text{ATIME}(\mathcal{O}(n^{2f})) \subseteq \text{AP}$$

Proof:

(1): We convert a  $f(n)$ -time-bounded NTM  $M$  into ~~an~~-space-bounded DTM  $M'$  s.t.h.  $L(M) = L(M')$ , a  $f(n)$

• On input  $x$ , the DTM  $M'$  performs a depth-first search

-5. on  $M$ 's computation tree,

to determine which configurations are accepting.

It accepts, if the root node (start configuration of  $M$ ) is found to be accepting.

We have seen a similar construction

when showing  $NTIME(t(n)) \subseteq DSPACE(t(n))$ .

↳ Initially, it seems we need a  $t(n)$ -high stack with entries that are configurations of length  $t(n)$ , hence we need  $O(t(n)^2)$  space.

↳ Then we find that we only have to store the non-deterministic choices.  $\leadsto O(t(n))$  space

(Here, it does not matter whether we have  $O(t(n))$  space, because for regular DTMs we have compression).

(2): We actually show the stronger inclusion

$$DSPACE(t(n)) \subseteq NSPACE(t(n)) \subseteq FTIME(O(t(n)^2)).$$

○ The idea is to use a parallel implementation of Savitch's procedure  $sav(\alpha, \beta, k)$

that determines whether configuration  $\alpha$  can go to  $\beta$  in  $\leq k$  steps.

↳ if  $k=0$  or  $k=1$ , we directly check  $\alpha \rightarrow \beta$  or  $\alpha \rightarrow \beta$ .

↳ if  $k \geq 2$ ,

we guess  $\gamma \in \Delta^{s(n)}$  using  $v$ -branching

Note that the guessing cannot be done instantaneously, but needs  $s(n)$  steps.

$\rightarrow$  we compute  $t(n)$  configurations.

one path should be accepting

paths in tree are of length  $t(n)$

we store  $\gamma$  if the other children are acc.  $|S| \cdot t(n) \in O(t(n))$

how many conf. can a NTM  $\rightarrow$  which is  $t$ -time bounded visit?  $\rightarrow s(n) = 2^{t(n)}$

The result is  $|A|^{s(n)}$  independent sub-processes, one for each  $\gamma$ .



For each  $\gamma$ , we check in parallel

$\text{sav}(\alpha, \gamma, \lceil \frac{s(n)}{2} \rceil)$  and  $\text{sav}(\gamma, \beta, \lfloor \frac{s(n)}{2} \rfloor)$

Using  $\alpha$ -branching.  $\rightarrow$  both parts of path must accept. paths in comp. tree of length  $O(s(n)^2)$ .

This (alternating) procedure can be implemented in  $O(\sqrt{s(n)})^2$  alternating time.  $\square$

### 13.3 From Alternating Space to Deterministic Time

Goal: Show that alternating space coincides with exponentially more deterministic time.

Theorem: Let  $s(n) \geq \log n$ .

$$\text{ASPACE}(O(s(n))) = \text{DTIME}(2^{O(s(n))})$$

$$\begin{aligned} \text{ASPACE} &= \bigcup_{s(n)} \text{ASPACE}(O(s(n))) \\ &= \bigcup_{s(n)} \text{DTIME}(2^{O(s(n))}) \\ &= \text{EXP} \end{aligned}$$

In particular,  $\text{AL} = \text{P}$  and  $\text{ASPACE} = \text{EXP}$ .

Proof:  $\text{AL} = \text{ASPACE}(O(\log n)) = \text{DTIME}(2^{O(\log n)}) = \bigcup_{k} \text{DTIME}(O(n^k)) = \text{P}$

$\subseteq$ : We construct a  $2^{O(s(n))}$ -time-bounded DTM  $M'$  that simulates an ATM  $M$  that is  $O(s(n))$ -space bounded.

- On input  $x$ ,  $M'$  constructs the configuration graph of  $M$  on  $x$ 
  - $\hookrightarrow$  the nodes are the configurations of  $M$  on  $x$  that use at most  $d \cdot s(n)$  space ( $d$  appropriate for  $O(s(n))$ );
  - $\hookrightarrow$  Edges go from a configuration  $\alpha$  to a configuration  $\beta$ , if  $\alpha$  can go to  $\beta$  in a single step.

- After constructing the graph,  $M'$  repeatedly scans it to mark configurations as accepting:
  - ↳ Initially, only the  $n$ -configurations without successors are marked. these are accepting
  - ↳ An  $n$ -configuration with only accepting successors is marked accepting.
  - ↳ An  $v$ -configuration with some accepting successor is also marked accepting.

•  $M'$  accepts, if the start configuration is marked. It rejects, if no more new nodes are marked (fixed point).

• The configuration graph can be constructed in  $2^{O(n \log n)}$  time. Scanning each step costs  $2^{O(n \log n)}$  time. There are at most  $2^{O(n \log n)}$  scans. Hence, the time bound is

$$\underbrace{2^{O(n \log n)} \cdot 2^{O(n \log n)}}_{\text{scanning } 2^{O(n \log n)} \text{ times}} + \underbrace{2^{O(n \log n)}}_{\text{constructing}} = 2^{O(n \log n)}$$

≥ We show how to simulate a  $2^{O(n \log n)}$ -time-bounded DTM  $M$  by an RTM  $M'$  with  $O(n \log n)$  space.

The simulation is tricky because the space available to  $M'$  is so much less than  $M$ 's compute time.

Essentially, the space is just sufficient to store

a pointer into the computation matrix of  $M$

from the Ladner-Cook-Levin theorem.



Indeed, we will build on Ladner's construction and show how to construct and evaluate on-the-fly

a CVP-instance of length  $2^{O(s(n))}$ .

Machine  $M'$  recursively guesses and verifies the values of the variables  $P_{i,j}^a$  and  $Q_{i,j}^p$

In step  $i$ , we see  $a$  in cell  $j$   
 In step  $i$ , the TM's head is in cell  $j$  and the TM is in state  $p$

To verify the content of a cell  $i,j$  with  $c > 0$ :

↳  $M'$  existentially guesses the values of the parent cells (for computation step  $i-1$ ),

↳ Checks whether the guessed content would yield the content of cell  $i,j$

according to  $M$ 's transition relation, and

$$(P_{i,j}^a = 0 \wedge Q_{i,j}^p = 1 \dots 1)$$

$\uparrow$   
 $\vee(\dots)$   
 $\vee(\dots)$

↳ universally branches to recursively verify these guesses.

For a cell in the first row ( $i=0$ ),

$M'$  can verify the guess directly,

because it knows  $M$ 's start configuration.

We can assume  $M$  has a single accept configuration where the head is on the left-end marker of the tape.

Hence,  $M'$  can check whether  $M$  accepts  $x$

by checking the value of the lowest leftmost cell of the computation matrix.



Since  $M'$  never needs to store more than a single pointer to a variable in the matrix,

and since such pointers can be stored in binary,

$M'$  needs  $\log_2 2^{O(s(n))} = O(s(n))$  space.

□