Other Models of Computation

1

Models of computation:

- •Turing Machines
- Recursive Functions
- Post Systems
- Rewriting Systems

Church's Thesis:

All models of computation are equivalent

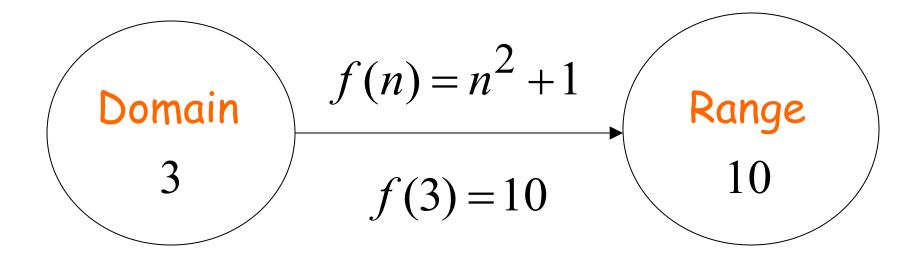
Turing's Thesis:

A computation is mechanical if and only if it can be performed by a Turing Machine Church's and Turing's Thesis are similar:

Church-Turing Thesis

Recursive Functions

An example function:



We need a way to define functions

We need a set of basic functions

Basic Primitive Recursive Functions

Zero function: z(x) = 0

Successor function: s(x) = x + 1

Projection functions: $p_1(x_1, x_2) = x_1$

 $p_2(x_1, x_2) = x_2$

Building complicated functions:

Composition:
$$f(x,y) = h(g_1(x,y),g_2(x,y))$$

Primitive Recursion:

$$f(x,0) = g_1(x)$$

 $f(x, y+1) = h(g_2(x, y), f(x, y))$

Any function built from the basic primitive recursive functions is called:

Primitive Recursive Function

A Primitive Recursive Function: add(x, y)

$$add(x,0) = x$$
 (projection)

add(x, y+1) = s(add(x, y))

(successor function)

add(3,2) = s(add(3,1))= s(s(add(3,0)))= s(s(3))= s(4)= 5

Another Primitive Recursive Function: mult(x, y)

mult(x,0) = 0

mult(x, y+1) = add(x, mult(x, y))

Theorem:

The set of primitive recursive functions is countable

Proof:

Each primitive recursive function can be encoded as a string

Enumerate all strings in proper order

Check if a string is a function

Theorem

there is a function that is not primitive recursive

Proof: Enumerate the primitive recursive functions

 f_1, f_2, f_3, K

Define function $g(i) = f_i(i) + 1$

g differs from every f_i

g is not primitive recursive

END OF PROOF

A specific function that is <u>not</u> Primitive Recursive:

Ackermann's function:

$$A(0, y) = y + 1$$

$$A(x, 0) = A(x - 1, 1)$$

$$A(x, y + 1) = A(x - 1, A(x, y))$$

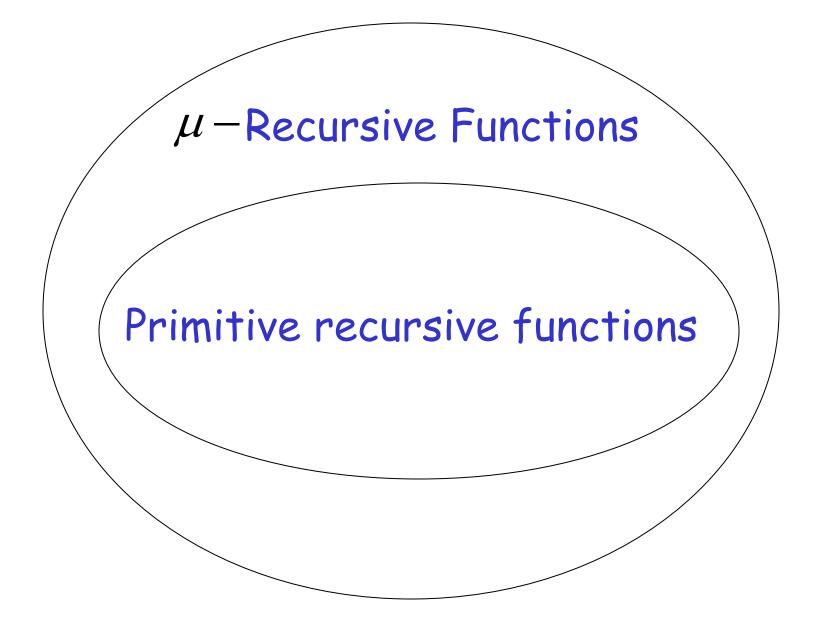
Grows very fast,

faster than any primitive recursive function

μ – Recursive Functions

$\mu y(g(x, y)) = \text{smallest } y \text{ such that } g(x, y) = 0$

Ackerman's function is a μ -Recursive Function



Post Systems

• Have Axioms

Have Productions

Very similar with unrestricted grammars

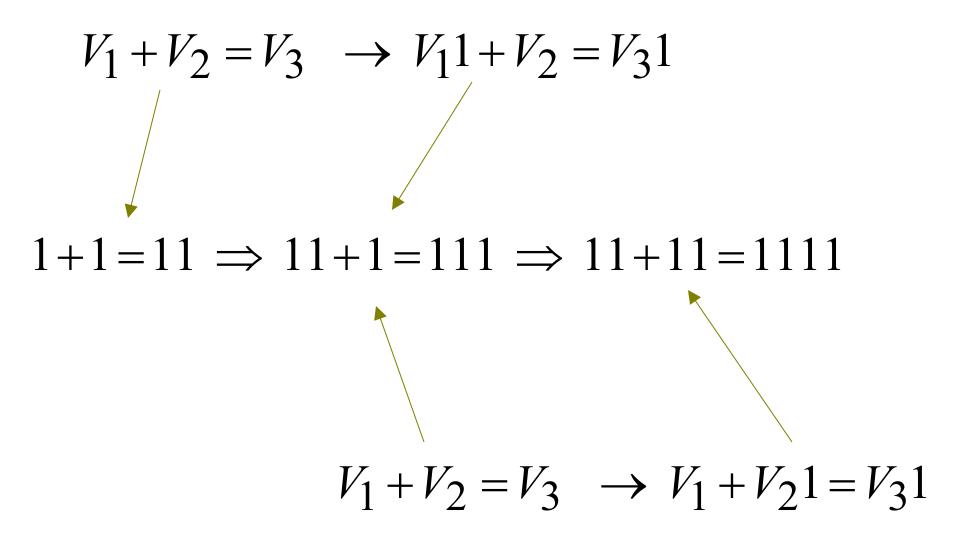
Example: Unary Addition

Axiom: 1+1=11

Productions:

$V_1 + V_2 = V_3 \rightarrow V_1 1 + V_2 = V_3 1$ $V_1 + V_2 = V_3 \rightarrow V_1 + V_2 1 = V_3 1$

A production:



Post systems are good for proving mathematical statements from a set of Axioms

Theorem:

A language is recursively enumerable if and only if a Post system generates it

Rewriting Systems

They convert one string to another

Matrix Grammars

- Markov Algorithms
- Lindenmayer-Systems

Very similar to unrestricted grammars

Matrix Grammars

Example:
$$P_1: S \rightarrow S_1S_2$$

 $P_2: S_1 \rightarrow aS_1, S_2 \rightarrow bS_2c$
 $P_3: S_1 \rightarrow \lambda, S_2 \rightarrow \lambda$

Derivation:

 $S \Rightarrow S_1S_2 \Rightarrow aS_1bS_2c \Rightarrow aaS_1bbS_2cc \Rightarrow aabbcc$

A set of productions is applied simultaneously

$$P_{1}: S \to S_{1}S_{2}$$

$$P_{2}: S_{1} \to aS_{1}, S_{2} \to bS_{2}c$$

$$P_{3}: S_{1} \to \lambda, S_{2} \to \lambda$$

$$L = \{a^n b^n c^n : n \ge 0\}$$

Theorem: A language is recursively enumerable if and only if a Matrix grammar generates it

Markov Algorithms

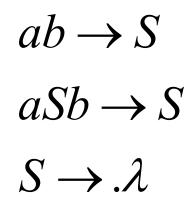
Grammars that produce λ

Example: $ab \rightarrow S$ $aSb \rightarrow S$

$$S \rightarrow .\lambda$$

Derivation:

$aaabbb \Rightarrow aaSbb \Rightarrow aSb \Rightarrow S \Rightarrow \lambda$



 $L = \{a^n b^n : n \ge 0\}$

In general:
$$L = \{w: w \Rightarrow \lambda\}$$

Theorem:

A language is recursively enumerable if and only if a Markov algorithm generates it

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Lindenmayer-Systems

They are parallel rewriting systems

Example: $a \rightarrow aa$

Derivation: $a \Rightarrow aa \Rightarrow aaaa \Rightarrow aaaaaaaa$

$$L = \{a^{2^n} : n \ge 0\}$$

Lindenmayer-Systems are not general As recursively enumerable languages

Extended Lindenmayer-Systems: $(x, a, y) \rightarrow u$

Theorem:

A language is recursively enumerable if and only if an Extended Lindenmayer-System generates it

Computational Complexity

Time Complexity:

The number of steps during a computation

Space Complexity:

Space used during a computation

Time Complexity

•We use a multitape Turing machine

•We count the number of steps until a string is accepted

•We use the O(k) notation

Example:
$$L = \{a^n b^n : n \ge 0\}$$

Algorithm to accept a string w:

Use a two-tape Turing machine

•Copy the a on the second tape

•Compare the a and b

$$L = \{a^n b^n : n \ge 0\}$$

Time needed:

•Copy the a on the second tape O(|w|)

•Compare the a and b O(|w|)

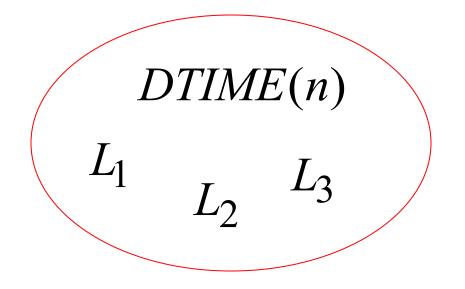
Total time: O(|w|)

 $L = \{a^n b^n : n \ge 0\}$

For string of length n

time needed for acceptance: O(n)

Language class: DTIME(n)



A Deterministic Turing Machine accepts each string of length nin time O(n)

DTIME(n) $\{a^n b^n : n \ge 0\}$ $\{WW\}$

In a similar way we define the class

DTIME(T(n))

for any time function: T(n)

Examples:

 $DTIME(n^2), DTIME(n^3),...$

Example: The membership problem for context free languages

 $L = \{w : w \text{ is generated by grammar } G\}$

$$L \in DTIME(n^3)$$
 (CYK - algorithm)

Polynomial time

Theorem: $DTIME(n^{k+1}) \subset DTIME(n^k)$

 $DTIME(n^{k+1})$ $DTIME(n^k)$

Polynomial time algorithms: $DTIME(n^k)$

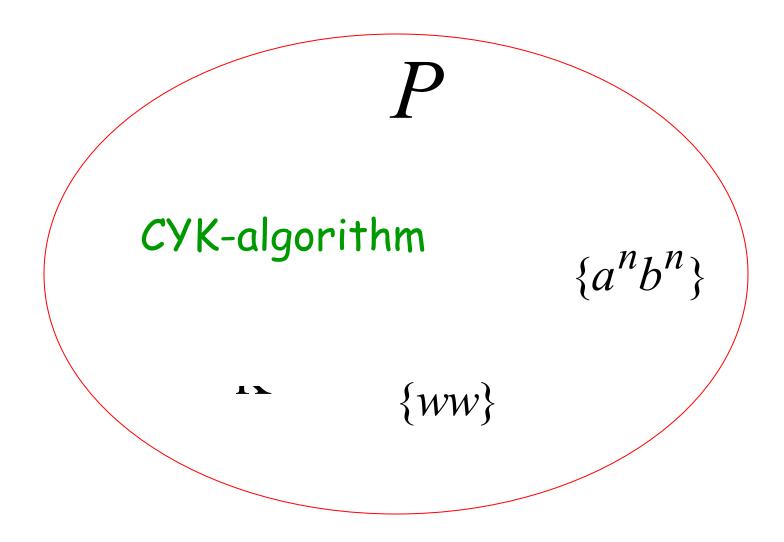
Represent tractable algorithms: For small k we can compute the result fast

The class P

$P = \cup DTIME(n^k) \quad \text{for all} \quad k$

Polynomial time

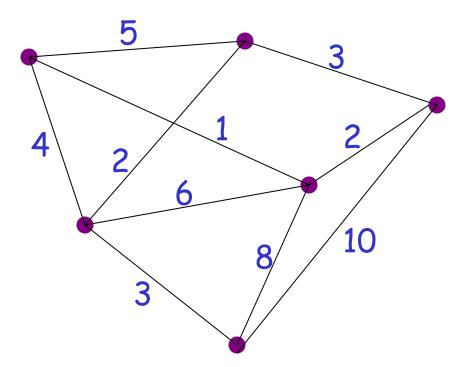
•All tractable problems



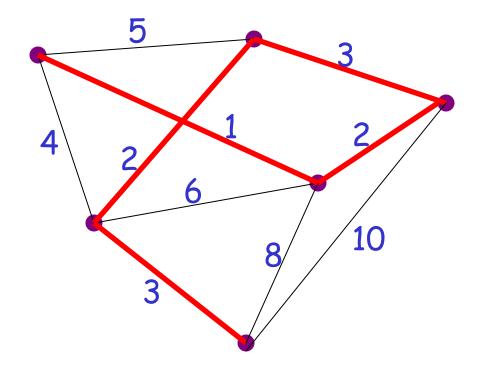
Exponential time algorithms: $DTIME(2^n)$

Represent intractable algorithms: Some problem instances may take centuries to solve

Example: the Traveling Salesperson Problem



Question: what is the shortest route that connects all cities?



Question: what is the shortest route that connects all cities?

A solution: search exhuastively all hamiltonian paths

L = {shortest hamiltonian paths}

 $L \in DTIME(n!) \approx DTIME(2^n)$

Exponential time

Intractable problem

Example: The Satisfiability Problem

Boolean expressions in Conjunctive Normal Form:

 $t_1 \wedge t_2 \wedge t_3 \wedge \Lambda \wedge t_k$

 $t_i = x_1 \lor \overline{x}_2 \lor x_3 \lor \Lambda \lor \overline{x}_p$ Variables

Question: is expression satisfiable?

Example:
$$(\overline{x}_1 \lor x_2) \land (x_1 \lor x_3)$$

Satisfiable:

 $x_1 = 0, x_2 = 1, x_3 = 1$

 $(\overline{x}_1 \lor x_2) \land (x_1 \lor x_3) = 1$

Example: $(x_1 \lor x_2) \land \overline{x}_1 \land \overline{x}_2$

Not satisfiable

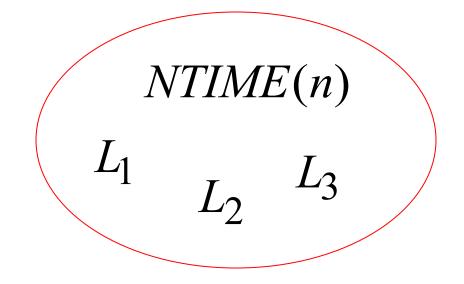
 $L = \{w: expression w \text{ is satisfiable}\}$

For *n* variables: $L \in DTIME(2^n)$ exponential

Algorithm: search exhaustively all the possible binary values of the variables

Non-Determinism

Language class: NTIME(n)



A Non-Deterministic Turing Machine accepts each string of length nin time O(n)

Example: $L = \{ww\}$

Non-Deterministic Algorithm to accept a string ww:

Use a two-tape Turing machine

•Guess the middle of the string and copy w on the second tape

•Compare the two tapes

 $L = \{ww\}$

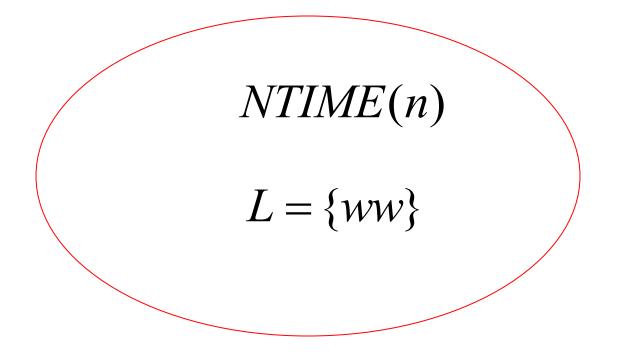
Time needed:

·Use a two-tape Turing machine

•Guess the middle of the string O(|w|)and copy w on the second tape

•Compare the two tapes O(|w|)

Total time: O(|w|)



In a similar way we define the class

NTIME(T(n))

for any time function: T(n)

Examples:

$$NTIME(n^2), NTIME(n^3),...$$

Non-Deterministic Polynomial time algorithms:

 $L \in NTIME(n^k)$

The class NP

$P = \bigcup NTIME(n^k) \quad \text{for all} \quad k$

Non-Deterministic Polynomial time

Example: The satisfiability problem

 $L = \{w: expression w \text{ is satisfiable}\}$

Non-Deterministic algorithm:

•Guess an assignment of the variables

Check if this is a satisfying assignment

 $L = \{w: expression w \text{ is satisfiable}\}$

Time for *n* variables:

•Guess an assignment of the variables O(n)

•Check if this is a satisfying assignment O(n)

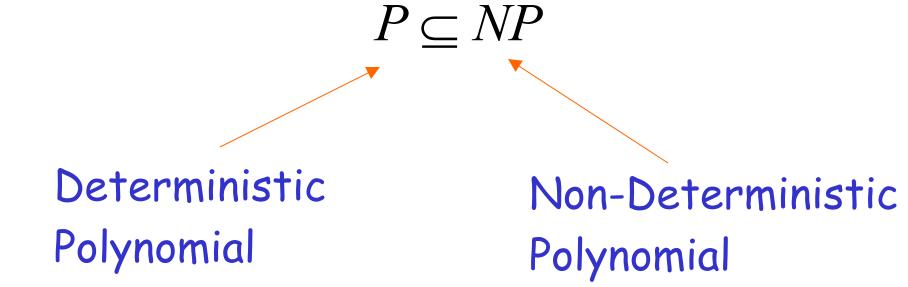
Total time: O(n)

$L = \{w: expression w \text{ is satisfiable}\}$

$L \in NP$

The satisfiability problem is an NP - Problem

Observation:



Open Problem: P = NP?

WE DO NOT KNOW THE ANSWER

Open Problem: P = NP?

Example: Does the Satisfiability problem have a polynomial time deterministic algorithm?

WE DO NOT KNOW THE ANSWER

NP-Completeness

A problem is NP-complete if:

•It is in NP

•Every NP problem is reduced to it (in polynomial time) Observation:

If we can solve any NP-complete problem in Deterministic Polynomial Time (P time) then we know:

$$P = NP$$

Observation:

If we prove that we cannot solve an NP-complete problem in Deterministic Polynomial Time (P time) then we know:

 $P \neq NP$

Cook's Theorem:

The satisfiability problem is NP-complete

Proof:

Convert a Non-Deterministic Turing Machine to a Boolean expression in conjunctive normal form Other NP-Complete Problems:

The Traveling Salesperson Problem

Vertex cover

Hamiltonian Path

All the above are reduced to the satisfiability problem **Observations**:

It is unlikely that NP-complete problems are in P

The NP-complete problems have exponential time algorithms

Approximations of these problems are in P