

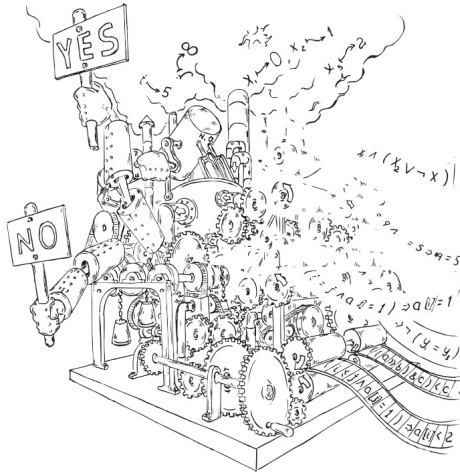
# Automata, Languages and Computation

## Chapter 1 : Automata Theory and Proof Techniques

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Lecture based on material originally developed by :  
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# Theoretical computer science



# Introduction

One of the main goals of theoretical computer science is the mathematical study of **computation**

- computability : what can be computed ?
- tractability : what can be **efficiently** computed ?

The mathematical study of computation requires

- abstract models of machine : **automata theory**
- abstract representations of data : **formal language theory**

# Introduction

Most well-known models of computation :

- Turing machines, introduced for the study of computability
- finite automata, introduced as models of neuronal calculus
- formal grammars, introduced by Noam Chomsky as linguistic models

- 1 Introduction to finite automata : pervasive model using a fixed amount of memory
- 2 Formal proof techniques : hypothesis, thesis, deduction, induction
- 3 Basic concepts of automata theory : alphabets, strings and languages

# Finite automata

**Finite automata**, or FA for short : Finite set of **states** with **transitions** from one state to another

Used as a model for :

- software for digital circuit design
- lexical analyzer within a compiler
- keyword search in a file or on the web
- communication protocols

We will see more later on applications

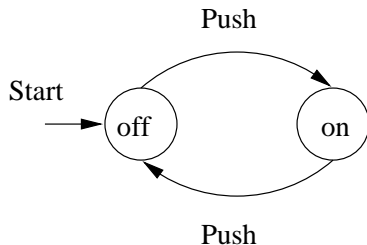
# Finite automata

The simplest representation for an FA is a **graph** :

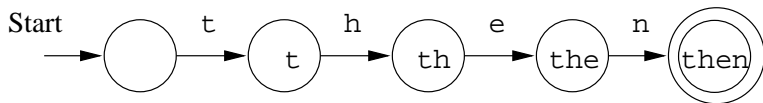
- **nodes** represent states
- **arcs** represent transitions
- **labels** on each arc indicate what is causing the transition

## Example

FA for on/off **switch**



FA that **recognizes** the keyword then in a programming language





# Structural Representation

An FA is a **recognition** model : it takes as input a sequence (string) and either accepts or rejects

Alternatively, we can use a **generative** model : such model generates all of the **desired** sequences (no input)

Recognition models are operational, generative models are declarative

# Structural Representation

**Grammars** : A rewriting rule

$$E \rightarrow E + E$$

specifies that an arithmetic expression may consist of two arithmetic expressions combined by the addition operator

**Regular expressions** : The expression

$$[A-Z] [a-z]^* [ ] [A-Z] [A-Z].$$

generates the string Ithaca NY, but does not generate the string Palo Alto CA

Generative models unveil structure underlying data

## Deductive proof

Typical form of the statement to be proved (H, C properties) :  
*If H, then C*

also written as  $H \Rightarrow C$ , where  $H =$  **hypothesis**,  $C =$  **conclusion**

This means

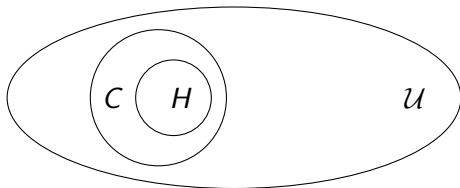
- $H$  is a **sufficient** condition for  $C$
- $C$  is a **necessary** condition for  $H$

See insiemistic interpretation in next slide

## Deductive proof

In an **insiemistic** interpretation,  $H$  and  $C$  are associated with all the elements of the universe  $\mathcal{U}$  that have that property

$H \Rightarrow C$  is equivalent to  $H \subseteq C$  : if  $H$  is true,  $C$  can't be false



Many students at the final exam use  $H \Rightarrow C$  and  $C \Rightarrow H$  interchangeably:  
**don't do that!**

## Deductive proof

**Deduction** : Sequence of statements that starts from one or more hypotheses and leads to a conclusion

Each step of the deduction uses some **logical rule**, applying it to the hypotheses or to one of the previously obtained statements

**Modus ponens** : logical rule to move from one statement to the next. If we know that “if  $H$  then  $C$ ” is true, and if we know that  $H$  is true, then we can conclude that  $C$  is true

## Example

**Theorem** If  $x$  is the sum of the squares of four positive integers, then  $2^x \geq x^2$

$x$  is a **parameter** and is universally quantified; the theorem is valid for all  $x$ 's that satisfy the hypotheses

See textbook for example of a deductive proof

## Deductive proof

Theorems having the form

$C_1$  *if and only if*  $C_2$

require proofs for both **directions** :

- “if  $C_2$  then  $C_1$ ”
- “if  $C_1$  then  $C_2$ ”, which is equivalent to “ $C_1$  only if  $C_2$ ”

## Additional techniques

**Reduction to definitions** : Convert all terms in the assumptions using the corresponding definitions

**Proof by contradiction** : To prove “if  $H$  then  $C$ ”, prove “ $H$  and not  $C$  implies falsehood”



## Example

**Theorem** Let  $S$  be a finite subset of an infinite set  $U$ . Let  $T$  be the complement set of  $S$  with respect to  $U$ . Then  $T$  is infinite

**Proof**  $S$  is finite, by definition, is equivalent to :  
there is an integer  $n$  such that  $|S| = n$

$U$  is infinite, by definition, is equivalent to :  
for no integer  $n$  we have  $|U| = n$

$T$  is the complement set of  $S$ , by definition, is equivalent to :  
 $S \cup T = U$  and  $S \cap T = \emptyset$

## Example

Let us consider the denial of the conclusion : “ $T$  is a finite set”  
(proof by contradiction)

$T$  is finite, by definition, is equivalent to :  
there is an integer  $m$  such that  $|T| = m$

Using  $|S| = n$  and using both  $S \cup T = U$  and  $S \cap T = \emptyset$ , we have that  $|U| = |S| + |T| = n + m$ , that is,  $U$  is finite. But this is against our hypothesis □

## Additional techniques

**Counterexample** : to prove that a theorem is false it is enough to show a case in which the statement is false

**Example** :

Is it true that if  $x$  is a prime number, then  $x$  is odd ?

No, in fact 2 is a prime number but it is not odd

# Quantifiers

**For each**  $x$  ( $\forall x$ ) : applies to all values of the variable

**Exists**  $x$  ( $\exists x$ ) : applies to at least one value of the variable

The **ordering** of the quantifiers affects the meaning of the statement

Very important for pumping lemma in chapters 4 and 7

## Example

**Theorem** If  $S$  is an infinite set, then for every integer  $n$  there exists at least one subset  $T$  of  $S$  with  $n$  elements

$\forall$  precedes  $\exists$ ; for the proof we must therefore (in that order)

- consider an arbitrary  $n$
- prove the existence of a subset  $T$  of  $S$  with  $n$  elements

# Set Equality

If  $E$  and  $F$  are sets, to prove  $E = F$  we have to prove  $E \subseteq F$  and  $F \subseteq E$

This amounts to show two statements of the form “if  $H$  then  $C$ ” :

- if  $x$  is in  $E$  then  $x$  is in  $F$
- if  $x$  is in  $F$  then  $x$  is in  $E$

# Contrapositive

The statement “if  $H$  then  $C$ ” is **equivalent** to the statement  
“if  $C$  is false then  $H$  is false”  
called **contrapositive**

Proof of equivalence uses **truth table**

In some cases, it may be easier to demonstrate the contrapositive  
Also known as *modus tollens*

# Inductive proof

Main technique when working on **recursively** defined objects (expressions, trees, etc.)

**Induction on integers** : we need to prove statement  $S(n)$ , for non-negative integer numbers  $n$

- in the **base** case we show  $S(i)$  for some specific integer  $i$  (usually  $i = 0$  or  $i = 1$ )
- in the **inductive** step, for  $n \geq i$  prove statement “if  $S(n)$  then  $S(n + 1)$ ”

We can then conclude that  $S(n)$  is true for every  $n \geq i$ , where  $i$  is the base case

**Think:** why is induction so powerful?



## Example

**Theorem** If  $x \geq 4$ , then  $2^x \geq x^2$

### Proof

**Base**  $x = 4 \Rightarrow 2^x = 16$  and  $x^2 = 16$

**Induction** Let us assume  $2^x \geq x^2$  for  $x \geq 4$

We need to show that  $2^{x+1} \geq (x+1)^2$  :

- $2^{x+1} = 2 \cdot 2^x \geq 2 \cdot x^2$ , from the inductive hypothesis
- we now show  $2x^2 \geq (x+1)^2 = x^2 + 2x + 1$
- dividing by  $x \neq 0$  :  $x \geq 2 + 1/x$
- if  $x \geq 4$ ,  $1/x \leq 1/4 \Rightarrow 2 + 1/x \leq 2.25$



## Inductive proof

We can **extend** the base part to a finite number of cases

We can **extend** the inductive step and demonstrate for a certain  $k > 0$ : “if  $S(n - k), S(n - k + 1), \dots, S(n - 1), S(n)$  then  $S(n + 1)$ ”

# Structural induction

Many structures can be defined recursively

Definition of **arithmetic expression**

**Base** Any variable or number is an arithmetic expression

**Induction** If  $E$  and  $F$  are arithmetic expressions, then also  $E + F$ ,  $E \times F$ , and  $(E)$  are arithmetic expressions

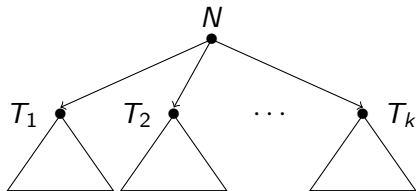
**Example** :  $3 + (x \times 2)$  and  $(2 \times (5 + 7)) \times y$  are arithmetic expressions

# Structural induction

Definition of **tree** (with root)

**Base** A single node  $N$  is a tree with root  $N$

**Induction** If  $T_1, T_2, \dots, T_k$ ,  $k \geq 1$ , are trees, the following structure is a tree with root  $N$



# Structural induction

To prove theorems for structure  $X$  which is recursively defined :

- show the statement for the base cases of the definition of  $X$
- show the statement for  $X$  on the basis of the same statement holding for the subparts of  $X$ , according to  $X$ 's definition

## Example

**Theorem** Each arithmetic expression has an equal number of open and closed parentheses

**Proof** We proceed by induction on the number of parentheses

**Base** Both variables and numbers have zero open parentheses and zero closed parentheses

**Induction** Let us assume that  $E$  has  $n$  open and closed parentheses and  $F$  has  $m$  of them

There are three ways to recursively construct an arithmetic expression :

- $E + F$  has  $n + m$  open brackets and  $n + m$  closed brackets
- $E \times F$  has  $n + m$  open brackets and  $n + m$  closed brackets
- $(E)$  has  $n + 1$  open brackets and  $n + 1$  closed brackets □

## Example

**Theorem** Let  $T$  be a tree with  $n$  nodes and  $e$  arcs. Then  
 $n = e + 1$

Before proving the theorem, try to get a visual intuition of why this is true

**Proof** By induction on  $T$ 's structure

**Base**  $T$  has  $n = 1$  and  $e = 0$

**Induction** Assume  $T_i$  has  $n_i$  nodes and  $e_i$  arcs. By inductive hypothesis,  $n_i = e_i + 1$

We have :

$$n = 1 + \sum_{i=1}^k n_i, \quad e = k + \sum_{i=1}^k e_i$$

## Example

We can write :

$$\begin{aligned}n &= 1 + \sum_{i=1}^k n_i \\ &= 1 + \sum_{i=1}^k (1 + e_i) && \text{inductive hypothesis} \\ &= 1 + k + \sum_{i=1}^k e_i \\ &= 1 + e\end{aligned}$$





# Alphabet & strings

**Alphabet** : **finite** and **nonempty** set of atomic symbols

**Example** :

- $\Sigma = \{0, 1\}$ , the binary alphabet
- $\Sigma = \{a, b, c, \dots, z\}$ , the set of all lowercase letters
- the set of all printable ASCII characters

**String** : **finite** sequence of symbols from some alphabet

- 0011001 string over  $\Sigma = \{0, 1\}$

# Alphabet & strings

**Empty string** : The string with zero symbols (taken from any alphabet) is denoted  $\epsilon$

**Length** of a string : Number of **occurrences** (standpoints) for the symbols in the string

- $|w|$  denotes the length of the string  $w$
- $|0110| = 4$ ,  $|\epsilon| = 0$

# Alphabet & strings

**Powers** of an alphabet :  $\Sigma^k$  is the set of all  $k$ -length strings with symbols from  $\Sigma$

- $\Sigma = \{0, 1\}$
- $\Sigma^1 = \{0, 1\}$ ; **ambiguity** between  $\Sigma$  and  $\Sigma^1$

Elements of  $\Sigma$  are alphabet symbols, elements of  $\Sigma^1$  are strings

- $\Sigma^2 = \{00, 01, 10, 11\}$
- $\Sigma^0 = \{\epsilon\}$

**Question** : How many strings are there in  $\Sigma^3$  ?

# Alphabet & strings

The set of all strings from  $\Sigma$  is denoted  $\Sigma^*$

We have

$$\Sigma^* = \Sigma^0 \cup \Sigma^1 \cup \Sigma^2 \cup \dots$$

$$\Sigma^+ = \Sigma^1 \cup \Sigma^2 \cup \Sigma^3 \cup \dots$$

$$\Sigma^* = \Sigma^+ \cup \{\epsilon\}$$

**Question** : It is a mistake to write  $\Sigma^+ \cup \epsilon$  : why?

# Alphabet & strings

**Concatenation** : If  $x$  and  $y$  are strings, then  $xy$  is the string obtained by putting a copy of  $y$  immediately after a copy of  $x$

**Example** :

$$x = 01101$$

$$y = 110$$

$$xy = 01101110$$

Sometimes we also use the the  $\cdot$  operator to represent concatenation and write  $x \cdot y$

Some textbooks use the notation  $x.y$

# Alphabet & strings

For each string  $x$  :

$$x\epsilon = \epsilon x = x$$

$\epsilon$  is the **neutral** element of the concatenation

You can always think of  $\epsilon$  occurring any number of times within a string :

$$\begin{aligned}x \cdot y &= x \cdot \epsilon \cdot y \\ &= x \cdot \epsilon \cdot \epsilon \cdot y \\ &= x \cdot \epsilon \cdot \epsilon \cdot \epsilon \cdot y \\ &= \dots\end{aligned}$$

Compare with  $2 + 3 = 2 + 0 + 3 = 2 + 0 + 0 + 3 = \dots$

# Alphabet & strings

## Notational conventions :

- $a, b, c, \dots, a_1, a_2, \dots, a_i, \dots$  alphabet symbols
- $u, w, x, y, z$  strings
- for  $n \geq 0$ ,  $a^n = aa \cdots a$  ( $a$  repeated  $n$  times)
- $a^0 = \epsilon$ ,  $a^1 = a$

# Languages

A **language** is a set of strings arbitrarily chosen from  $\Sigma^*$ , where  $\Sigma$  is an alphabet.  $L \subseteq \Sigma^*$  is a language

## Example :

- set of all the words in some English dictionary
- set of all Java programs without syntactic errors
- set of strings consisting of  $n$  zeros followed by  $n$  ones, with  $n \geq 0$

$$\{\epsilon, 01, 0011, 000111, \dots\}$$

- set of strings with an equal number of 0's and 1's

$$\{\epsilon, 01, 10, 0011, 0101, 1001, \dots\}$$

What is  $\Sigma$  in the first two cases above?



# Languages

## Example :

- set of binary numbers whose value is a prime

$$L_p = \{10, 11, 101, 111, 1011, \dots\}$$

- empty language  $\emptyset$ , contains no string
- language  $\{\epsilon\}$ , contains only the empty string

Do not confuse these two languages :

$$\emptyset \neq \{\epsilon\}$$

# Languages

**Extensional** representation of a language :

$$L = \{\epsilon, 01, 0011, 000111, 00001111, \dots\}$$

**Intensional** representation of a language, using a **set-former** :

$$L = \{w \mid \text{statement specifying } w\}$$

**Example** :

- $\{w \mid w \text{ consists of an equal number of 0's and 1's}\}$
- $\{w \mid w \text{ is an integer binary number whose value is prime}\}$
- $\{w \mid w \text{ is a syntactically correct Java program}\}$

# Languages

Set-formers are often expressed in mathematical form :

$$L = \{w \mid w = 0^n 1^n, n \geq 0\}$$

or, in simplified form, also as :

$$L = \{0^n 1^n \mid n \geq 0\}$$

which is equivalent to :

$$L = \{\epsilon, 01, 0011, 000111, \dots\}$$

Note the **implicit** universal quantifier for  $n$  in the set-former above

When needed, existential quantifiers are written explicitly

# Languages

## Example :

- $\{0^i1^j \mid i, j \geq 1, i \geq j\}$
- $\{0^i1^j \mid i, j \geq 1, i > j \text{ or } i < j\}$

The comma punctuation symbol is an implicit 'and' operator above

## Note :

 do not confuse the two notations

- $\{0^n1^n \mid n \geq 0\}$
- $\{0^n1^n\}, n \geq 0$

There is **no precise syntax** for the use of set-formers

This requires some experience, many students get confused about this

## Decision problems

Let  $P(x)$  be a **predicate** expressing some mathematical property of element  $x$

**Decision problem** associated with  $P$  : on input  $x$ , decide whether  $P(x)$  holds true

Associated formal language ( $x$  viewed as a string) :

$$L_P = \{x \mid P(x) \text{ holds true}\}$$

The decision problem can be reformulated as : Given as input string  $x$ , decide whether  $x \in L_P$

## Example

For natural number  $x$ ,  $P(x)$  is true if  $x$  is a prime number. We represent  $x$  as a binary string

We define the language of prime numbers

$$L_p = \{10, 11, 101, 111, 1011, \dots\}$$

Assigned as input the binary string  $x$ , decide whether  $x \in L_p$

# Decision problems

Many mathematical problems are not decision problems, but require instead a computation that constructs an output **result**

Think about search problems, optimization problems, etc.

We can reformulate these problems as decision problems

**Example :**

- given matrices  $A$ ,  $B$ , construct the matrix  $C = A \times B$
- associated decision problem : given a triple  $\langle A, B, C \rangle$ , decide whether  $C = A \times B$

## Decision problems

The general (non-decision) problem **is no easier** than the associated decision problem

You can solve the decision problem if you have a subroutine for the general problem

**Example** : Algorithm for decision problem using the algorithm for the general problem as a subroutine (**reduction** technique)

- input  $\langle A, B, C \rangle$
- use subroutines on  $A, B$  to produce  $C' = A \times B$
- if  $C' = C$  answer yes, otherwise answer no

If you have enough computational resources to solve the general problem, then you can also solve the decision problem