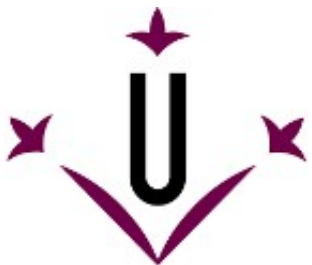


Advanced Programming in Artificial Intelligence

Local Search and SAT

Grau en Enginyeria Informàtica
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Local Search

- Introduction
- The Satisfiability Problem
- Random SAT Problem generator
- Local Search Algorithms for SAT

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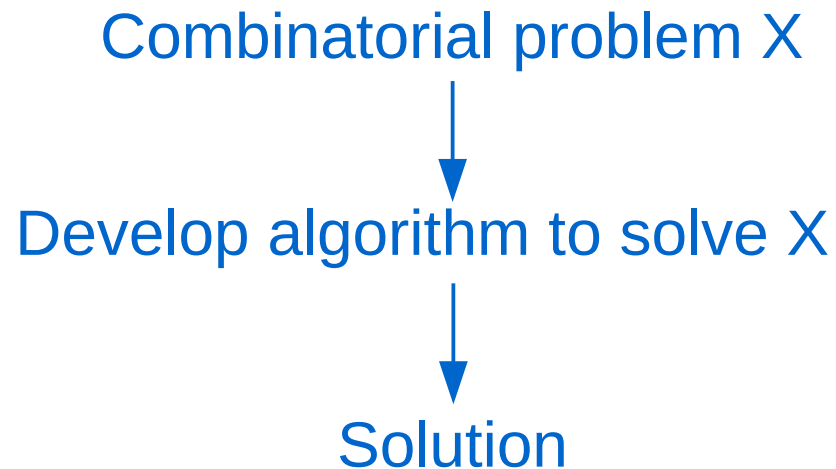
Local Search and SAT

Introduction



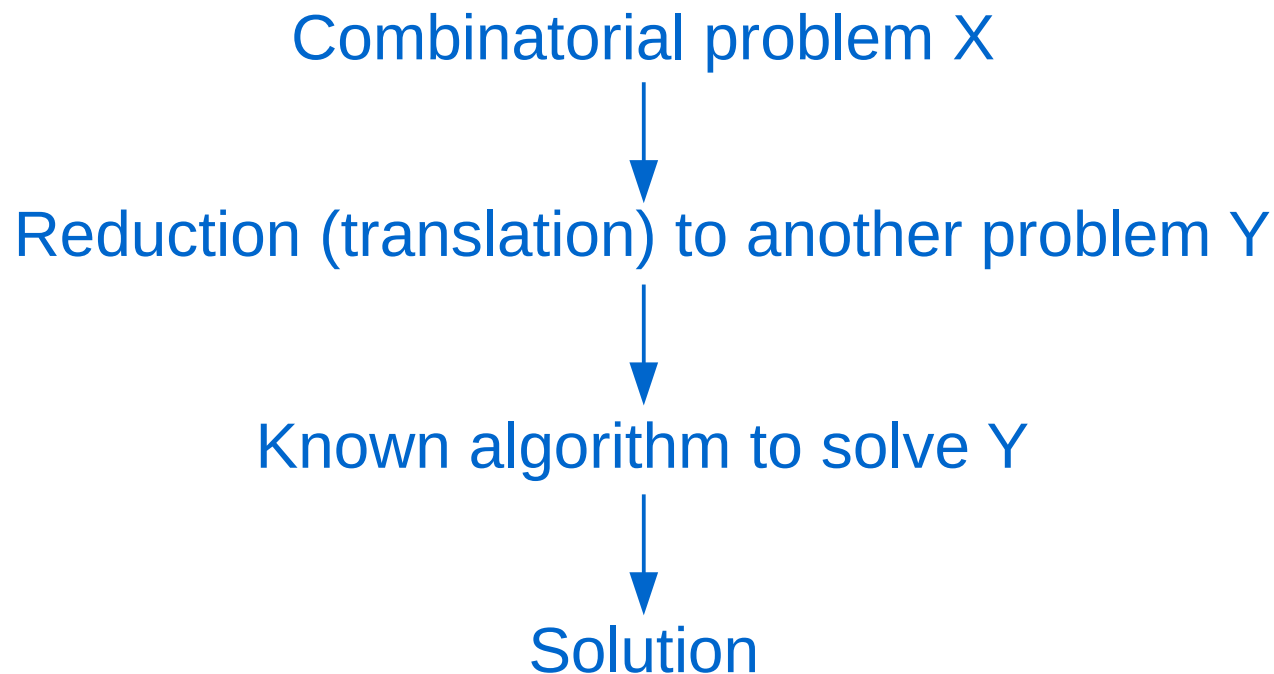
Introduction

- **Real-world problems:** Planning, Scheduling, Circuit design, Software verification, Automated theorem proving...



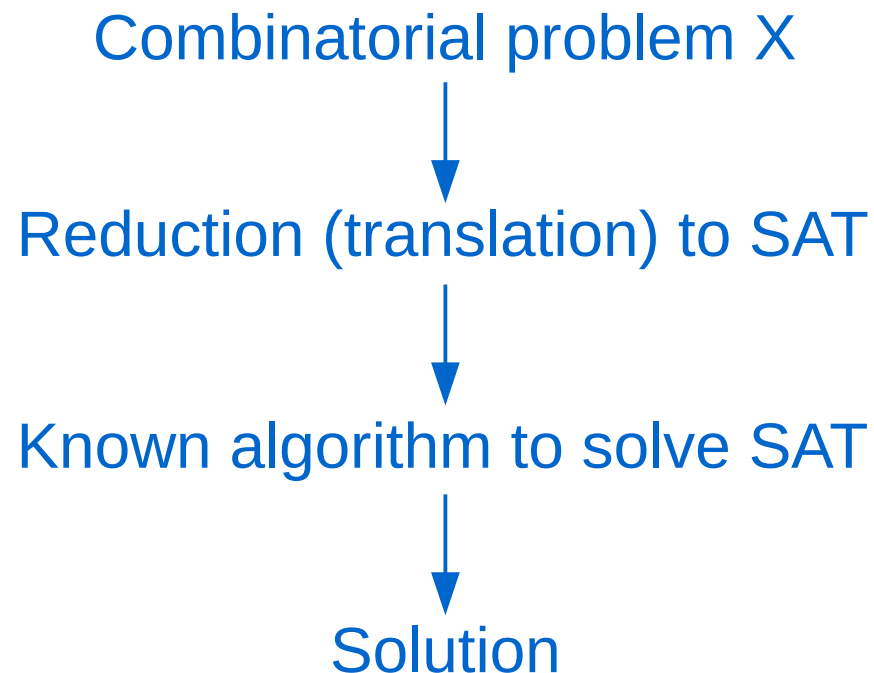
Introduction

- **Real-world problems:** Planning, Scheduling, Circuit design, Software verification, Automated theorem proving → SAT



Introduction

- **Real-world problems:** Planning, Scheduling, Circuit design, Software verification, Automated theorem proving → SAT



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Local Search and SAT

The Satisfiability Problem



The Satisfiability Problem

The Boolean Satisfiability problem (SAT)

- SAT is the problem of determining if a propositional formula in Conjunctive Normal Form (CNF) is satisfiable
 - In other words, if there exists an interpretation which satisfies the formula

The CNF Formulas

Set of variables $X = \{x_1, x_2, \dots, x_n\}$

- **Literal:** a positive or negative variable
 - Ex: $x_i, \neg x_i$
- **Clause:** a disjunction of literals
 - Ex: $x_1 \vee \neg x_2 \vee x_3, x_2, \neg x_2 \vee \neg x_3 \dots$
- **CNF Formula:** a conjunction of clauses
 - Ex: $(x_1 \vee \neg x_2 \vee x_3) \wedge (x_2) \wedge (\neg x_2 \vee \neg x_3)$

Note: \vee = logical **or**, \wedge = logical **and**

The CNF Formulas

- **CNF Formula:** a conjunction of clauses
 - Ex: $(x_1 \vee \neg x_2 \vee x_3) \wedge (x_2) \wedge (\neg x_2 \vee \neg x_3)$
Note: \vee = logical **or**, \wedge = logical **and**
 - Alternative notation: sets notation
 - Ex: $\{ \{ x_1 \vee \neg x_2 \vee x_3 \}, \{ x_2 \}, \{ \neg x_2 \vee \neg x_3 \} \}$
 - \square denotes an empty clause
 - \emptyset denotes an empty formula

The CNF Formulas

- **Interpretation (I):** is an assignment of truth values $\{0, 1\}$ (false, true) to variables
- We say that I **satisfies** a:
 - **Literal** x_i iff $I(x_i) = 1$
 - **Literal** $\neg x_i$ iff $I(x_i) = 0$
 - **Clause** c iff I satisfies at least on literal of c
 - **CNF Formula** F iff I satisfies all the clauses of F
- A **CNF Formula** F is **satisfiable** iff exist an interpretation that satisfies F . Otherwise F is **unsatisfiable**.

The CNF Formulas

- Examples

- $\{ \square, \{ x_2 \}, \{ \neg x_2 \vee \neg x_3 \} \}$ is unsatisfiable
 - Every formula with an empty clause is trivially unsatisfiable
- $\{ \emptyset \}$ is satisfiable
 - The empty formula is satisfiable
- $\{ \{ x_1 \vee \neg x_2 \vee x_3 \}, \{ x_2 \}, \{ \neg x_2 \vee \neg x_3 \} \}$?

The CNF Formulas

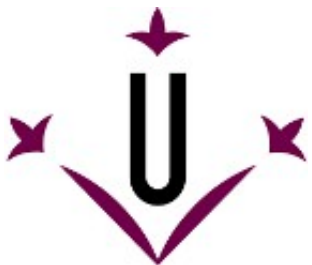
- Examples

- $\{ \square, \{ x_2 \}, \{ \neg x_2 \vee \neg x_3 \} \}$ is unsatisfiable
 - Every formula with an empty clause is trivially unsatisfiable
- $\{ \emptyset \}$ is satisfiable
 - The empty formula is satisfiable
- $\{ \{ x_1 \vee \neg x_2 \vee x_3 \}, \{ x_2 \}, \{ \neg x_2 \vee \neg x_3 \} \}$ is satisfiable
 - $x_1 = 1, x_2 = 1, x_3 = 0$ is an interpretation that satisfies the formula
 - There are more interpretations ?

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Local Search and SAT

Random SAT Problem generator



CNF format

- **Comments:** Line starting with character 'c'
 - Ex: c Random CNF formula
- **Problem Line:** p cnf *num_variables num_clauses*
 - Ex: p cnf 4 10
- **Clauses:** List of clauses, one per line
 - Clause: sequence of integers between *num_variables* and *-num_variables* ending with 0 on the same line
 - Positive number → positive literal of the variable
 - Negative number → negative literal of the variable
 - Ex: -3 5 24 -18 0

CNF format

SAT input format example

c

c A random CNF formula

c

p cnf 3 4

1 -2 0

-1 2 -3 0

-3 2 0

1 3 0

CNF format

SAT input format example

c

c A random CNF formula

c

p cnf 3 4

1 -2 0

-1 2 -3 0

-3 2 0

1 3 0

Is there a solution?

1 = true

2 = true

3 = false

Solver output solution:

SATISFIABLE

1 2 -3 0

Random CNF generator

- Example of a Random k-SAT generator
 - Random generator of CNF problems
 - K is the length of the clauses
 - Parameters of the generator
 - Number of variables
 - Number of clauses
 - Length of the clauses (K)
 - Random seed

Random 3-SAT Hardness

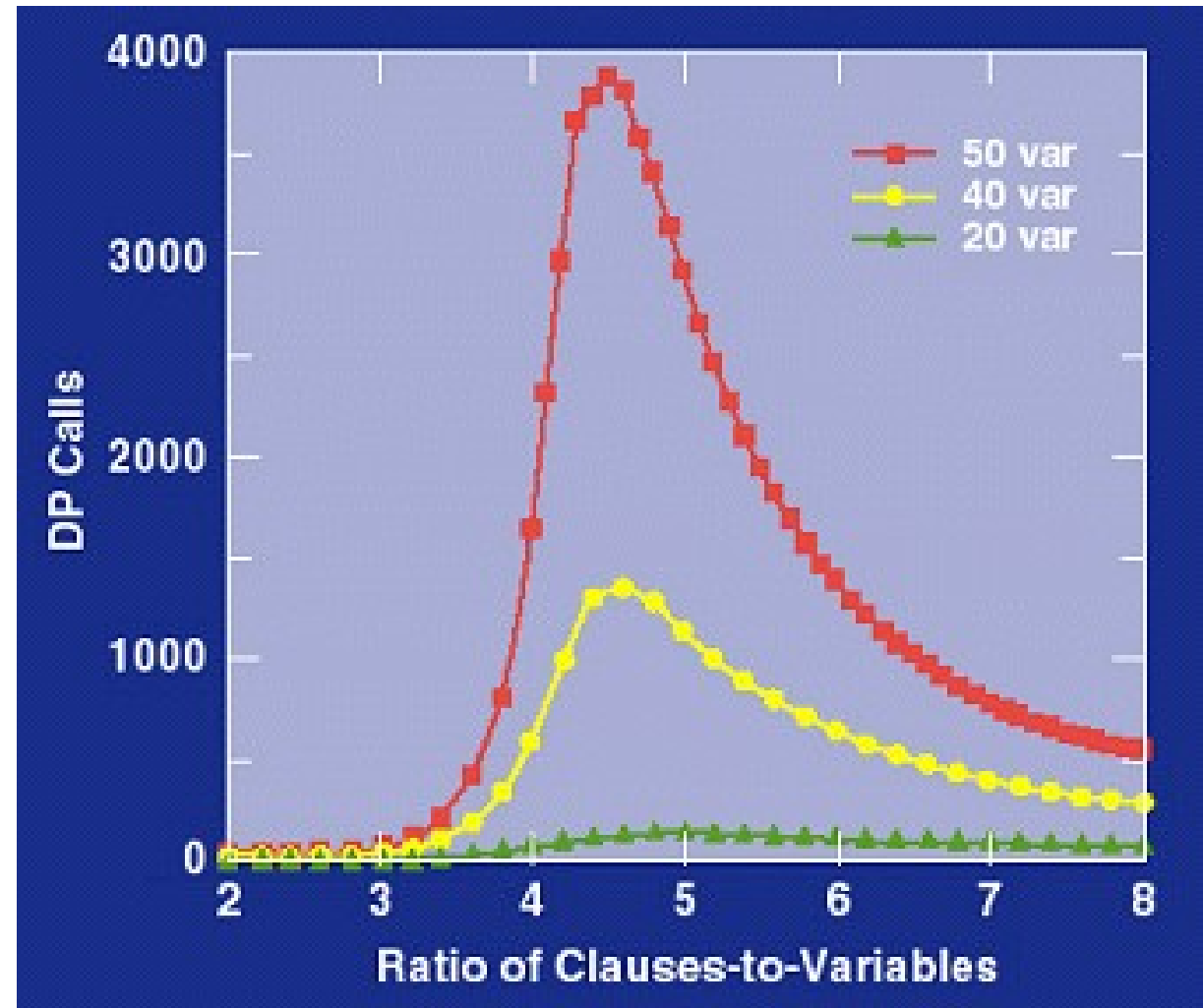
Easy-hard-easy pattern

The **hardness** depends on:

#clauses

#variables

(peak around 4.25)

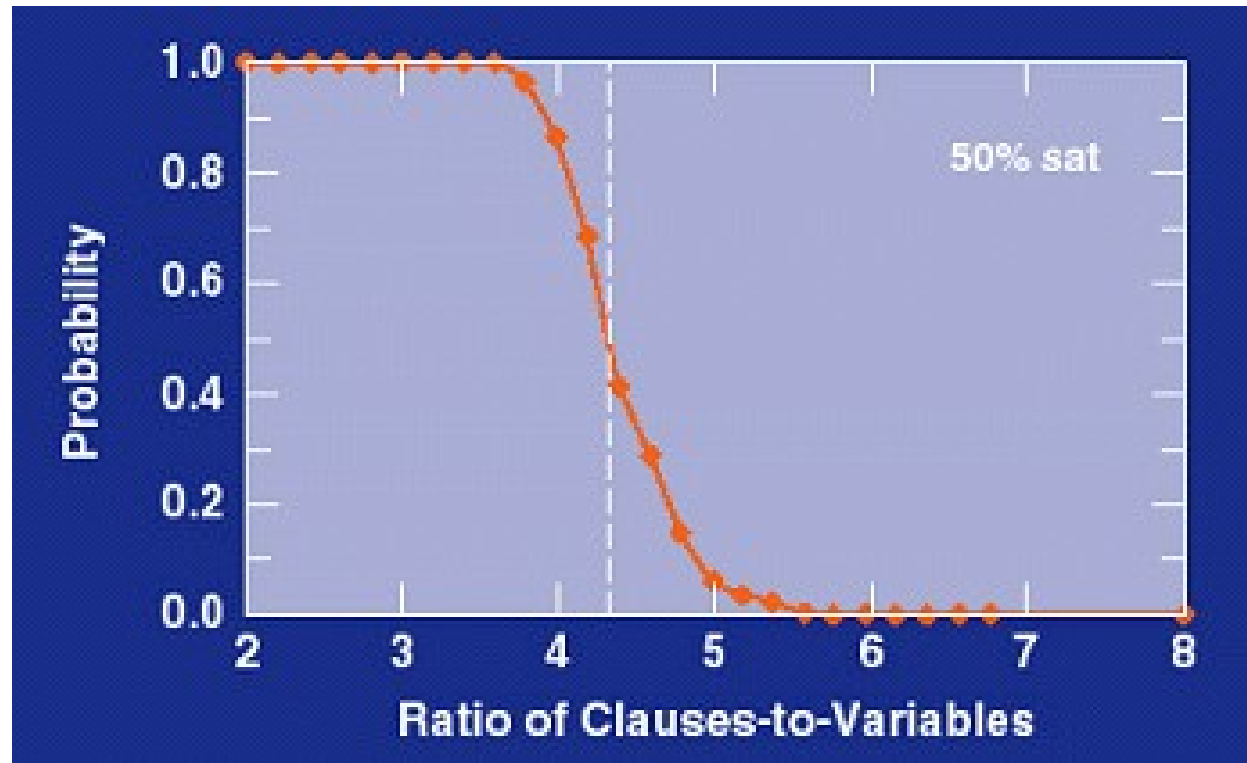


Random 3-SAT Probability

Phase transition

From **satisfiable** formulas
to **unsatisfiable** formulas

(peak around 4.25)



Experimentation

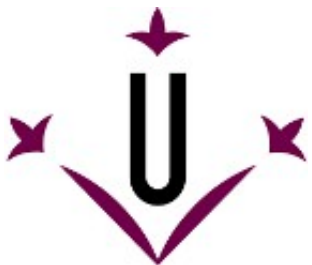
Reproduce the two previous plots

- Generate the benchmarks (random generator)
- Run the solver with all the benchmarks
 - Send the output to a file (for all the benchmarks)
- Parse the output needed from the results
- Summarize the results
 - Table
 - Plot
 - ...

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Local Search and SAT

Local Search algorithms for SAT



Local Search algorithms for SAT

- Do not follow a systematic method to explore all the search space
 - Not conscious of the explored space
- When they finish...
 - they have found a solution, or
 - they reached a limit of resources
 - In this case, they can not say anything about the solution
- Examples
 - GSAT
 - WalkSat

GSAT

```
F ← input CNF formula
for i = 1 to max_tries do
  I ← random interpretation for F
  for j = 1 to max_flips do
    if I satisfies F then
      return I
    endif
    S ← set of variables s.t. flipping them
        increments more the total number of satisfied
        clauses
    p ← a variable of S
    I ← I with the value of p flipped
  endfor
endfor
return "No solution found"
```


GSAT

- GSAT can not be used to proof that a formula is unsatisfiable
 - Incomplete procedure
 - If GSAT gets an interpretation, then the formula is SAT
 - If GSAT can not find an interpretation, then we do not know if the formula is SAT or UNSAT
- GSAT is able to solve hard instances with more than 2.000 variables
- Different runs with the same parameters → different solutions
- Problems with local minimums

GSAT: Escaping from local minimums

- **Restarts**
 - Restarts the search with a new random interpretation after a given number of flips
- **Random walk (GWSAT)**
 - with probability ω :
 - Flips a variable from a random unsatisfied clause
 - with probability $1 - \omega$:
 - Does like GSAT
- **Tabu search**
 - Some recent flips that take to visited solutions are forbidden
 - GSAT forces to explore new regions in flat areas

Walksat

```
F ← input CNF formula
for i = 1 to max_tries do
  I ← random interpretation for F
  for j = 1 to max_flips do
    if I satisfies F then
      return I
    endif
    C ← a clause of F not satisfied by I
    S ← set of variables that appear in C
    b ← min({broken(p, F, I) | p in S})
    if b > 0 and with probability ω then
      p ← a variable of S
    else
      p ← a variable of S s.t. broken(p, F, I) = b
    endif
    I ← I with the value of p flipped
  endfor
endfor
return "No solution found"
```

GSAT and Walksat implementations

- There are several implementations of GSAT and Walksat
 - The original: Walksat/GSAT Home Page
 - GSAT is no longer maintained
 - **UBCSAT:** UBCSAT SATLib
 - Implements a very large sample of solvers
 - Current version 1.1.0

UBCSAT

- `./ubcsat -alg gsat -solve -i test.cnf`
 - `-alg <name>`: name of the algorithm (-ha for the list)
 - `-solve`: stops when it finds a solution (see -target)
 - `-i <file_name>`: input CNF instance
 - `-runs INT`: number of independent tries (default 1)
 - `-cutoff INT`: number of flips per run (default 100.000)
 - `-target INT`: target solution quality (default 0)
 - `-seed INT`: initial random seed
 - `-wp <[0,1]>`: random walk probability (default 0.5)

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Bibliography

- “Artificial Intelligence: A Modern Approach”. S. Russell and P. Norvig. Prentice Hall.
- Wikipedia and several Internet sources.

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