

Introduction to Propositional Satisfiability

IA085: Satisfiability and Automated Reasoning

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FI MUNI, Spring 2025

Contents

Propositional satisfiability (SAT)

- $(A \vee \neg B) \wedge (\neg A \vee C)$
- is it **satisfiable**?

Satisfiability modulo theories (SMT)

- $x = 1 \wedge x = y + y \wedge y > 0$
- is it satisfiable over **reals**?
- is it satisfiable over **integers**?

Automated theorem proving (ATP)

- axioms: $\forall x (x + x = 0), \forall x \forall y (x + y = y + x)$
- do they **imply** $\forall x \forall y ((x + y) + (y + x) = 0)$?

For each of the problems (SAT/SMT/ATP)

- necessary definitions and theoretical results
- algorithms to solve the problem
- usage in practice and practical considerations

Organization of the Course

Schedule and Requirements

During semester

- lecture every week (except May 7)
- seminar every other week
- project (write your own small SAT solver) – mandatory

Exam

- oral exam

Implement your own SAT solver

- you can use any **reasonable** programming language (C, C++, C#, Go, Java, Python, Rust, ...)
- you **are encouraged** to work **in pairs** (but you do not have to)
- technical requirements are specified in the information system
- more advanced features → bonus points for the exam
- the scores will be evaluated periodically through the semester, you will see the ranking

Who am I?

- author of SMT solver Q3B for quantified formulas over bit-vector theory
- for 3 years post-doctoral researcher in Fondazione Bruno Kessler: research focused on SMT-based verification of software and SAT-based verification of hardware
- PhD thesis about satisfiability of quantified formulas over bit-vector theory
- author of several research papers about solving SMT and using it in practice
- co-organizer of SMT-COMP 2024, 2025

Propositional Logic

Propositional logic

Propositional logic deals with propositions, their relationships, and arguments based on them.

Does not deal objects and their properties, just with separate **atomic** claims.

“Martin has brown hair”

“Martin does not have hair”

No relationship as far as **propositional** logic is concerned.

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“Martin has brown hair”

if “Martin has brown hair” then “Martin does have hair”

Implies *“Martin does have hair”* .

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“Martin has brown hair” (A)

if *“Martin has brown hair”* then *“Martin does have hair”* ($A \rightarrow B$)

Implies *“Martin does have hair”* (B).

Let $V = \{A, B, C, \dots\}$ be a countable set of **propositional variables**. The set of propositional formulas is defined inductively as

- \top and \perp are propositional formulas,
- v is a propositional formula for each $v \in V$ (called **propositional atom**),
- if φ is a propositional formula, $\neg\varphi$ is a propositional formula,
- if φ and ψ are propositional formulas, $\varphi \wedge \psi$, $\varphi \vee \psi$, $\varphi \rightarrow \psi$, and $\varphi \leftrightarrow \psi$ are propositional formulas.

Example

- $A \wedge B$
- $(A \vee B) \leftrightarrow \neg C$

Formulas of form v or $\neg v$ are called **literals**.

Semantics: Truth Assignments

$Atoms(\varphi)$ = the set of all atoms of formula φ

(Total) truth assignment for formula φ

- assigns true (\top) or false (\perp) to each propositional variable in φ
- a function $\mu: V' \rightarrow \{\top, \perp\}$ where $Atoms(\varphi) \subseteq V'$
- can be written as a set of non-contradictory literals containing all variables of φ

Example

- formula $\varphi = A \vee B$,
- total assignment $\mu(A) = \top, \mu(B) = \perp$,
- written as $\mu = \{A, \neg B\}$.

Semantics: Satisfaction

Define when a truth assignment μ **satisfies** the formula φ , written $\mu \models \varphi$:

- $\mu \models \top$
- $\mu \models v$ if $\mu(v) = \top$
- $\mu \models \neg\psi$ if **not** $\mu \models \psi$
- $\mu \models \psi_1 \wedge \psi_2$ if $\mu \models \psi_1$ **and** $\mu \models \psi_2$
- $\mu \models \psi_1 \vee \psi_2$ if $\mu \models \psi_1$ **or** $\mu \models \psi_2$
- $\mu \models \psi_1 \rightarrow \psi_2$ if **not** $\mu \models \psi_1$ **or** $\mu \models \psi_2$
- $\mu \models \psi_1 \leftrightarrow \psi_2$ if $\mu \models \psi_1$ **if and only if** $\mu \models \psi_2$

Semantics: Model

If $\mu \models \varphi$, we say that μ is a model of φ

Example

$\{A, \neg B, C\}$ is a model of $A \wedge (B \leftrightarrow \neg C)$

An assignment μ is a partial model of φ if each extension of μ that is a truth assignment to φ (i.e., $Atoms(\varphi) \subseteq dom(\mu)$) is a model of φ

Example

$\{A, B\}$ is a partial model of $(A \wedge B) \vee (A \wedge C)$

Propositional Entailment

Formula φ **propositionally entails** formula ψ (written $\varphi \models \psi$) if every μ that is a truth assignment for both φ and ψ (i.e., $(Atoms(\varphi) \cup Atoms(\psi)) \subseteq dom(\mu)$) satisfies

if $\mu \models \varphi$ then also $\mu \models \psi$

Example

- $A \models A \vee B$
- $(A \rightarrow B) \wedge A \models B$
- $(A \vee B) \wedge (\neg A \vee C) \models (B \vee C)$
- $A \not\models A \wedge B$

Propositional Equivalence

Formulas φ and ψ are **propositionally equivalent** (written $\varphi \equiv \psi$) if

$$\varphi \models \psi \quad \text{and} \quad \psi \models \varphi$$

Example

- $A \wedge (B \vee C) \equiv (A \wedge B) \vee (A \wedge C)$
- $A \wedge (A \vee B) \equiv A$
- $\neg(A \wedge B) \equiv \neg A \vee \neg B$

Negation Normal Form

Negation Normal Form (NNF)

- negations are applied only to propositional atoms
- the formula does not contain implication (\rightarrow) and equivalence (\leftrightarrow)

Transformation to NNF

1. rewrite all $\varphi \leftrightarrow \psi$ to $(\varphi \rightarrow \psi) \wedge (\varphi \leftarrow \psi)$
2. rewrite all $\varphi \rightarrow \psi$ to $\neg\varphi \vee \psi$
3. apply De Morgan rules until fixed point
 - rewrite $\neg(\varphi \wedge \psi)$ to $(\neg\varphi) \vee (\neg\psi)$
 - rewrite $\neg(\varphi \vee \psi)$ to $(\neg\varphi) \wedge (\neg\psi)$

Conversion to NNF: Complexity

What is the complexity of conversion to NNF?

$$\varphi \leftrightarrow \psi \sim (\neg\varphi \vee \psi) \wedge (\varphi \vee \neg\psi)$$

Each equivalence doubles the size of the formula \rightarrow translation can be **exponential!**

Conversion to NNF: Complexity

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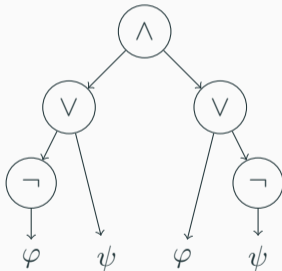
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Or is it? It depends on the **representation of the formulas**

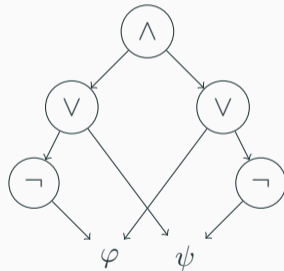
Representation of Formulas

$$(\neg\varphi \vee \psi) \wedge (\varphi \vee \neg\psi)$$

Tree



Directed acyclic graph (DAG)



In practice, we represent formulas as **DAGS**.

Conversion to NNF: Complexity

Theorem

*When representing formulas as DAGs, the transformation to NNF is **linear**.*

Proof idea.

The DAG contains two nodes for each subformula φ : one for φ , one for $\neg\varphi$. □

Conversion to NNF: Complexity

Theorem

When representing formulas as DAGs, the transformation to NNF is *linear*.

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The DAG contains two nodes for each subformula φ : one for φ , one for $\neg\varphi$. □

Proof details (bonus).

Recursively define function $NNF(\varphi) = (\varphi^+, \varphi^-)$. Given $NNF(\psi) = (\psi^+, \psi^-)$ and $NNF(\rho) = (\rho^+, \rho^-)$:

$$NNF(\psi \wedge \rho) = (\psi^+ \wedge \rho^+, \psi^- \vee \rho^-)$$

$$NNF(\neg\psi) = (\psi^-, \psi^+)$$

$$NNF(\psi \leftrightarrow \rho) = (\underbrace{(\psi^- \vee \rho^+) \wedge (\psi^+ \vee \rho^-)}_{\text{positive}}, \underbrace{(\psi^+ \wedge \rho^-) \vee (\psi^- \wedge \rho^+)}_{\text{negative}}).$$

For more details, see *Property 1* in Gabriele Masina, Giuseppe Spallitta, Roberto Sebastiani: *On CNF Conversion for SAT Enumeration*.

Conjunctive Normal Form

Clause

- disjunction of literals
- $A \vee \neg B \vee C$
- written as $\{A, \neg B, C\}$ thanks to idempotence, commutativity, and associativity
- what is $\{\}$?

Formula in Conjunctive Normal Form (CNF)

- conjunction of clauses
- $(A \vee \neg B \vee C) \wedge (B \vee \neg C) \wedge C$
- written as $\{\{A, \neg B, C\}, \{B, \neg C\}, \{C\}\}$ thanks to idempotence, commutativity, and associativity
- what are $\{\}$? and $\{\emptyset\}$?

Conjunctive Normal Form

- easy to represent (`clause = list[int]`, `formula = list[clause]`)
- easy to write algorithms, do not have to deal with the structure of the formula
- most of modern SAT solvers have input in CNF

Conjunctive Normal Form

Transformation to CNF (naive)

1. transform to NNF
2. apply distributivity until fixed point
 - rewrite $\varphi \vee (\psi \wedge \rho)$ to $(\varphi \vee \psi) \wedge (\varphi \vee \rho)$
 - rewrite $(\psi \wedge \rho) \vee \varphi$ to $(\psi \vee \varphi) \wedge (\rho \vee \varphi)$

This is again **exponential**, try with $\bigvee_{1 \leq i \leq n} (A_i \wedge B_i)$. ☹

Can we do better? What if the DAG representation is used? What if we use a different algorithm?

Conversion to CNF: Naive

Theorem

There exists an infinite family of formulas $\Phi = \{\varphi_i \mid i \in \mathbb{N}\}$ such that *for each* equivalent family of formulas with $\varphi_i^{CNF} \equiv \varphi_i$, the size $|\varphi_i^{CNF}|$ grows *exponentially* with respect to $|\varphi_i|$ (even for DAG representation).

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Proof.

Let $parity_i(A_1, A_2, \dots, A_i) = A_1 \oplus A_2 \oplus \dots \oplus A_i$. We can show that

- $parity_i$ can be defined by a formula φ_i of size $\mathcal{O}(i)$,
- each formula φ_i^{CNF} in CNF that defines $parity_i$ has 2^{i-1} clauses. □

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We cannot do better than exponential. 😞

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Or can we? Yes, we can! Later today.

Cube

- conjunction of literals
- $A \wedge \neg B \wedge C$

Formula in Disjunctive Normal Form (DNF)

- disjunction of cubes
- $(A \wedge \neg B \wedge C) \vee (B \wedge \neg C) \vee C$

We will not be dealing with DNF often in this course.

Propositional Satisfiability (SAT)

Satisfiability Problem

Problem (SAT)

Given a propositional formula, decide whether it is satisfiable.

Problem (CNF-SAT)

Given a propositional formula in CNF, decide whether it is satisfiable.

Problem (3-SAT)

Given a propositional formula in CNF with each clause of size 3, decide whether it is satisfiable.

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Theorem

SAT, CNF-SAT, and 3-SAT are NP-complete.

Hardness of Propositional Satisfiability: Theory

Theorem

SAT, CNF-SAT, and 3-SAT are NP-complete.

Proof ideas.

- Whether an assignment is a model can be checked in polynomial time.
- A computation of Turing machine of polynomial length can be encoded by a CNF formula of polynomial size. □

There are **no known polynomial algorithms** for propositional satisfiability.

Hardness of Propositional Satisfiability: Practice

Modern SAT solvers can decide satisfiability of formulas with **thousands** of variables and **millions** of clauses thanks to

- clever algorithms (worst case exponential)
- clever data structures
- clever heuristics

Give it a try:

- MiniSAT (<http://minisat.se/>)
- CaDiCaL (<https://github.com/arminbiere/cadical>)
- Kissat (<https://github.com/arminbiere/kissat>)

Applications: Other Logical Problems

Other logical problems can be reduced¹ to satisfiability

Validity

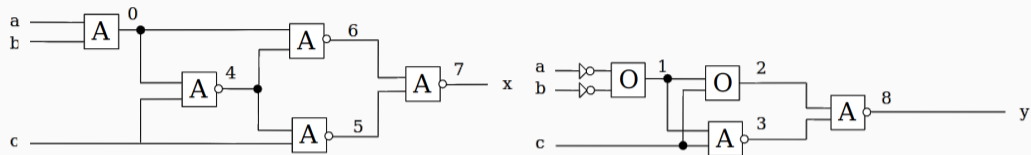
- a φ is **valid** if every total assignment for φ is its model
- φ is **valid** $\Leftrightarrow \neg\varphi$ is **not satisfiable**

Entailment

- $\varphi \models \psi \Leftrightarrow (\varphi \rightarrow \psi)$ is **valid** $\Leftrightarrow (\varphi \wedge \neg\psi)$ is **not satisfiable**

¹in the sense of Turing reductions

Applications: Hardware Design



[Example from: https://www21.in.tum.de/~lammich/2015_SS_Seminar_SAT/resources/Equivalence_Checking_11_30_08.pdf]

Are circuits C_1 and C_2 equivalent?

Is $\neg(\text{formula}(C_1) \leftrightarrow \text{formula}(C_2))$ UNSAT? (called a **miter** formula)

Works only for reasonably small circuits. For larger circuits (millions of gates), more involved techniques are necessary, e.g., SAT-sweeping.

Applications: Package Dependency

- package P has n versions:

$$x_1^P, x_2^P, \dots, x_n^P$$

- only one can be installed at a time:

$$\neg x_i^P \vee \neg x_j^P \text{ for all packages } P \text{ and versions } i \neq j$$

- packages have dependencies:

$$x_3^P \rightarrow (x_1^Q \vee x_2^Q) \wedge x_8^R$$

- I have version 1 of package Q and want to install version 3 of package P :

$$x_3^P \wedge x_1^Q$$

- what dependencies I need to install:

Is the formula SAT? What is its model?

Used for example by package manager Cabal for Haskell.

Definition

A triple $(a, b, c) \in \mathbb{N}$ is called **Pythagorean** if $a^2 + b^2 = c^2$.

Question

Can every set of numbers $N = \{1, 2, \dots, n\}$ be colored by two colors such that there is no monochromatic Pythagorean triple?

²<https://www.cs.utexas.edu/~marijn/ptn/>

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The answer is no ($n = 7825$) and was found by a SAT solver in 2016². Previous lower bound was that $n = 7664$ can be colored.

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Applications: Open Problems in Mathematics

1. Define a formula F_i whose models are two-colorings of $\{1, 2, \dots, i\}$ with no monochromatic Pythagorean triples.

$$F_i = \bigwedge_{(a,b,c) \text{ is a Pythagorean triple}} (x_a \vee x_b \vee x_c) \wedge (\neg x_a \vee \neg x_b \vee \neg x_c)$$

2. F_{7824} : 6492 variables and 18930 clauses; F_{7825} : 6494 variables and 18944 clauses.
3. Preprocessing: reduce this to 3740 variables and 14652 clauses; and 3745 variables and 14672 clauses.
4. Use parallel SAT solver and tweak some of its heuristics.
5. Use a parallel machine with 800 cores for 2 days.
6. Find that F_{7824} is satisfiable and F_{7825} is unsatisfiable.
7. Get a largest unsatisfiability proof ever (200 terabytes).

Thinking with Clauses

Clauses = Implications

Important view during this course: clauses = implications.

$\{A, B\}$ (i.e., $A \vee B$)

- $\neg A \rightarrow B$
- $\neg B \rightarrow A$

$\{A, B, C\}$ (i.e., $A \vee B \vee C$)

- $(\neg A \wedge \neg B) \rightarrow C$
- $(\neg A \wedge \neg C) \rightarrow B$
- ...

2-CNF = formula in CNF with clauses of size 2

2-SAT = decide satisfiability of formula in 2-CNF

Example

Is the following 2-CNF formula satisfiable?

$\{A, B\},$

$\{\neg B, C\},$

$\{\neg C, A\}$

$\{\neg A, \neg B\},$

$\{B, \neg A\},$

$\{C, \neg D\}$

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$\{C, \neg D\}$

Theorem

2-SAT can be solved in linear time.

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Proof.

Let φ be in 2-CNF. Construct a graph $G = (V, E)$ with

- $V = \{v \mid v \in \text{Atoms}(\varphi)\} \cup \{\neg v \mid v \in \text{Atoms}(\varphi)\}$
- $E = \{(\neg a, b) \mid \{a, b\} \in \varphi\}$

φ is satisfiable $\Leftrightarrow G$ has no cycle that contains both v and $\neg v$ for some v □

Encoding Graph Coloring

Given an undirected graph $G = (V, E)$, can it be colored by three colors (red, green, blue) so that no edge has endpoints of the same color?

Encoding

- variables v_r, v_g, v_b for each $v \in V$

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- coloring constraint $u_c \rightarrow \neg v_c$ for each edge $\{u, v\} \in E$ and each color $c \in \{r, g, b\} \equiv$ clause $\{\neg u_c, \neg v_c\}$

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- coloring constraint $u_c \rightarrow \neg v_c$ for each edge $\{u, v\} \in E$ and each color $c \in \{r, g, b\} \equiv$ clause $\{\neg u_c, \neg v_c\}$
- models of the formula \simeq valid colorings

Conversion to CNF: Tseitin encoding

Conversion to **equivalent** CNF can be exponential, but do we really need equivalence?

Definition

The formulas φ and ψ are **equisatisfiable** if both are satisfiable or both unsatisfiable.

Theorem

For each formula φ there exists an **equisatisfiable** formula φ^{CNF} with $\mathcal{O}(|\varphi|)$ clauses of size at most three.

Proof.

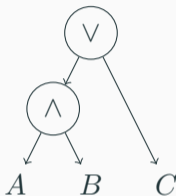
Tseitin encoding. □

Conversion to CNF: Tseitin encoding by example

$$\varphi = (A \wedge B) \vee C$$

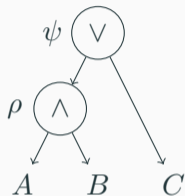
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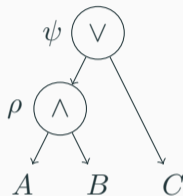
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Conversion to CNF: Tseitin encoding by example

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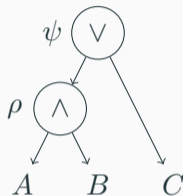
$$(A_\rho \leftrightarrow (A \wedge B)) \wedge$$

$$(A_\psi \leftrightarrow (A_\rho \vee C)) \wedge$$

$$A_\psi$$

Conversion to CNF: Tseitin encoding by example

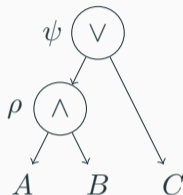
$$\varphi = (A \wedge B) \vee C$$



$$\begin{aligned} (A_\rho \leftrightarrow (A \wedge B)) \wedge & \quad (A_\rho \rightarrow (A \wedge B)) \wedge \\ (A_\rho \leftarrow (A \wedge B)) \wedge & \quad (A_\rho \leftarrow (A \wedge B)) \wedge \\ (A_\psi \leftrightarrow (A_\rho \vee C)) \wedge & \equiv (A_\psi \rightarrow (A_\rho \vee C)) \wedge \\ A_\psi & \quad (A_\psi \leftarrow (A_\rho \vee C)) \wedge \\ & \quad A_\psi \end{aligned}$$

Conversion to CNF: Tseitin encoding by example

$$\varphi = (A \wedge B) \vee C$$



$$\begin{aligned} (A_\rho \leftrightarrow (A \wedge B)) \wedge & & (A_\rho \rightarrow (A \wedge B)) \wedge & & \{\neg A_\rho, A\}, \{\neg A_\rho, B\}, \\ (A_\rho \leftarrow (A \wedge B)) \wedge & & (A_\rho \leftarrow (A \wedge B)) \wedge & & \{\neg A, \neg B, A_\rho\}, \\ (A_\psi \leftrightarrow (A_\rho \vee C)) \wedge & \equiv & (A_\psi \rightarrow (A_\rho \vee C)) \wedge & \equiv & \{\neg A_\psi, A_\rho, C\}, \\ (A_\psi \leftarrow (A_\rho \vee C)) \wedge & & (A_\psi \leftarrow (A_\rho \vee C)) \wedge & & \{\neg A_\rho, A_\psi\}, \{\neg C, A_\psi\}, \\ A_\psi & & A_\psi & & \{A_\psi\} \end{aligned}$$

Conversion to CNF: Tseitin encoding

1. Create a new **Tseitin variable** A_ψ for each subformula of φ .
2. Add unit clause $\{A_\varphi\}$.
3. Define semantics of the new Tseitin variables A_ψ :

ψ	definition of A_ψ	added clauses
$\rho_1 \vee \rho_2$	$A_\psi \rightarrow (A_{\rho_1} \vee A_{\rho_2})$ $A_\psi \leftarrow (A_{\rho_1} \vee A_{\rho_2})$	$\{\neg A_\psi, A_{\rho_1}, A_{\rho_2}\}$ $\{\neg A_{\rho_1}, A_\psi\}, \{\neg A_{\rho_2}, A_\psi\}$
$\rho_1 \wedge \rho_2$	$A_\psi \rightarrow (A_{\rho_1} \wedge A_{\rho_2})$ $A_\psi \leftarrow (A_{\rho_1} \wedge A_{\rho_2})$	$\{\neg A_\psi, A_{\rho_1}\}, \{\neg A_\psi, A_{\rho_2}\}$ $\{\neg A_{\rho_1}, \neg A_{\rho_2}, A_\psi\}$
$\neg \rho$	$A_\psi \rightarrow \neg A_\rho$ $A_\psi \leftarrow \neg A_\rho$	$\{\neg A_\psi, \neg A_\rho\}$ $\{A_\rho, A_\psi\}$
$\rho_1 \leftrightarrow \rho_2$	$A_\psi \rightarrow (A_{\rho_1} \leftrightarrow A_{\rho_2})$ $A_\psi \leftarrow (A_{\rho_1} \leftrightarrow A_{\rho_2})$	$\{\neg A_\psi, \neg A_{\rho_1}, A_{\rho_2}\}, \{\neg A_\psi, A_{\rho_1}, \neg A_{\rho_2}\}$ $\{\neg A_{\rho_1}, \neg A_{\rho_2}, A_\psi\}, \{A_{\rho_1}, A_{\rho_2}, A_\psi\}$

Tseitin encoding

- often used in practice
- also works for DAG representation of formulas: one Tseitin variable for each node in the DAG
- transforming to increase shared subexpression helps $(B \wedge A) \rightsquigarrow (A \wedge B)$
- additional preprocessing helps: $(A \vee (B \vee C)) \rightsquigarrow (A \vee B \vee C)$ and then encode $A_\varphi \leftrightarrow (A \vee B \vee C)$ as **one Tseitin variable** and **four implications**
- some of the clauses are not needed (Plaisted-Greenbaum)

Classical SAT algorithms

- propositional resolution
- Davis-Putnam-Logemann-Loveland algorithm (DPLL)