# Propositional Satisfiability 

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## (1) Models of a formula

## (2) Satisfiability

## Models of a formula

- We can define $M(\Gamma)=$ the set of models of a theory (or formula) $\Gamma$. Example: $M(a \vee b)=\{\{a, b\},\{a\},\{b\}\}$


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\begin{aligned}
M(\alpha \vee \beta) & =M(\alpha) \cup M(\beta) \\
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- Two formulas $\alpha, \beta$ are equivalent if $M(\alpha)=M(\beta)$ (same models)


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- Does this formula $\alpha$ always exist?
- We will see a method (minterms) to obtain a minimal representation


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## Propositional Logic: Semantics

## Theorem

$\vDash \alpha \rightarrow \beta$ is equivalent to $\alpha=\beta$.

Definition (Weaker/stronger formula)
When $\models \alpha \rightarrow \beta$, or just $M(\alpha) \subseteq M(\beta)$, we say that $\alpha$ is stronger than $\beta$ (or $\beta$ is weaker $\alpha$ ).

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(2) Satisfiability

## Satisfiability

## Definition (SAT decision problem)

Decision problem $\operatorname{SAT}(\alpha) \in\{y e s$, no $\}$ checks whether a formula $\alpha$ has some model. That is: $\operatorname{SAT}(\alpha)=$ yes iff $M(\alpha) \neq \emptyset$.

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- Nowadays, SAT is an outstanding state-of-the-art research area for search algorithms. There exist many efficient tools and commercial applications. See www.satlive.com
- SAT keypoint: instead of designing an ad hoc search algorithm, encode the problem into propositional logic and use SAT as a backend.


## Exercise

- Programming the satisfaction relation $\vDash$ in Prolog:

```
:- op (210, yfx, &).
:- op (220, yfx, v).
:- op(1060, yfx, <->).
sat(_I, false) :- !, fail.
sat(_I, true) :- !.
sat(I, P) :- atom(P),!,member(P,I),!.
sat(I, -A) :- \+ sat(I, A).
sat(I, A & B) :- sat(I, A), sat(I, B).
sat(I, A v B) :- sat(I,A),! ; sat(I, B).
sat(I, A -> B) :- sat(I,-A v B).
sat(I, A <-> B) :- sat(I,(A -> B)&(B -> A)).
```


## Exercise

- Testing whether a formula is a tautology:

```
tautology(S, F) :- \+ (subset(S,I), \+ sat(I,F)).
subset([],[]) :- !.
subset([X | Xs],S) :- subset(Xs,S).
subset([X | Xs],[X | S]) :- subset(Xs,S).
```


## Conjunctive Normal Form

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- This is a conjunction of clauses $=$ disjunctions of literals. Example: $\left(p_{1} \vee \neg p_{2}\right) \wedge\left(\neg p_{3} \vee p_{1} \vee p_{2}\right) \wedge \neg p_{1} \wedge\left(p_{2} \vee p_{4}\right)$


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- This is represented as a text file in DIMACS format. For instance, the formula above becomes

```
p cnf 34 3 variables, 4 clauses
1 -2 0 0 marks the end of a clause
-3}112% 
-1 0
2 4 0
```


## Reduction to CNF

- Reduction to CNF: several methods can be used (for instance, semantic tableaux)


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- Reducing a formula $\alpha$ to CNF causes an exponential cost
- Distributivity blows up

$$
\left(p_{1} \wedge q_{1}\right) \vee\left(p_{2} \wedge q_{2}\right) \vee \cdots \vee\left(p_{n} \wedge q_{n}\right)
$$

$2^{n}$ disjunctions depending on whether we take $p$ or $q$ for each $i$

Reducing $\varphi \longrightarrow \operatorname{CNF}(\varphi)$ in classical logic is NP-hard.

## Reduction to CNF (with auxiliary atoms)

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- Key idea: introduce auxiliary variables per each non-atomic subformula, then add equivalences to fix their truth


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\begin{gathered}
a_{1} \vee a_{2} \vee \cdots \vee a_{n} \\
a_{i} \leftrightarrow p_{i} \wedge q_{i}
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$$
\begin{array}{ll} 
& a_{1} \vee a_{2} \vee \cdots \vee a_{n} \\
\neg a_{i} \vee p_{i} & \neg a_{i} \vee q_{i} \quad a_{i} \vee \neg p_{i} \vee \neg q_{i}
\end{array}
$$

$1+3 \cdot n$ clauses. We have $n$ new atoms: we would hide in models

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\begin{array}{cc} 
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a_{1} \rightarrow q & a_{1} \rightarrow r
\end{array} \quad q \wedge r \rightarrow a_{1},
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& \neg p \quad \neg a_{1} & a_{2} \\
\neg a_{1} \vee q & \neg a_{1} \vee r & \neg q \vee \neg r \vee a_{1} \\
\neg a_{2} \vee p & \neg a_{2} \vee \neg r & \neg p \vee \neg r \vee a_{2}
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## SAT solvers

Basic Methods: (we will see them in detail later)

- DPLL (Davis-Putnam-Logemann-Loveland)
- Backtracking algorithm: picks some atom $p$ and tries two branches: one with $p=$ true, one with $p=$ false.


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- CDCL (Conflict-Driven Conflict Learning)
- Maintains an implication graph (each node is a literal, each arrow an implication)
- When an inconsistent assignment is reached, it extracts from the graph a new clause (reflecting the conflict)
- back jump: it backtracks several steps backwards to the first-assigned variable involved in the conflict

