

CS:4420 Artificial Intelligence

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Propositional Logic

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Readings

- Chap. 7 of [Russell and Norvig, 3rd Edition]

Knowledge Representation and Logic

The field of **Mathematical Logic** provides powerful, formal knowledge representation languages and inference systems to build reasoning agents

We will consider two languages, and associated inference systems, from mathematical logic:

- Propositional Logic
- First-order Logic

Logics

A **logic** is a triple $\langle \mathcal{L}, \mathcal{S}, \mathcal{R} \rangle$ where

- \mathcal{L} , the logic's **language**, is a class of sentences described by a formal grammar
- \mathcal{S} , the logic's **semantics** is a formal specification of how to assign *meaning* in the “real world” to the elements of \mathcal{L}
- \mathcal{R} , the logic's **inference system**, is a set of formal derivation *rules* over \mathcal{L}

There are **several** logics: propositional, first-order, higher-order, modal, temporal, intuitionistic, linear, equational, non-monotonic, fuzzy, ...

We will concentrate on **propositional logic** and **first-order logic**

Propositional Logic

Each sentence is made of

- propositional variables (A, B, \dots, P, Q, \dots)
- logical constants (**True, False**)
- logical connectives ($\wedge, \vee, \Rightarrow, \dots$)

Every propositional variable stands for a basic **fact**

Examples: *I'm hungry, Apples are red, Joe and Jill are married*

Propositional Logic

Ontological Commitments

Propositional Logic is about **facts** in the world that are either true or false, nothing else

Semantics of Propositional Logic

Since each propositional variable stands for a fact about the world, its meaning ranges over the Boolean values $\{true, false\}$

Note: Do not confuse

- *true, false*, which are values (i.e., semantical entities) here with
- **True, False**, which are logical constants (i.e., symbols of the language)

Propositional Logic

The Language

- Each propositional variable (A, B, \dots, P, Q, \dots) is a sentence
- Each logical constant (**True, False**) is a sentence
- If φ and ψ are sentences, all of the following are also sentences

(φ) $\neg\varphi$ $\varphi \wedge \psi$ $\varphi \vee \psi$ $\varphi \Rightarrow \psi$ $\varphi \Leftrightarrow \psi$

- Nothing else is a sentence

The Language of Propositional Logic

More formally, it is the language generated by the following grammar

Symbols:

- Propositional variables: A, B, \dots, P, Q, \dots
- Logical constants:

True	(true)	\wedge	(and)	\Rightarrow	(implies)	\neg	(not)
False	(false)	\vee	(or)	\Leftrightarrow	(equivalent)		

Grammar Rules:

Sentence ::= *AtomicS* | *ComplexS*

AtomicS ::= **True** | **False** | A | B | ... | P | Q | ...

ComplexS ::= (*Sentence*) | *Sentence* *Connective* *Sentence* | \neg *Sentence*

Connective ::= \wedge | \vee | \Rightarrow | \Leftrightarrow

Wumpus world sentences

Let $P_{i,j}$ denote that there is a pit in position (i, j)

Let $B_{i,j}$ denote that there is a breeze in position (i, j)

“There is no pit in the initial position but there is one in $(2, 2)$ ”

$$\neg P_{1,1} \wedge P_{2,2}$$

Wumpus world sentences

Let $P_{i,j}$ denote that there is a pit in position (i, j)

Let $B_{i,j}$ denote that there is a breeze in position (i, j)

“There is no pit in the initial position but there is one in $(2, 2)$ ”

$$\neg P_{1,1} \wedge P_{2,2}$$

“A square is breezy **if and only if** there is an adjacent pit”

$$B_{1,1} \iff (P_{2,1} \vee P_{1,2})$$

$$B_{2,2} \iff (P_{2,1} \vee P_{3,2} \vee P_{2,1} \vee P_{1,2})$$

$$\vdots \quad \vdots \quad \vdots$$

Semantics of Propositional Logic

The meaning of **True** is always *true*

The meaning of **False** is always *false*

The meaning of the other sentences depends on the meaning of the propositional variables

- **Base cases:** truth tables

P	Q	$\neg P$	$P \wedge Q$	$P \vee Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>
<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>
<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>

- **Non-base Cases:** given by reduction to the base cases

Ex: the meaning of $(P \vee Q) \wedge R$ is the same as the meaning of $A \wedge R$ where A has the same meaning as $P \vee Q$

Semantics of Propositional Logic

An assignment of Boolean values to the propositional variables of a sentence is an **interpretation** of the sentence

P	H	$P \vee H$	$(P \vee H) \wedge \neg H$	$((P \vee H) \wedge \neg H) \Rightarrow P$
<i>False</i>	<i>False</i>	<i>False</i>	<i>False</i>	<i>True</i>
<i>False</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>
<i>True</i>	<i>False</i>	<i>True</i>	<i>True</i>	<i>True</i>
<i>True</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>

Interpretations: $\{P \mapsto \text{false}, H \mapsto \text{false}\}, \{P \mapsto \text{false}, H \mapsto \text{true}\}, \dots$

An interpretation is a **model** of a sentence φ if it makes the sentence true

Note: The semantics of Propositional Logic is **compositional** — the meaning of a sentence is defined recursively in terms of the meaning of the sentence's components

Semantics of Propositional Logic

The meaning of a sentence in general depends on its interpretation
Some sentences, however, have always the same meaning

P	H	$P \vee H$	$(P \vee H) \wedge \neg H$	$((P \vee H) \wedge \neg H) \Rightarrow P$
<i>False</i>	<i>False</i>	<i>False</i>	<i>False</i>	<i>True</i>
<i>False</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>
<i>True</i>	<i>False</i>	<i>True</i>	<i>True</i>	<i>True</i>
<i>True</i>	<i>True</i>	<i>True</i>	<i>False</i>	<i>True</i>

A sentence is

- **satisfiable** if it is true in **some** interpretation
- **unsatisfiable** if it is true in **no** interpretation
- **valid** if it is true in **every** possible interpretation
- **invalid** if it is false in **some** possible interpretation

A Warning

Disjunction

- $A \vee B$ is true when A or B or *or both* are true (inclusive or)
- $A \oplus B$ is sometimes used to mean “either A or B but not both” (exclusive or)

Implication

- $A \Rightarrow B$ does not require a causal connection between A and B
Ex: *Sky-is-blue* \Rightarrow *Snow-is-white*
- When A is false, $A \Rightarrow B$ is always true *regardless* of B
Ex: *Two-equals-four* \Rightarrow *Apples-are-red*
- Beware of negations in implications
Ex: *Is-a-female-bird* \Rightarrow *Lays-eggs*
 \neg *Is-a-female-bird* \Rightarrow \neg *lays-eggs*

Entailment in Propositional Logic

Given

- a set Γ of sentences and
- a sentence φ ,

we write

$$\Gamma \models \varphi$$

iff every interpretation that makes all sentences in Γ true makes φ also true

$\Gamma \models \varphi$ is read as “ Γ entails φ ” or “ φ logically follows from Γ ”

Entailment in Propositional Logic

Examples

$$\{A, A \Rightarrow B\} \models B$$

$$\{A\} \models A \vee B$$

$$\{A, B\} \models A \wedge B$$

$$\{\} \models A \vee \neg A$$

$$\{A\} \not\models A \wedge B$$

$$\{A \vee \neg A\} \not\models A$$

	A	B	$A \Rightarrow B$	$A \vee B$	$A \wedge B$	$A \vee \neg A$
1.	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>
2.	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>
3.	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>true</i>
4.	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>

Properties of Entailment

- $\Gamma \models \varphi$, for all $\varphi \in \Gamma$ (inclusion property of PL)
- if $\Gamma \models \varphi$, then $\Gamma' \models \varphi$ for all $\Gamma' \supseteq \Gamma$ (monotonicity of PL)
- φ is valid iff $\{\} \models \varphi$ (also written as $\models \varphi$)
- φ is unsatisfiable iff $\varphi \models \mathbf{False}$
- $\Gamma \models \varphi$ iff the set $\Gamma \cup \{\neg\varphi\}$ is unsatisfiable

Logical Equivalence

Two sentences φ_1 and φ_2 are **logically equivalent**, written

$$\varphi_1 \equiv \varphi_2$$

if $\varphi_1 \models \varphi_2$ and $\varphi_2 \models \varphi_1$

Note:

- $\varphi_1 \equiv \varphi_2$ if and only if every interpretation assigns the same Boolean value to φ_1 and φ_2
- Implication and equivalence (\Rightarrow , \Leftrightarrow), which are **syntactic entities**, are intimately related to entailment and logical equivalence (\models , \equiv), which are **semantic notions**:

$$\varphi_1 \models \varphi_2 \quad \text{iff} \quad \models \varphi_1 \Rightarrow \varphi_2$$

$$\varphi_1 \equiv \varphi_2 \quad \text{iff} \quad \models \varphi_1 \Leftrightarrow \varphi_2$$

Properties of Logical Connectives

- \wedge and \vee are **commutative**

$$\varphi_1 \wedge \varphi_2 \equiv \varphi_2 \wedge \varphi_1$$

$$\varphi_1 \vee \varphi_2 \equiv \varphi_2 \vee \varphi_1$$

- \wedge and \vee are *associative*

$$\varphi_1 \wedge (\varphi_2 \wedge \varphi_3) \equiv (\varphi_1 \wedge \varphi_2) \wedge \varphi_3$$

$$\varphi_1 \vee (\varphi_2 \vee \varphi_3) \equiv (\varphi_1 \vee \varphi_2) \vee \varphi_3$$

- \wedge and \vee are mutually *distributive*

$$\varphi_1 \wedge (\varphi_2 \vee \varphi_3) \equiv (\varphi_1 \wedge \varphi_2) \vee (\varphi_1 \wedge \varphi_3)$$

$$\varphi_1 \vee (\varphi_2 \wedge \varphi_3) \equiv (\varphi_1 \vee \varphi_2) \wedge (\varphi_1 \vee \varphi_3)$$

- \wedge and \vee are related by \neg (DeMorgan's Laws)

$$\neg(\varphi_1 \wedge \varphi_2) \equiv \neg\varphi_1 \vee \neg\varphi_2$$

$$\neg(\varphi_1 \vee \varphi_2) \equiv \neg\varphi_1 \wedge \neg\varphi_2$$

Properties of Logical Connectives

\wedge , \Rightarrow , and \Leftrightarrow are actually redundant:

$$\varphi_1 \wedge \varphi_2 \equiv \neg(\neg\varphi_1 \vee \neg\varphi_2)$$

$$\varphi_1 \Rightarrow \varphi_2 \equiv \neg\varphi_1 \vee \varphi_2$$

$$\varphi_1 \Leftrightarrow \varphi_2 \equiv (\varphi_1 \Rightarrow \varphi_2) \wedge (\varphi_2 \Rightarrow \varphi_1)$$

We keep them all mainly for convenience

Exercise Use the truth tables to verify all the logical equivalences seen so far

Inference Systems for Propositional Logic

An **inference system** \mathcal{I} for PL is a procedure that given a set $\Gamma = \{\alpha_1, \dots, \alpha_m\}$ of sentences and a sentence φ , may reply “yes”, “no”, or run forever

If \mathcal{I} replies positively to input (Γ, φ) , we say that Γ **derives** φ **in** \mathcal{I} (or, \mathcal{I} derives φ from Γ , or, φ derives from Γ in \mathcal{I}) and write

$$\Gamma \vdash_{\mathcal{I}} \varphi$$

Intuitively, \mathcal{I} should be such that it replies “yes” on input (Γ, φ) only if φ is in fact entailed by Γ

All These Fancy Symbols!

- $A \wedge B \Rightarrow C$ is a sentence, a bunch of **symbols** manipulated by an **inference system \mathcal{I}**
- $A \wedge B \models C$ is a mathematical abbreviation standing for the statement: “every interpretation that makes $A \wedge B$ true, makes C also true”
- $A \wedge B \vdash_{\mathcal{I}} C$ is a mathematical abbreviation standing for the statement: “ \mathcal{I} derives C from $A \wedge B$ ”

In other words,

- \Rightarrow is a formal symbol of the logic, which is used by the inference system
- \models is a shorthand we use to talk about the meaning of formal sentences
- $\vdash_{\mathcal{I}}$ is a shorthand we use to talk about the output of the inference system \mathcal{I}

All These Fancy Symbols!

The connective \Rightarrow and the shorthand \models are related as follows

The sentence $\varphi_1 \Rightarrow \varphi_2$ is valid (always true) if and only if $\varphi_1 \models \varphi_2$

Example: $A \Rightarrow (A \vee B)$ is valid and $A \models (A \vee B)$

	A	B	$A \vee B$	$A \Rightarrow (A \vee B)$
1.	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>
2.	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>
3.	<u><i>true</i></u>	<i>false</i>	<i>true</i>	<i>true</i>
4.	<u><i>true</i></u>	<i>true</i>	<i>true</i>	<i>true</i>

All These Fancy Symbols!

The shorthands \models and $\vdash_{\mathcal{I}}$ are related as follows

- A **sound** inference system can derive **only** sentences that logically follow from a given set of sentences:

$$\text{if } \Gamma \vdash_{\mathcal{I}} \varphi \text{ then } \Gamma \models \varphi$$

- A **complete** inference system can derive **all** sentences that logically follow from a given set of sentences:

$$\text{if } \Gamma \models \varphi \text{ then } \Gamma \vdash_{\mathcal{I}} \varphi$$

Inference systems for PL

Divided into (roughly) two kinds:

Rule-based

- Sound generation of new sentences from old
- **Proof** = a sequence of inference rule applications
Can use inference rules as operators as in a standard search procedures
- Typically require translation of sentences into some **normal form**

Model-based

- Truth table enumeration (always exponential in n)
- Improved backtracking, e.g., Davis–Putnam–Logemann–Loveland
- Heuristic search in model space (incomplete)
e.g., min-conflicts-like hill-climbing algorithms

Truth table enumeration

The proof system \mathcal{TT} is specified as follows:

$\{\alpha_1, \dots, \alpha_m\} \vdash_{\mathcal{TT}} \varphi$ iff all the values in the truth table of $(\alpha_1 \wedge \dots \wedge \alpha_m) \Rightarrow \varphi$ are true

Inference by Truth Tables

- The truth-tables-based inference system is sound:

$\alpha_1, \dots, \alpha_m \vdash_{\mathcal{T}\mathcal{T}} \varphi$ implies truth table of $(\alpha_1 \wedge \dots \wedge \alpha_m) \Rightarrow \varphi$ all true
implies $(\alpha_1 \wedge \dots \wedge \alpha_m) \Rightarrow \varphi$ is valid
implies $\models (\alpha_1 \wedge \dots \wedge \alpha_m) \Rightarrow \varphi$
implies $(\alpha_1 \wedge \dots \wedge \alpha_m) \models \varphi$
implies $\alpha_1, \dots, \alpha_m \models \varphi$

- It is also complete (exercise: prove it)
- Its time complexity is $O(2^n)$ where n is the number of propositional variables in $\alpha_1, \dots, \alpha_m, \varphi$
- We cannot hope to do better in general because the dual problem: determining the satisfiability of a sentence, is NP-complete
- However, really hard cases of propositional inference are somewhat rare in practice

Rule-Based Inference in PL

An inference system in Propositional Logic can also be specified as a set \mathcal{R} of inference (or derivation) rules

- Each rule is just a *pattern* premises/conclusion
- A rule **applies** to Γ and **derives** φ if
 - some of the sentences in Γ match with the premises of the rule and
 - φ matches with the conclusion
- A rule is **sound** if the set of its premises entails its conclusion

Some Inference Rules

- And-Introduction

$$\frac{\alpha \quad \beta}{\alpha \wedge \beta}$$

- And-Elimination

$$\frac{\alpha \wedge \beta}{\alpha}$$

$$\frac{\alpha \wedge \beta}{\beta}$$

- Or-Introduction

$$\frac{\alpha}{\alpha \vee \beta}$$

$$\frac{\alpha}{\beta \vee \alpha}$$

Some Inference Rules (cont'd)

- Implication-Elimination (aka Modus Ponens)

$$\frac{\alpha \Rightarrow \beta \quad \alpha}{\beta}$$

- Unit Resolution

$$\frac{\alpha \vee \beta \quad \neg\beta}{\alpha}$$

- Resolution

$$\frac{\alpha \vee \beta \quad \neg\beta \vee \gamma}{\alpha \vee \gamma} \quad \text{or, equivalently,}$$

$$\frac{\neg\alpha \Rightarrow \beta, \quad \beta \Rightarrow \gamma}{\neg\alpha \Rightarrow \gamma}$$

Some Inference Rules (cont'd)

- Double-Negation-Elimination

$$\frac{\neg\neg\alpha}{\alpha}$$

- False-Introduction

$$\frac{\alpha \quad \neg\alpha}{\mathbf{False}}$$

- False-Elimination

$$\frac{\mathbf{False}}{\beta}$$

Inference by Proof

We say there is a **proof** of φ from Γ in an inference system \mathcal{R} if we can derive φ by applying the rules of \mathcal{R} repeatedly to Γ and its derived sentences

Example: a proof of P from $\{(P \vee H) \wedge \neg H\}$

1. $(P \vee H) \wedge \neg H$ *by assumption*
2. $P \vee H$ *by \wedge -elimination applied to (1)*
3. $\neg H$ *by \wedge -elimination applied to (1)*
4. P *by unit resolution applied to (2),(3)*

We can represent a proof more visually as a **proof tree**:

Example:

$$\frac{\frac{(P \vee H) \wedge \neg H}{P \vee H} \quad \frac{(P \vee H) \wedge \neg H}{\neg H}}{P}$$

Rule-Based Inference in Propositional Logic

More precisely, there is a proof of φ from Γ in \mathcal{R} if

1. $\varphi \in \Gamma$ or,
2. there is a rule in \mathcal{R} that applies to Γ and produces φ or,
3. there is a proof of each $\varphi_1, \dots, \varphi_m$ from Γ in \mathcal{R} and a rule that applies to $\{\varphi_1, \dots, \varphi_m\}$ and produces φ

Then, the inference system \mathcal{R} is specified as follows:

$\Gamma \vdash_{\mathcal{R}} \varphi$ iff there is a proof of φ from Γ in \mathcal{R}

An Inference System \mathcal{R}

$$\frac{\alpha \quad \beta}{\alpha \wedge \beta}$$

$$\frac{\alpha}{\alpha \vee \beta}$$

$$\frac{\alpha}{\beta \vee \alpha}$$

$$\frac{\alpha \wedge \beta}{\alpha}$$

$$\frac{\alpha \wedge \beta}{\beta}$$

$$\frac{\alpha \Rightarrow \beta \quad \alpha}{\beta}$$

$$\frac{\alpha \vee \beta \quad \neg \beta}{\alpha}$$

$$\frac{\alpha \vee \beta \quad \neg \beta \vee \gamma}{\alpha \vee \gamma}$$

$$\frac{\neg \neg \alpha}{\alpha}$$

$$\frac{\alpha \quad \neg \alpha}{\mathbf{False}}$$

$$\frac{\mathbf{False}}{\beta}$$

Soundness of \mathcal{R}

The given system \mathcal{R} is sound because all of its rules are

Example: the Resolution rule
$$\frac{\alpha \vee \beta, \quad \neg\beta \vee \gamma}{\alpha \vee \gamma}$$

	α	β	γ	$\neg\beta$	$\alpha \vee \beta$	$\neg\beta \vee \gamma$	$\alpha \vee \gamma$
1.	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>false</i>
2.	<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>
3.	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>
4.	<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<u><i>true</i></u>	<u><i>true</i></u>	<i>true</i>
5.	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<u><i>true</i></u>	<u><i>true</i></u>	<i>true</i>
6.	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<u><i>true</i></u>	<u><i>true</i></u>	<i>true</i>
7.	<i>true</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>true</i>
8.	<i>true</i>	<i>true</i>	<i>true</i>	<i>false</i>	<u><i>true</i></u>	<u><i>true</i></u>	<i>true</i>

All the interpretations that satisfy both $\alpha \vee \beta$ and $\neg\beta \vee \gamma$ (4,5,6,8) satisfy $\alpha \vee \gamma$ as well

Soundness of \mathcal{R}

The given system \mathcal{R} is sound because all of its rules are

Exercise: prove that the other inference rules are sound as well

Is \mathcal{R} also complete?

Another Inference System: Resolution

The previous inference system is difficult to implement efficiently in practice

Resolution is a more efficient inference system (in practice)

The main reason is that

- it requires formulas to be in a **special form**: **CNF**
- it has only **one inference rule**

Resolution

Literal: prop. symbol (P) or negated prop. symbol ($\neg P$)

Clause: set of literals $\{l_1, \dots, l_k\}$ (understood as $l_1 \vee \dots \vee l_k$)

Conjunctive Normal Form: set of clauses $\{C_1, \dots, C_n\}$ (understood as $C_1 \wedge \dots \wedge C_n$)

Resolution

Literal: prop. symbol (P) or negated prop. symbol ($\neg P$)

Clause: set of literals $\{l_1, \dots, l_k\}$ (understood as $l_1 \vee \dots \vee l_k$)

Conjunctive Normal Form: set of clauses $\{C_1, \dots, C_n\}$ (understood as $C_1 \wedge \dots \wedge C_n$)

Resolution rule for CNF:

$$\frac{l_1 \vee \dots \vee l_k \vee P \quad \neg P \vee m_1 \vee \dots \vee m_n}{l_1 \vee \dots \vee l_k \vee m_1 \vee \dots \vee m_n}$$

E.g.,

$$\frac{P \vee Q \quad \neg Q}{P}$$

$$\frac{P \vee Q \quad R \vee \neg Q \vee \neg S}{P \vee R \vee \neg S}$$

$$\frac{P \vee Q \quad \neg Q \vee P \vee R}{P \vee R}$$

Resolution

Literal: prop. symbol (P) or negated prop. symbol ($\neg P$)

Clause: set of literals $\{l_1, \dots, l_k\}$ (understood as $l_1 \vee \dots \vee l_k$)

Conjunctive Normal Form: set of clauses $\{C_1, \dots, C_n\}$ (understood as $C_1 \wedge \dots \wedge C_n$)

Resolution rule for CNF:

$$\frac{l_1 \vee \dots \vee l_k \vee P \quad \neg P \vee m_1 \vee \dots \vee m_n}{l_1 \vee \dots \vee l_k \vee m_1 \vee \dots \vee m_n}$$

E.g.,

$$\frac{P \vee Q \quad \neg Q}{P} \quad \frac{P \vee Q \quad R \vee \neg Q \vee \neg S}{P \vee R \vee \neg S} \quad \frac{P \vee Q \quad \neg Q \vee P \vee R}{P \vee R}$$

Resolution is sound and **complete** for CNF KBs

Conversion to CNF

Ex.: $A \Leftrightarrow (B \vee C)$

1. Eliminate \Leftrightarrow , replacing $\alpha \Leftrightarrow \beta$ with $(\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)$

$$(A \Rightarrow (B \vee C)) \wedge ((B \vee C) \Rightarrow A)$$

2. Eliminate \Rightarrow , replacing $\alpha \Rightarrow \beta$ with $\neg\alpha \vee \beta$

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

3. Move \neg inwards using de Morgan's rules and double-negation

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

4. Apply distributivity law (\vee over \wedge) and flatten

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

Resolution Procedure

Proof by contradiction: show $KB \models \alpha$ by showing $KB \wedge \neg\alpha$ unsatisfiable.

Do the latter by deriving **False** from CNF of $KB \wedge \neg\alpha$

Resolution Procedure

Proof by contradiction: show $KB \models \alpha$ by showing $KB \wedge \neg\alpha$ unsatisfiable.

Do the latter by deriving **False** from CNF of $KB \wedge \neg\alpha$

function PL-RESOLUTION(KB, α) **returns** *true* or *false*

clauses \leftarrow the set of clauses in the CNF representation of $KB \wedge \neg\alpha$

new \leftarrow { }

loop do

for each C_i, C_j **in** *clauses* **do**

resolvents \leftarrow PL-RESOLVE(C_i, C_j)

if *resolvents* contains the empty clause **then return** *true*

new \leftarrow *new* \cup *resolvents*

if *new* \subseteq *clauses* **then return** *false*

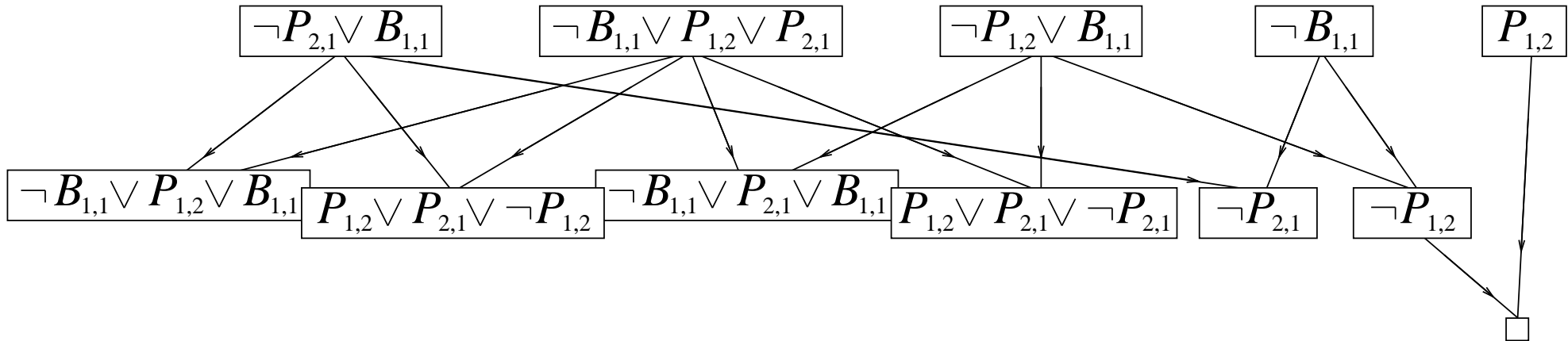
clauses \leftarrow *clauses* \cup *new*

Resolution Example

$$KB = \{ B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1}), \neg B_{1,1} \}$$

$$\alpha = \neg P_{1,2}$$

$$\text{CNF} = \{ \neg P_{1,2} \vee B_{1,1}, \neg B_{1,1} \vee P_{1,2} \vee P_{2,1}, \neg P_{1,2} \vee B_{1,1}, \neg B_{1,1}, P_{1,2} \}$$



Forward and backward chaining

Horn clause: prop. symbol (p) or implication $p_1 \wedge \dots \wedge p_n \Rightarrow p$

Horn Form: set of Horn clauses $\{C_1, \dots, C_n\}$ (understood as $C_1 \wedge \dots \wedge C_n$)

E.g., $\{ C, B \Rightarrow A, C \wedge D \Rightarrow B \}$

Modus Ponens for Horn Form

$$\frac{\alpha_1 \quad \dots \quad \alpha_n \quad \alpha_1 \wedge \dots \wedge \alpha_n \Rightarrow \beta}{\beta}$$

Sound and complete for Horn Form KBs

Can be used with forward chaining or backward chaining

These algorithms are very natural and run in linear time

Forward chaining

Idea:

Fire any rule whose premises are satisfied in the KB, add its conclusion to the KB, until query is found

Ex.: query is Q

KB is

$$P \implies Q$$

$$L \wedge M \implies P$$

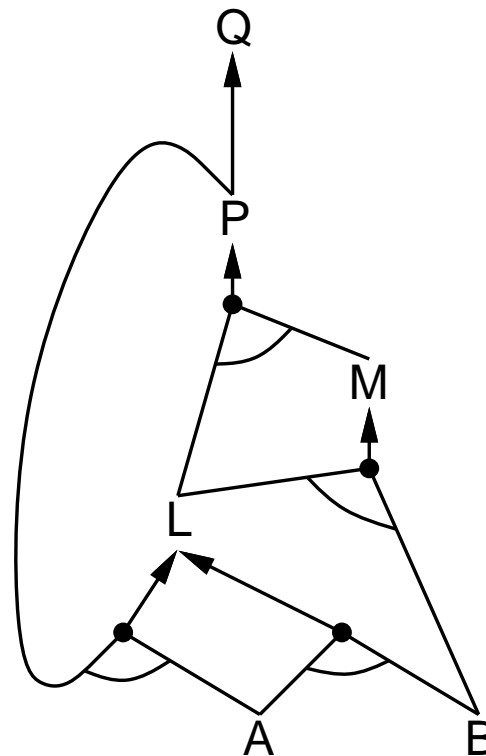
$$B \wedge L \implies M$$

$$A \wedge P \implies L$$

$$A \wedge B \implies L$$

A

B



Forward chaining algorithm

```
function PL-FC-ENTAILS?(KB, q) returns true or false  
  count, a table, indexed by clause, initially the number of premises  
  inferred, a table, indexed by symbol, each entry initially false  
  agenda, a list of symbols, initially the symbols known to be true  
  
  while agenda is not empty do  
     $p \leftarrow \text{POP}(\textit{agenda})$   
    unless inferred[p] do  
       $\textit{inferred}[p] \leftarrow \textit{true}$   
      for each Horn clause c in whose premise p appears do  
        decrement count[c]  
        if count[c] = 0 then do  
          if HEAD[c] = q then return true  
          PUSH(HEAD[c], agenda)  
  
  return false
```

Forward chaining example

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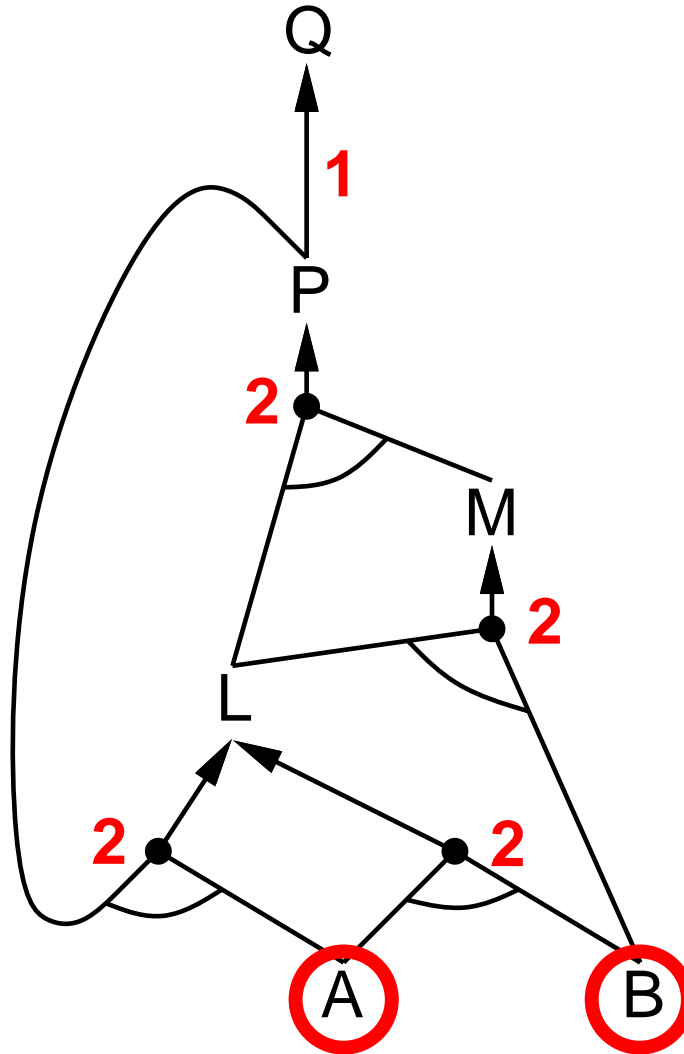
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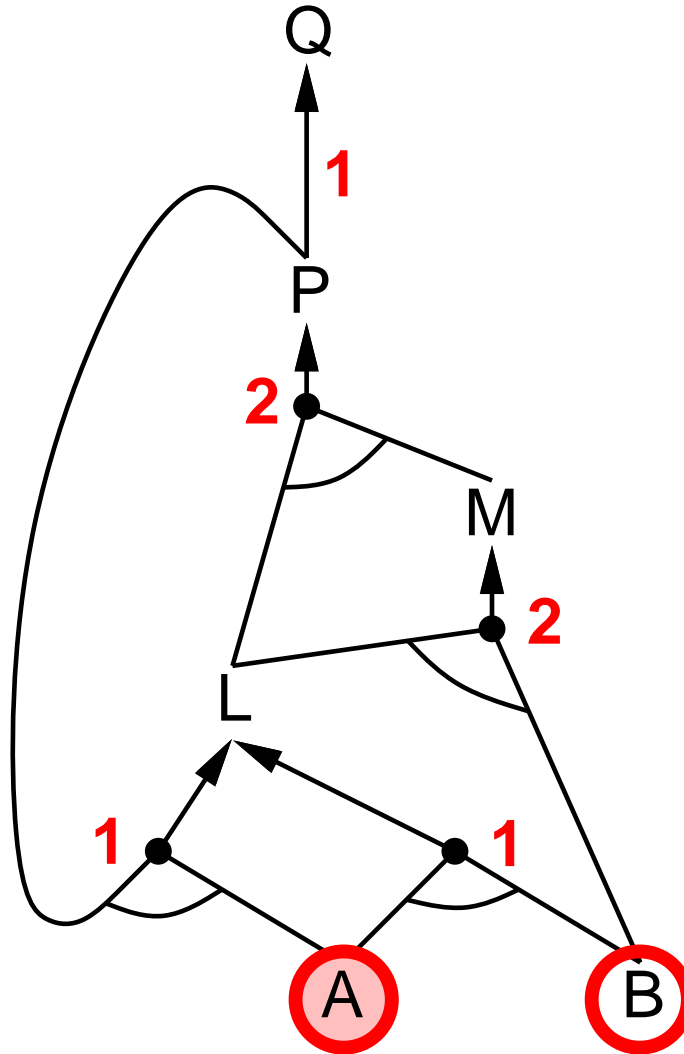
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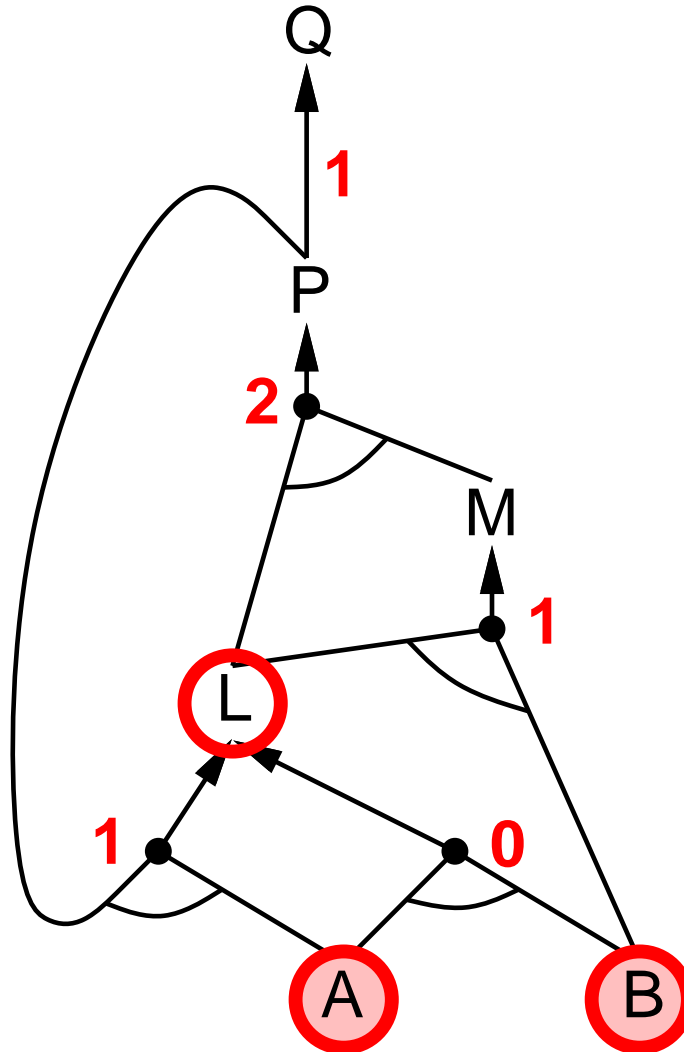
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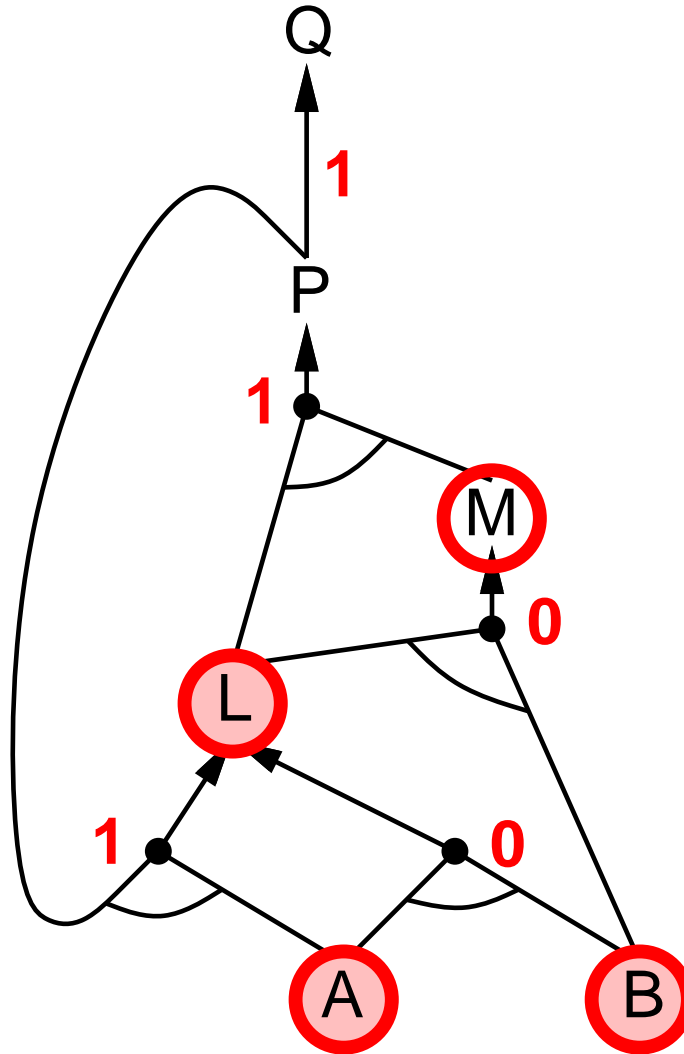
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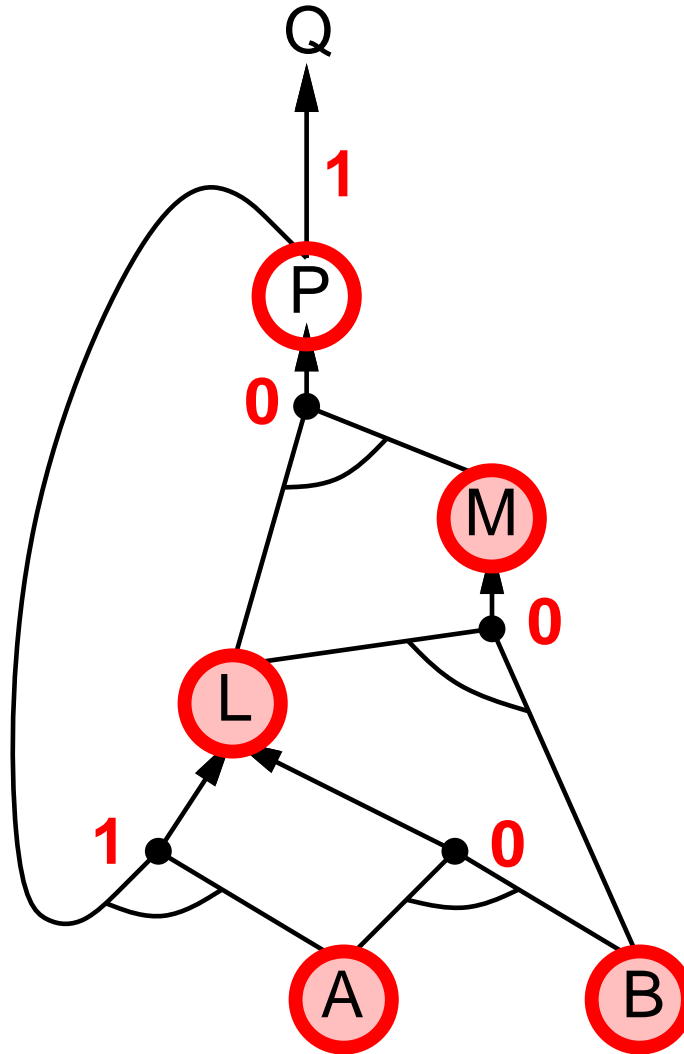
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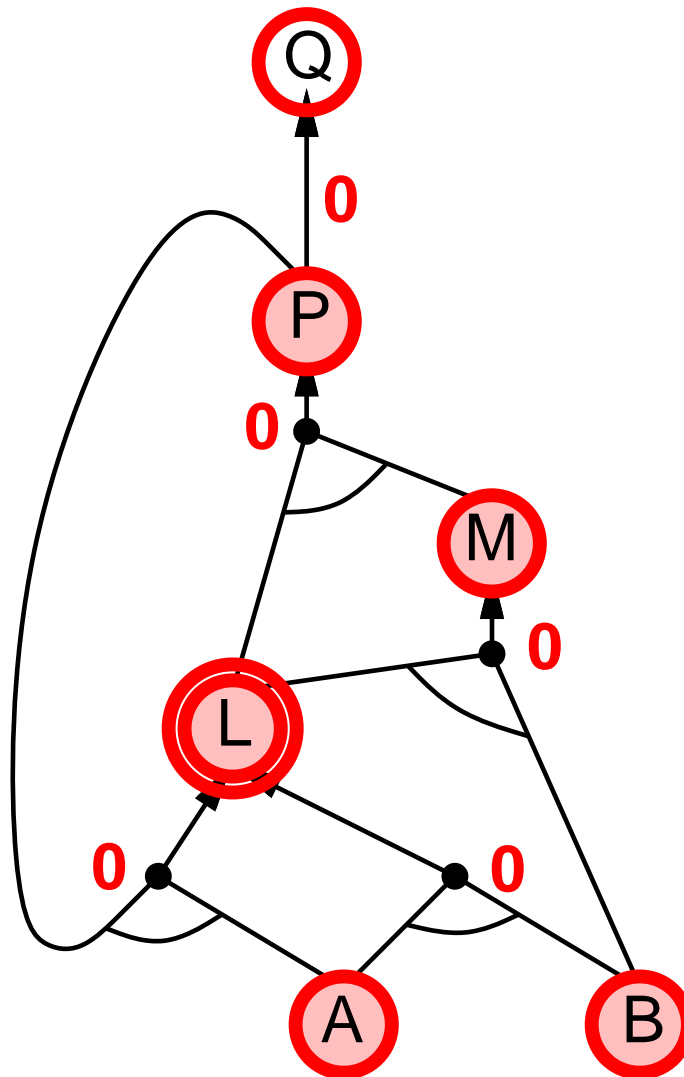
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A

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Proof of completeness (sketch)

FC derives every atomic sentence that is entailed by KB

1. FC reaches a **fixed point** where no new atomic sentences (prop. symbols) are inferred
2. Consider the final state as an interpretation m , assigning true to the inferred symbols and false to the other symbols

3. **Claim:** Every clause in the original KB is satisfied by m

Proof: Suppose a clause $a_1 \wedge \dots \wedge a_k \Rightarrow b$ is falsified by m

Then $a_1 \wedge \dots \wedge a_k$ is satisfied by m while b is not

But then b was not inferred, contradicting the assumption that the algorithm had reached a fixed point!

4. Hence m is a model of KB

5. If $KB \models q$ then q is true in **every** model of KB, including m

Backward chaining

Idea: work backwards from the query q

to infer q by BC,

check if q is known already, or

infer by BC all premises of some rule concluding q

Avoid loops: check if new subgoal is already on the goal stack

Avoid repeated work: check if new subgoal

1. has already been inferred, or
2. has already failed

Backward chaining example

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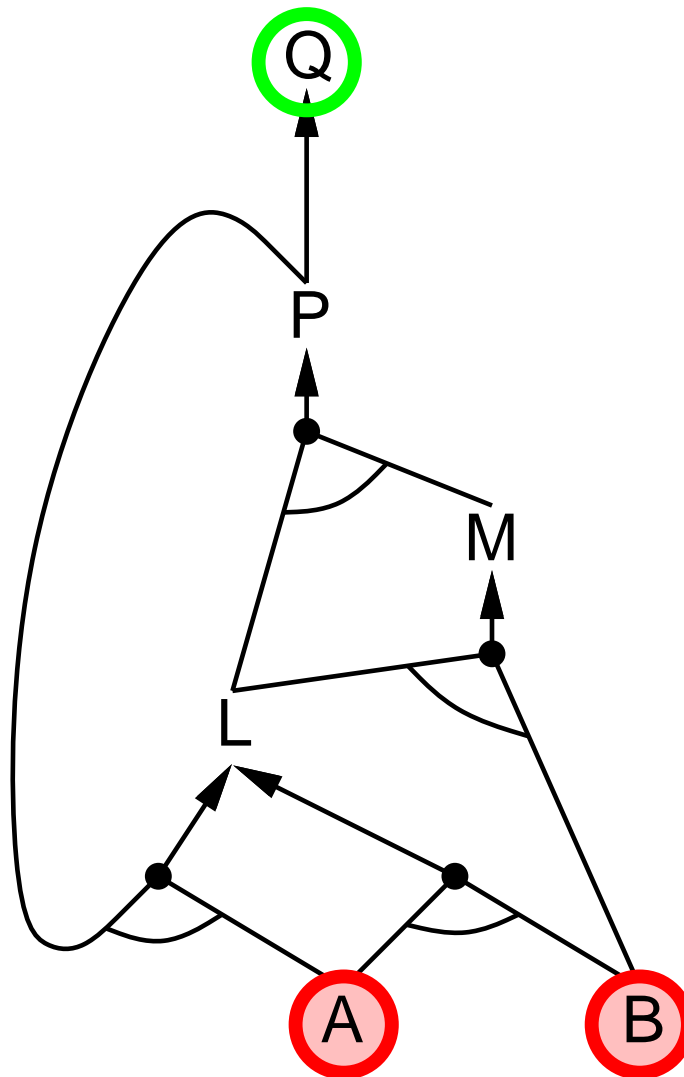
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Backward chaining example

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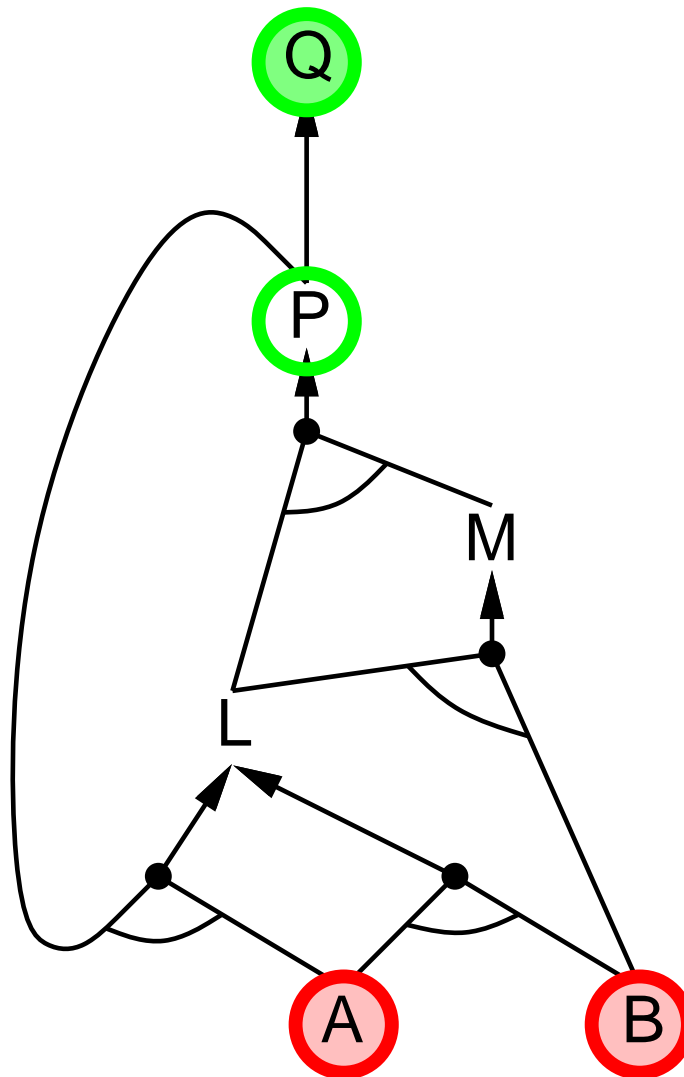
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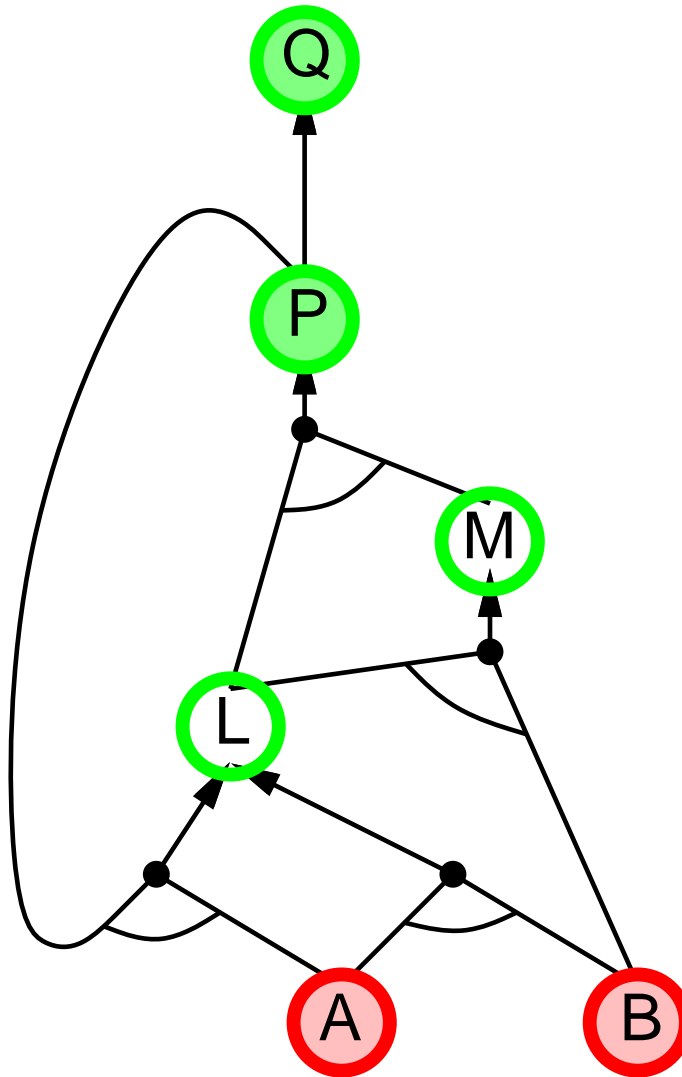
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Backward chaining example

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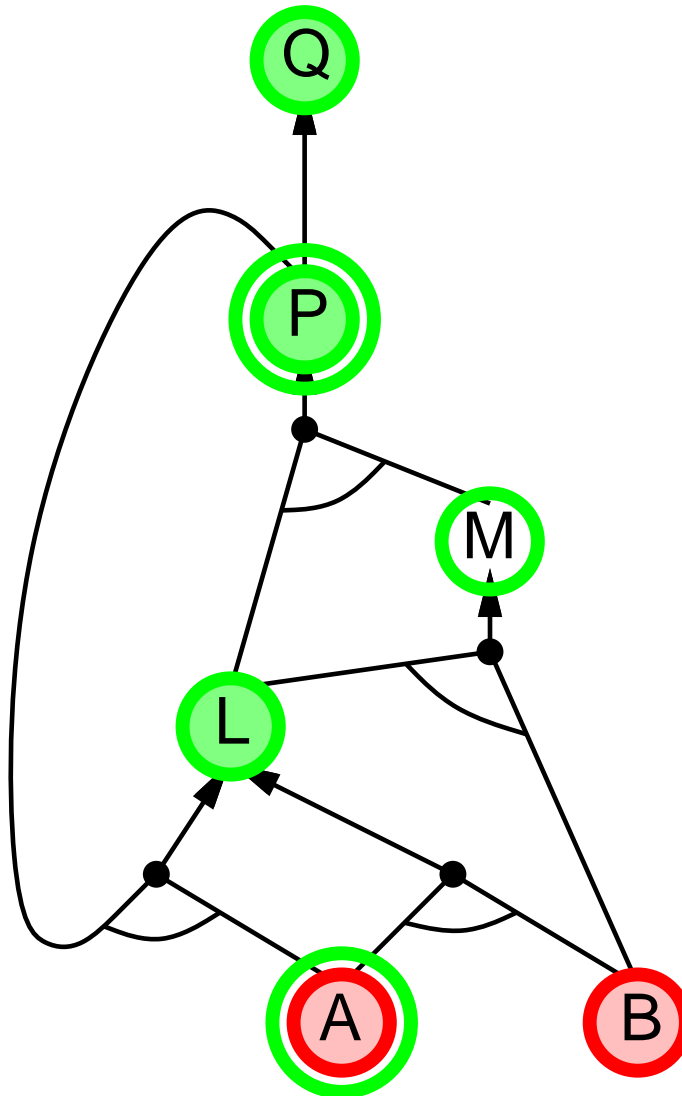
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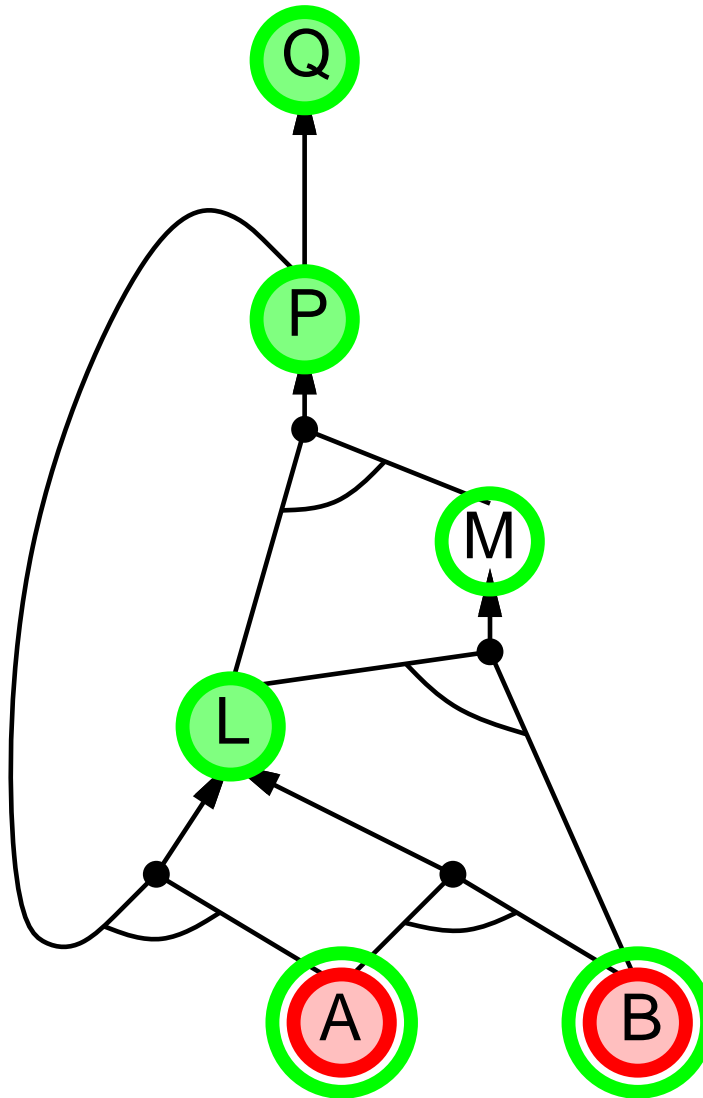
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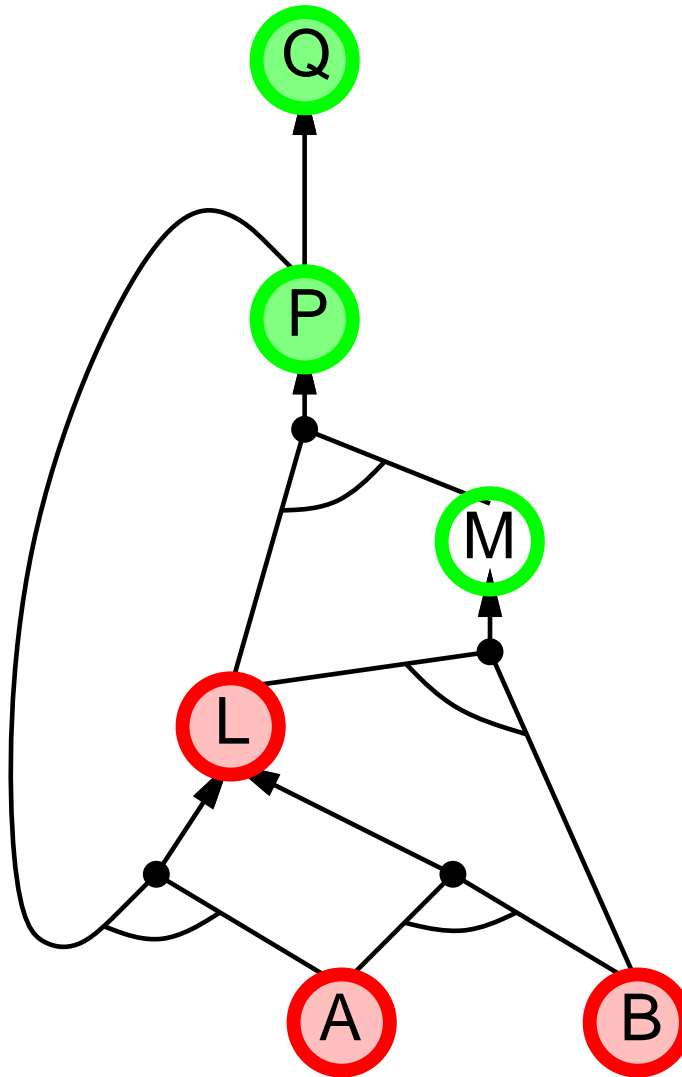
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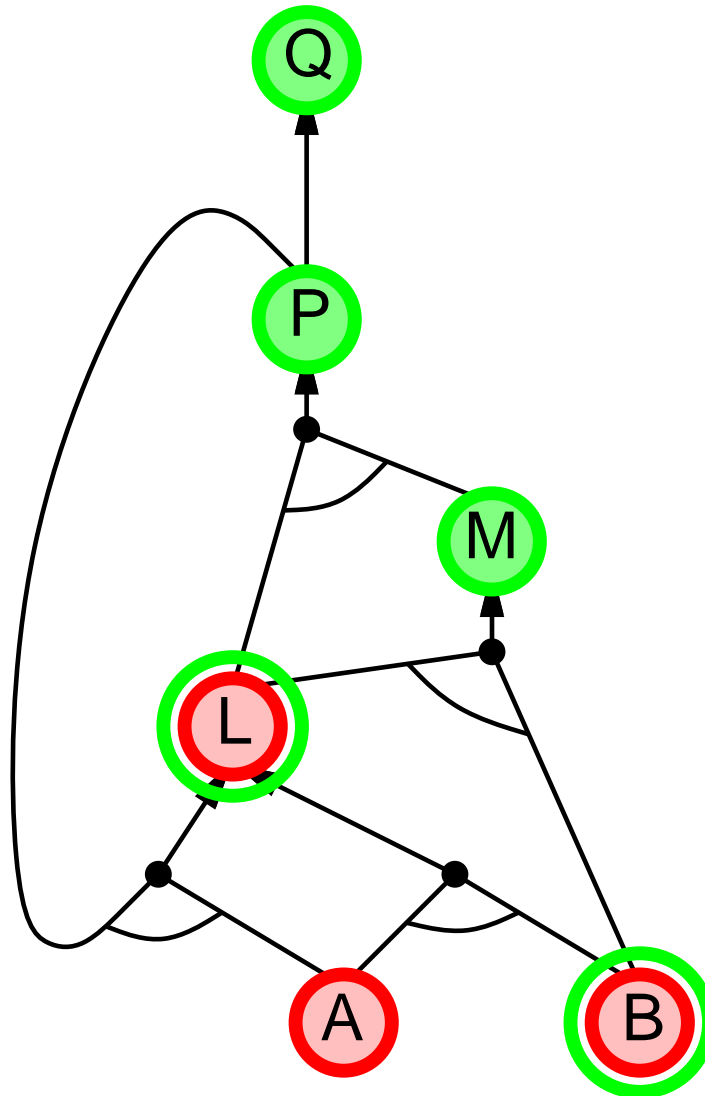
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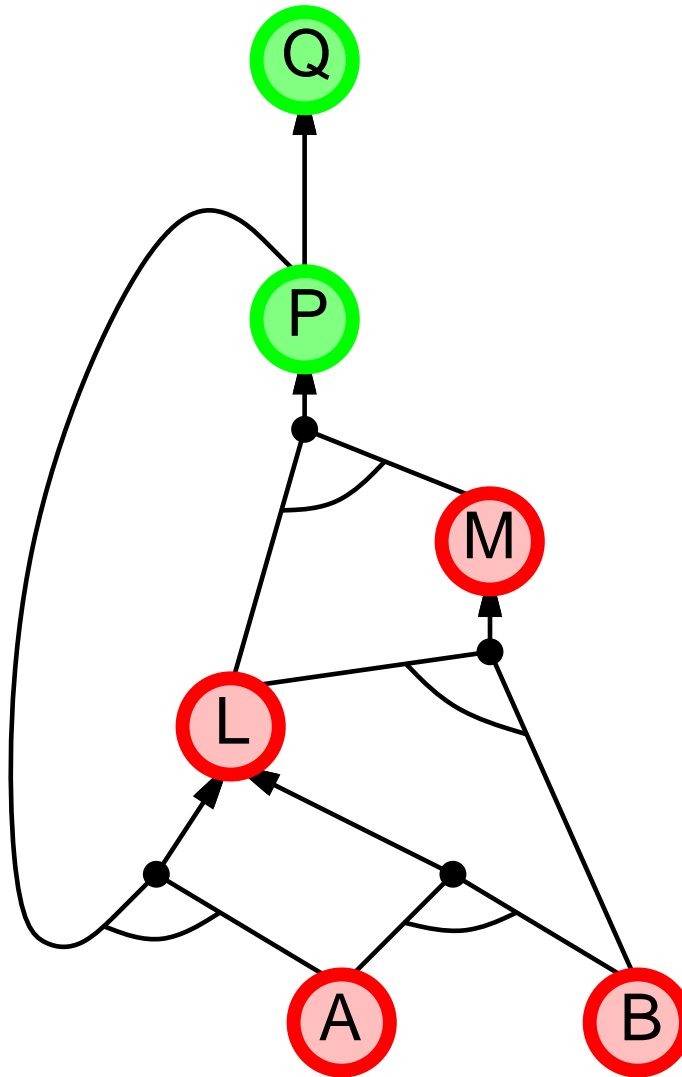
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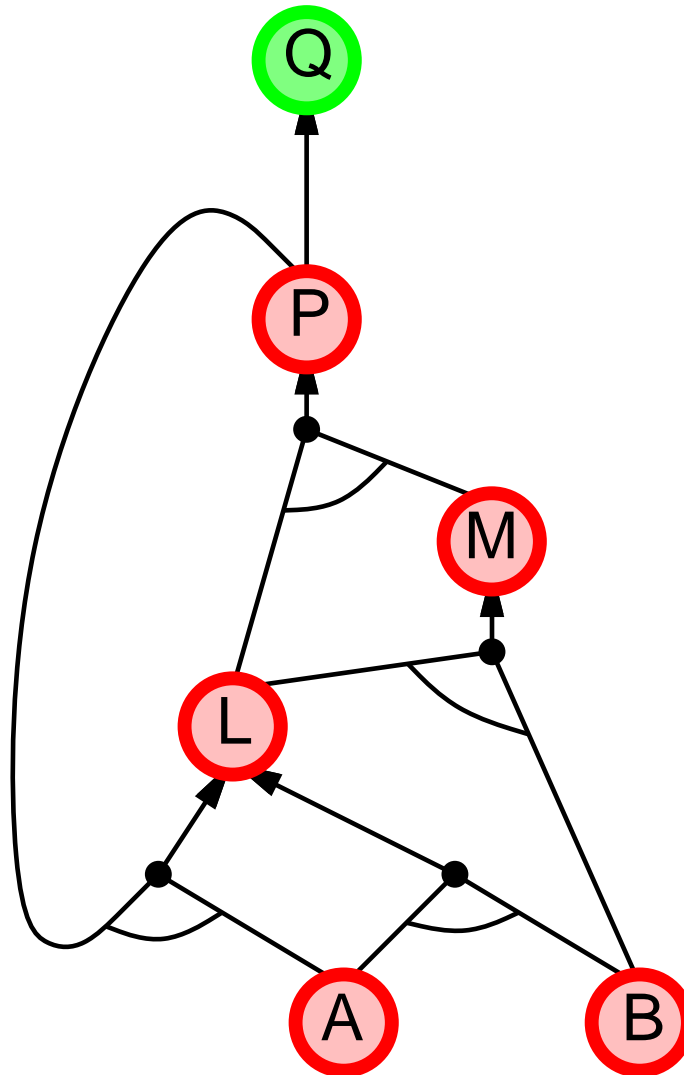
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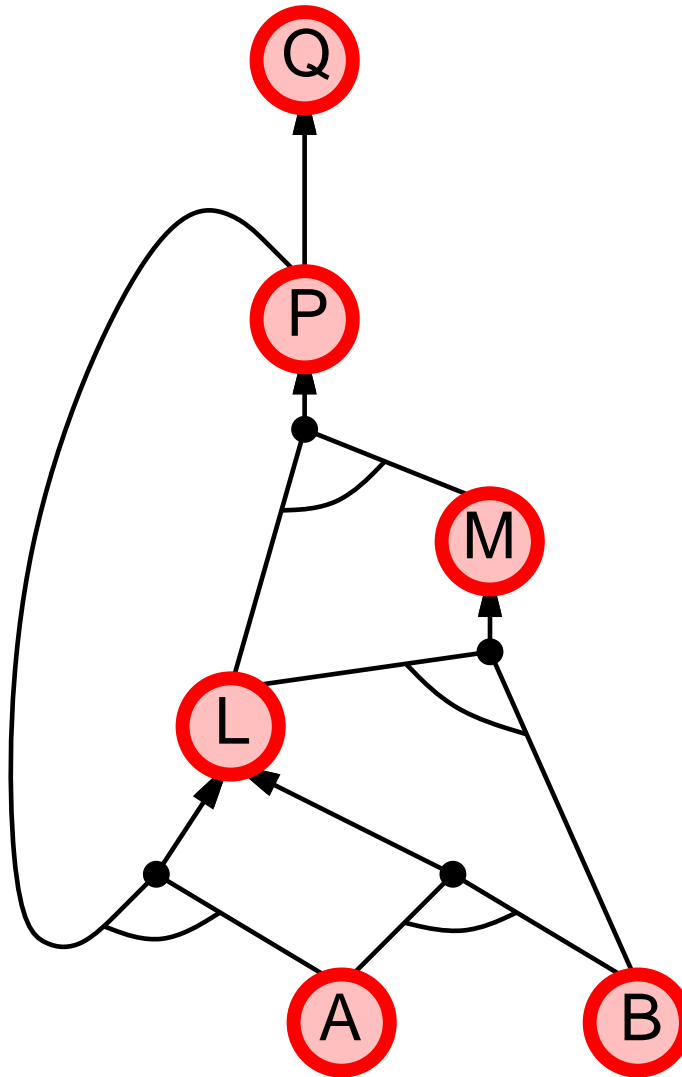
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Forward vs. Backward Chaining

FC is **data-driven**, cf. automatic, unconscious processing
e.g., object recognition, routine decisions

May do lots of work that is irrelevant to the goal

BC is **goal-driven**, appropriate for problem-solving,
e.g., Where are my keys? How do I get into a PhD program?

Complexity of BC can be **much smaller** than linear in size of KB

Model Checking methods

The most effective procedures for propositional satisfiability are based on CSP techniques

Variable domain: $\{true, false\}$

Constraints: sets of clauses

Heuristic Improvements:

- unit propagation
- variable and value ordering
- intelligent backtracking
- clause learning
- random restarts
- clever indexing
- subproblem decomposition

DPLL Procedure

function DPLL-SATISFIABLE?(*s*) **returns** *true* or *false*

inputs: *s*, a sentence in propositional logic

clauses \leftarrow the set of clauses in the CNF representation of *s*

symbols \leftarrow a list of the proposition symbols in *s*

return DPLL(*clauses*, *symbols*, [])

DPLL Procedure (cont.)

function DPLL(*clauses*, *symbols*, *model*) **returns** *true* or *false*

if every clause in *clauses* is satisfied by *model* **then return** *true*

if some clause in *clauses* is falsified by *model* **then return** *false*

P, *value* \leftarrow FIND-PURE-SYMBOL(*symbols*, *clauses*, *model*)

if *P* is non-null **then**

return DPLL(*clauses*, *symbols* - *P*, (*P* \mapsto *value*) :: *model*)

P, *value* \leftarrow FIND-UNIT-CLAUSE(*clauses*, *model*)

if *P* is non-null **then**

return DPLL(*clauses*, *symbols* - *P*, (*P* \mapsto *value*) :: *model*)

P \leftarrow FIRST(*symbols*)

rest \leftarrow REST(*symbols*)

return DPLL(*clauses*, *rest*, (*P* \mapsto *true*) :: *model*) **or**

 DPLL(*clauses*, *rest*, (*P* \mapsto *false*) :: *model*)

DPLL Exercise

Use DPLL to check the satisfiability of the following set of clauses

$$\begin{array}{lll} (1) \neg p_1 \vee p_2 & (2) \neg p_3 \vee p_4 & (3) \neg p_6 \vee \neg p_5 \vee \neg p_2 \\ (4) \neg p_5 \vee p_6 & (5) p_5 \vee p_7 & (6) \neg p_1 \vee p_5 \vee \neg p_7 \end{array}$$