

Logic: Top-down proof procedure, Datalog, Big Picture

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UBC CS 322 - Logic 4

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Textbook §5.2, §12.3

Lecture Overview

➡ Recap: Bottom-up proof procedure is sound and complete

- Top-down Proof Procedure
- Datalog
- Logics: Big Picture

Logical consequence and BU proofs

Definition (logical consequence)

If KB is a set of clauses and g is a conjunction of atoms, g is a **logical consequence** of KB, written $KB \models g$, if g is true in every model of KB

Example: $KB = \{h \leftarrow a, a, a \leftarrow c\}$. Then $KB \models ?$

	a	c	h	$h \leftarrow a$	a	$a \leftarrow c$	Model of KB
I_1	F	F	F	T	F	T	no
I_2	F	F	T	T	F	T	no
I_3	F	T	F	T	F	F	no
I_4	F	T	T	T	F	F	no
I_5	T	F	F	F	T	T	no
I_6	T	F	T	T	T	T	yes
I_7	T	T	F	F	T	T	no
I_8	T	T	T	T	T	T	yes

Which atoms are entailed?

Logical consequence and BU proofs

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I_5	T	F	F	F	T	T	no
I_6	T	F	T	T	T	T	yes
I_7	T	T	F	F	T	T	no
I_8	T	T	T	T	T	T	yes

Which atoms are entailed?

$KB \models a$ and
 $KB \models h$

Logical consequence and BU proofs

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Example: $KB = \{h \leftarrow a, a, a \leftarrow c\}$. Then $KB \models a$ and $KB \models h$.

$C := \{\}$;

repeat

select clause $h \leftarrow b_1 \wedge \dots \wedge b_m$ in KB
such that $b_i \in C$ for all i , and $h \notin C$;

$C := C \cup \{h\}$

until no more clauses can be selected. $KB \models_{BU} g$ if and only if $g \in C$

BU proof procedure

What does BU derive for the KB above?

Logical consequence and BU proofs

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If KB is a set of clauses and g is a conjunction of atoms, g is a **logical consequence** of KB, written $KB \models g$, if g is true in every model of KB

Example: $KB = \{h \leftarrow a, a, a \leftarrow c\}$. Then $KB \models a$ and $KB \models h$.

$C := \{\}$;

repeat

select clause $h \leftarrow b_1 \wedge \dots \wedge b_m$ in KB
such that $b_i \in C$ for all i , and $h \notin C$;

$C := C \cup \{h\}$

until no more clauses can be selected. $KB \vdash_{BU} g$ if and only if $g \in C$

BU proof procedure

What does BU derive for the KB above?

Trace: $\{a\}, \{a, h\}$. Thus $KB \vdash_{BU} a$ and $KB \vdash_{BU} h$.

Exactly the logical consequences!

Summary for bottom-up proof procedure BU

- Proved last time
 - BU is sound:
it derives **only** atoms that logically follow from KB
 - BU is complete:
it derives **all** atoms that logically follow from KB
- Together:
it derives **exactly** the atoms that logically follow from KB !
 - That's why the results for \models and \vdash_{BU} matched for the example above
- And, it is efficient!
 - Outer loop linear in the number of clauses in KB
 - Each clause is used maximally once by BU

Learning Goals Up To Here

- PDCL syntax & semantics
 - Verify whether a logical statement belongs to the language of propositional definite clauses
 - Verify whether an **interpretation** is a **model** of a PDCL KB.
 - Verify when a conjunction of atoms is a **logical consequence** of a knowledge base
- Bottom-up proof procedure
 - Define/read/write/trace/debug the Bottom Up (**BU**) proof procedure
 - Prove that the BU proof procedure is sound and complete

Lecture Overview

- Recap: Bottom-up proof procedure is sound and complete

Top-down Proof Procedure

- Datalog
- Logics: Big Picture

Bottom-up vs. Top-down

Bottom-up



g is proved if $g \in C$

When does BU look at the query g ?

In every loop iteration Never

At the end At the beginning

Bottom-up vs. Top-down

- **Key Idea of top-down:** search backward from a query g to determine if it can be derived from KB .

Bottom-up

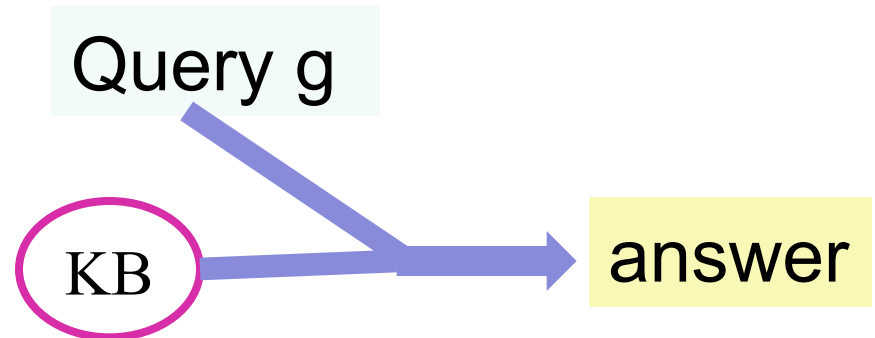


g is proved if $g \in C$

When does BU look at the query g ?

- Never
- It derives the same C regardless of the query

Top-down



We'll see how g is proved

TD performs a backward search starting at g

Top-down Ground Proof Procedure

Idea: search backward from a query

An **answer clause** is of the form: $\text{yes} \leftarrow a_1 \wedge \dots \wedge a_m$
where a_1, \dots, a_m are atoms

We express the query as an answer clause

– E.g. query $q_1 \wedge \dots \wedge q_k$ is expressed as $\text{yes} \leftarrow q_1 \wedge \dots \wedge q_k$

Basic operation: **SLD Resolution** of an answer clause

$$\text{yes} \leftarrow c_1 \wedge \dots \wedge c_{i-1} \wedge c_i \wedge c_{i+1} \dots \wedge c_m$$

on an atom c_i with another clause

$$c_i \leftarrow b_1 \wedge \dots \wedge b_p$$

yields the clause

$$\text{yes} \leftarrow c_1 \wedge \dots \wedge c_{i-1} \wedge b_1 \wedge \dots \wedge b_p \wedge c_{i+1} \dots \wedge c_m$$

Rules of derivation in top-down and bottom-up

Top-down:
SLD Resolution

$$\frac{\text{yes} \leftarrow c_1 \wedge \dots \wedge c_{i-1} \wedge c_i \wedge c_{i+1} \dots \wedge c_m \quad c_i \leftarrow b_1 \wedge \dots \wedge b_p}{\text{yes} \leftarrow c_1 \wedge \dots \wedge c_{i-1} \wedge b_1 \wedge \dots \wedge b_p \wedge c_{i+1} \dots \wedge c_m}$$

Bottom-up:
Generalized modus ponens

$$\frac{h \leftarrow b_1 \wedge \dots \wedge b_m \quad b_1 \wedge \dots \wedge b_m}{h}$$

Example for (successful) SLD derivation

$a \leftarrow b \wedge c.$ 1 $a \leftarrow e \wedge f.$ $b \leftarrow f \wedge k.$
 $c \leftarrow e.$ $d \leftarrow k$ 3 $e.$
 $f \leftarrow j \wedge e.$ 2 $f.$ $j \leftarrow c.$

Query: ?a

$\gamma_0: \text{yes} \leftarrow a$

$\gamma_1: \text{yes} \leftarrow e \wedge f$

$\gamma_2: \text{yes} \leftarrow e$

$\gamma_3: \text{yes} \leftarrow$

Done. The question was
“Can we derive a?”

The answer is “Yes, we can”

SLD Derivations

- An **answer** is an answer clause with $m = 0$.

$\text{yes} \leftarrow .$

- A **successful derivation** from KB of query $?q_1 \wedge \dots \wedge q_k$ is a sequence of answer clauses $\gamma_0, \gamma_1, \dots, \gamma_n$ such that
 - γ_0 is the answer clause $\text{yes} \leftarrow q_1 \wedge \dots \wedge q_k$.
 - γ_i is obtained by resolving γ_{i-1} with a clause in KB, and
 - γ_n is an answer $\text{yes} \leftarrow$
- An **unsuccessful derivation** from KB of query $?q_1 \wedge \dots \wedge q_k$
 - We get to something like $\text{yes} \leftarrow b_1 \wedge \dots \wedge b_k$, where there is no clause in KB with any of the b_i as its head

Top-down Proof Procedure for PDCL

To solve the query $? q_1 \wedge \dots \wedge q_k$:

ac:= yes \leftarrow body, where body is $q_1 \wedge \dots \wedge q_k$

repeat

select $q_i \in$ body;

choose clause $C \in$ KB, C is $q_i \leftarrow b_c$;

replace q_i in body by b_c

until ac is an answer (fail if no clause with q_i as head)

select: any choice will work

 (“Don’t care non-determinism”)

choose: truly non-deterministic, must pick the right one

 (“Don’t know non-determinism”)

Example for failing SLD derivation

$a \leftarrow b \wedge c.$ 1 $a \leftarrow e \wedge f.$ $b \leftarrow f \wedge k.$
 $c \leftarrow e.$ $d \leftarrow k$ 3 $e.$
2 $f \leftarrow k.$ $f.$ $j \leftarrow c.$

Query: ?a

$\gamma_0: \text{yes} \leftarrow a$
 $\gamma_1: \text{yes} \leftarrow e \wedge f$
 $\gamma_2: \text{yes} \leftarrow e \wedge k$
 $\gamma_3: \text{yes} \leftarrow k$

“Can we derive a?”
“This time we failed”

There is no rule
with k as its head,
thus ... **fail**

Correspondence between BU and TD proofs

If the following is a top-down (TD) derivation in a given KB, what would be the bottom-up (BU) derivation of the same query?

TD derivation

yes \leftarrow a.

yes \leftarrow b \wedge f.

yes \leftarrow b \wedge g \wedge h.

yes \leftarrow c \wedge d \wedge g \wedge h.

yes \leftarrow d \wedge g \wedge h.

yes \leftarrow g \wedge h.

yes \leftarrow h.

yes \leftarrow .

BU derivation

{

Correspondence between BU and TD proofs

If the following is a top-down (TD) derivation in a given KB, what would be the bottom-up (BU) derivation of the same query?

TD derivation

yes \leftarrow a.

yes \leftarrow b \wedge f.

yes \leftarrow b \wedge g \wedge h.

yes \leftarrow c \wedge d \wedge g \wedge h.

yes \leftarrow d \wedge g \wedge h.

yes \leftarrow g \wedge h.

yes \leftarrow h.

yes \leftarrow .

BU derivation

{}

{h}

{g,h}

{d,g,h}

{c,d,g,h}

{b,c,d,g,h}

{b,c,d,f,g,h}

{a,b,c,d,f,g,h}

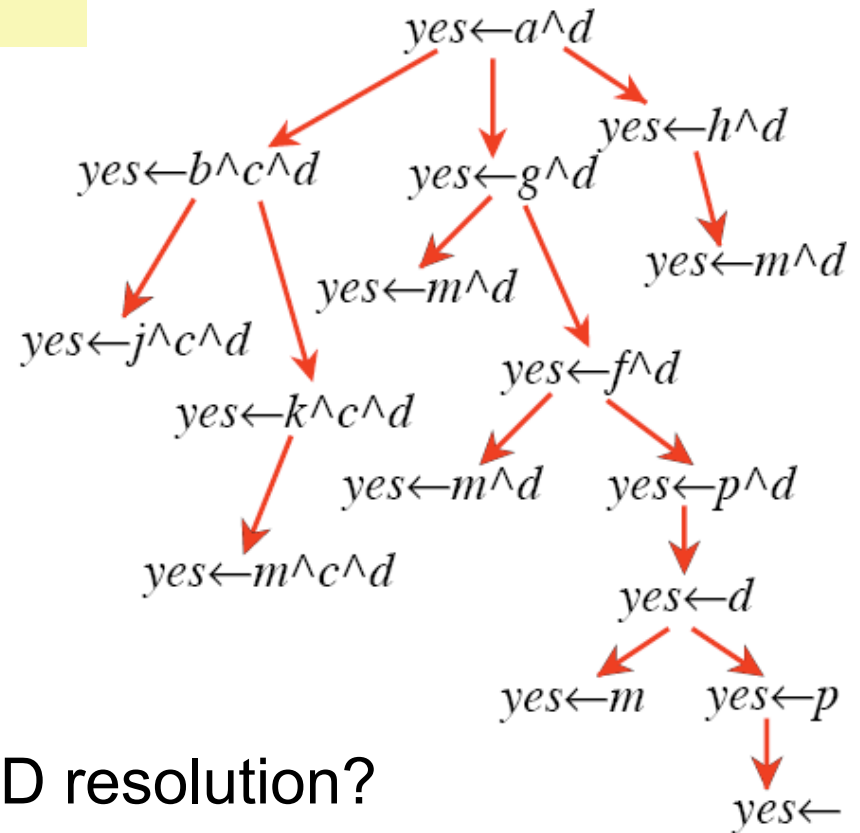
Is the Top-down procedure sound and complete?

- Yes, since there is a 1:1 correspondence between top-down and bottom-up proofs
 - The two methods derive exactly the same atoms (if the SLD resolution picks the successful derivations)

Search Graph for Top-down proofs

Query: $?a \wedge d$.

$a \leftarrow b \wedge c.$	$a \leftarrow g.$
$a \leftarrow h.$	$b \leftarrow j.$
$b \leftarrow k.$	$d \leftarrow m.$
$d \leftarrow p.$	$f \leftarrow m.$
$f \leftarrow p.$	$g \leftarrow m.$
$g \leftarrow f.$	$k \leftarrow m.$
$h \leftarrow m.$	$p.$



What kind of search is SLD resolution?

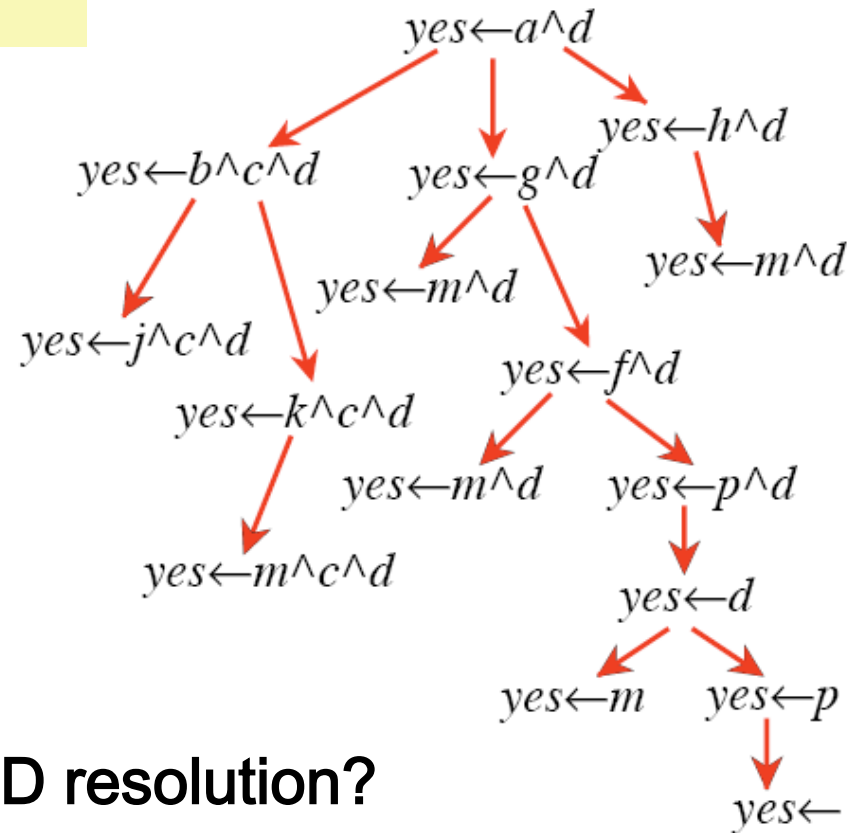
Breadth-first search

Depth-first-search

Search Graph for Top-down proofs

Query: $?a \wedge d$.

$a \leftarrow b \wedge c.$	$a \leftarrow g.$
$a \leftarrow h.$	$b \leftarrow j.$
$b \leftarrow k.$	$d \leftarrow m.$
$d \leftarrow p.$	$f \leftarrow m.$
$f \leftarrow p.$	$g \leftarrow m.$
$g \leftarrow f.$	$k \leftarrow m.$
$h \leftarrow m.$	$p.$



What kind of search is SLD resolution?

It's a depth-first-search. Failing resolutions are paths where the search has to backtrack.

Search Graph for Top-down proofs

Query: $?a \wedge d$.

$a \leftarrow b \wedge c.$ $a \leftarrow g.$

$a \leftarrow h.$ $b \leftarrow j.$

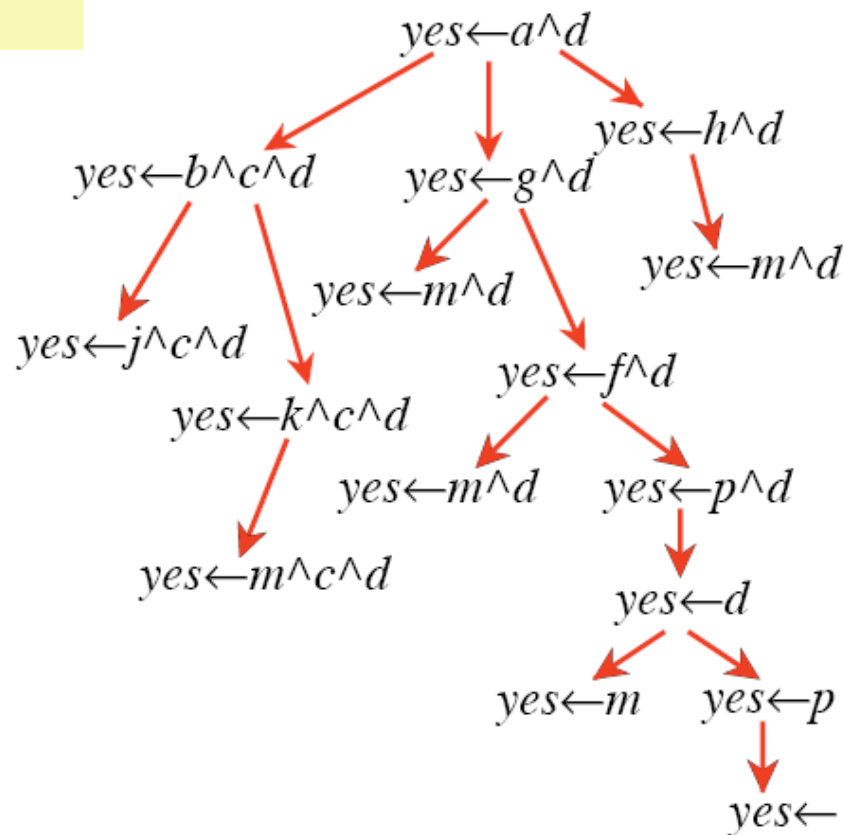
$b \leftarrow k.$ $d \leftarrow m.$

$d \leftarrow p.$ $f \leftarrow m.$

$f \leftarrow p.$ $g \leftarrow m.$

$g \leftarrow f.$ $k \leftarrow m.$

$h \leftarrow m.$ $p.$



Q: Can we use heuristics?

A: Yes! *E.g. number of atoms in the answer clause*

Admissible?

Yes

No

Search Graph for Top-down proofs

Query: $?a \wedge d$.

$a \leftarrow b \wedge c.$ $a \leftarrow g.$

$a \leftarrow h.$ $b \leftarrow j.$

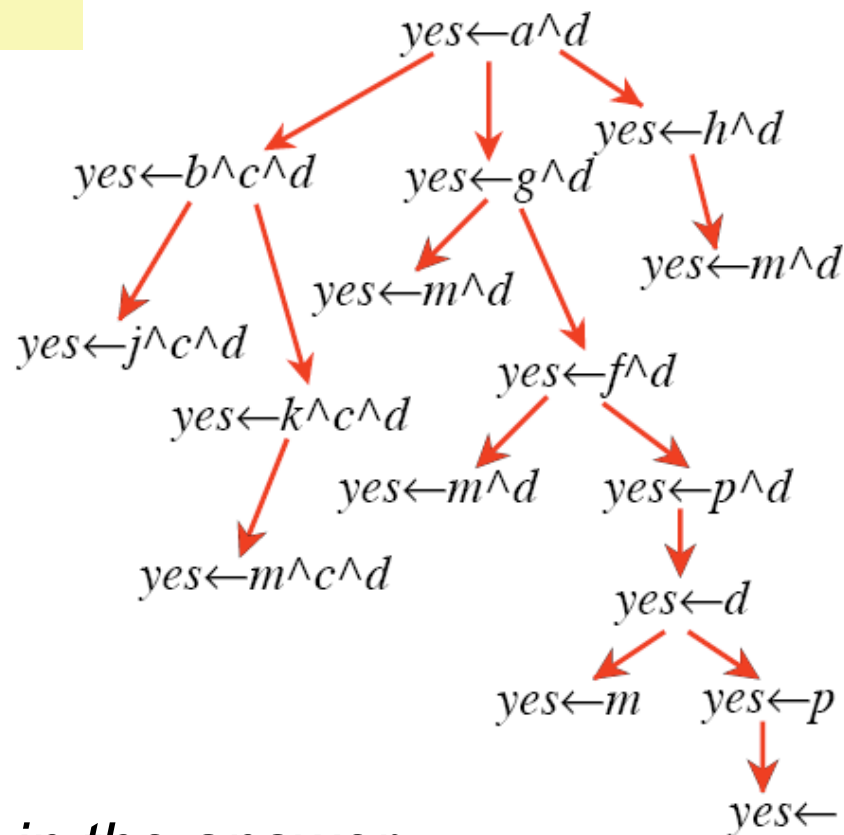
$b \leftarrow k.$ $d \leftarrow m.$

$d \leftarrow p.$ $f \leftarrow m.$

$f \leftarrow p.$ $g \leftarrow m.$

$g \leftarrow f.$ $k \leftarrow m.$

$h \leftarrow m.$ $p.$



Q: Can we use heuristics?

A: Yes! *E.g. number of atoms in the answer clause*

Admissible? *Yes, you need at least these many SLD steps to get an answer*

Inference as Standard Search

- Constraint Satisfaction (Problems):
 - State: assignments of values to a subset of the variables
 - Successor function: assign values to a “free” variable
 - Goal test: set of constraints
 - Solution: possible world that satisfies the constraints
 - Heuristic function: none (all solutions at the same distance from start)
- Planning :
 - State: full assignment of values to features
 - Successor function: states reachable by applying valid actions
 - Goal test: partial assignment of values to features
 - Solution: a sequence of actions
 - Heuristic function: relaxed problem! E.g. “ignore delete lists”
- Inference (Top-down/SLD resolution)
 - State: answer clause of the form $\text{yes} \leftarrow q_1 \wedge \dots \wedge q_k$
 - Successor function: all states resulting from substituting first atom a with $b_1 \wedge \dots \wedge b_m$ if there is a clause $a \leftarrow b_1 \wedge \dots \wedge b_m$
 - Goal test: is the answer clause empty (i.e. $\text{yes} \leftarrow$) ?
 - Solution: the proof, i.e. the sequence of SLD resolutions
 - Heuristic function: number of atoms in the query clause

Lecture Overview

- Recap: Bottom-up proof procedure is sound and complete
- Top-down Proof Procedure

 Datalog

- Logics: Big Picture

Representation and Reasoning in complex domains

- Expressing knowledge with **propositions** can be quite limiting

up_s₂
up_s₃
ok_cb₁
ok_cb₂
live_w₁
connected_w₁_w₂

E.g. **there is no notion** that *w₁* is the same in *live_w₁* and in *connected_w₁_w₂*

- It is often **natural** to consider **individuals** and their **properties**

up(s₂)
up(s₃)
ok(cb₁)
ok(cb₂)
live(w₁)
connected(w₁, w₂)

Now there is a notion that *w₁* is the same in *live(w₁)* and in *connected(w₁, w₂)*

Datalog: What do we gain?

- An extension of propositional definite clause (PDC) logic
 - We now have **variables**
 - We now have **relationships** between variables
 - We can express knowledge that holds for a set of individuals, writing more powerful clauses by introducing variables, such as:

$$\text{live}(W) \leftarrow \text{wire}(W) \wedge \text{connected_to}(W, W_1) \wedge \text{wire}(W_1) \wedge \text{live}(W_1).$$

- We can ask **generic queries**,
 - E.g. “which wires are connected to w_1 ?”

$$? \text{ connected_to}(W, w_1)$$

Datalog: a relational rule language

Datalog expands the syntax of PDCL....

A **variable** is a symbol starting with an upper case letter

Examples: X, Y

A **constant** is a symbol starting with lower-case letter or a sequence of digits.

Examples: alan, w1

A **term** is either a variable or a constant.

Examples: X, Y, alan, w1

A **predicate symbol** is a symbol starting with a lower-case letter.

Examples: live, connected, part-of, in

Datalog Syntax (cont'd)

An **atom** is a symbol of the form p or $p(t_1 \dots t_n)$ where p is a predicate symbol and t_i are terms

Examples: sunny, in(alan,X)

A **definite clause** is either an atom (a fact) or of the form:

$$h \leftarrow b_1 \wedge \dots \wedge b_m$$

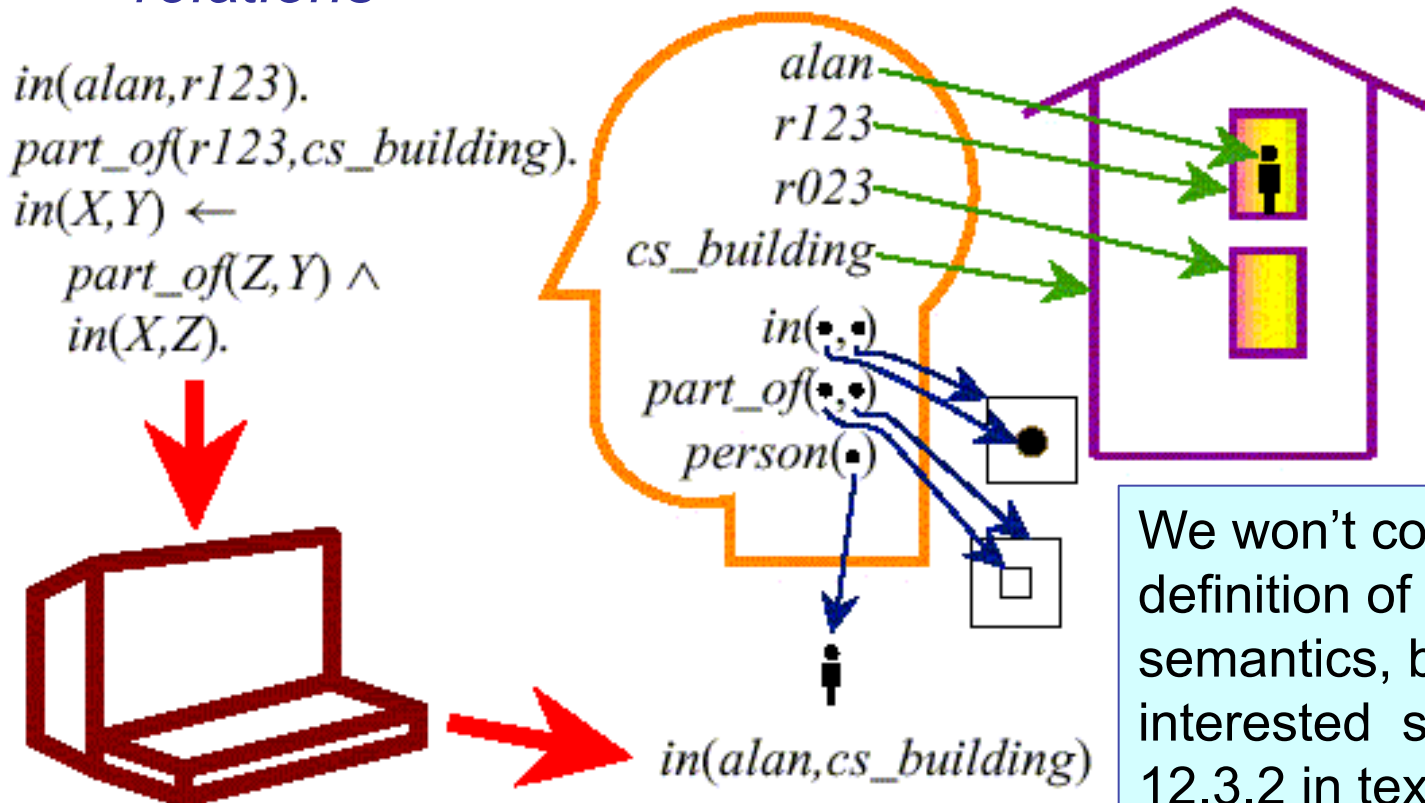
where h and the b_i are atoms (Read this as "` h if b .")

Example: $\text{in}(X,Z) \leftarrow \text{in}(X,Y) \wedge \text{part-of}(Y,Z)$

A **knowledge base** is a set of definite clauses

Datalog Semantics

- Role of semantics is still to connect symbols and sentences in the language with the target domain. Main difference:
 - need to create correspondence both *between terms and individuals*, as well as *between predicate symbols and relations*



We won't cover the formal definition of Datalog semantics, but if you are interested see 12.3.1 and 12.3.2 in textbook

Datalog: Top Down Proof Procedure

```
in(alan, r123).  
part_of(r123,cs_building).  
in(X,Y) ← part_of(Z,Y) & in(X,Z).
```

- Extension of Top-Down procedure for PDCL.
How do we deal with variables?
 - Idea:
 - Find clauses with heads that match the query
 - Substitute variable in the clause with the matching constant
 - Example:

Query: $yes \leftarrow in(alan, cs_building).$



$in(X,Y) \leftarrow part_of(Z,Y) \& in(X,Z).$
with $Y = cs_building$

$yes \leftarrow part_of(Z,cs_building), in(alan, Z).$

- We will not cover the formal details of this process, called *unification*.
See P&M Section 12.4.2, p. 511 for the details.

Example proof of a Datalog query

in(alan, r123).
part_of(r123,cs_building).
in(X,Y) ← part_of(Z,Y) & in(X,Z).

Query: yes ← in(alan, cs_building).

Using clause: in(X,Y) ←
part_of(Z,Y) & in(X,Z),
with Y = cs_building

yes ← part_of(Z,cs_building), in(alan, Z).

Using clause:
part_of(r123,cs_building)
with Z = r123

yes ← in(alan, r123).

Using clause:
in(alan, r123).

Using clause: in(X,Y) ←
part_of(Z,Y) & in(X,Z).
With Z = alan

yes ←.

yes ← part_of(Z, r123), in(alan, Z).

No clause with
matching head:
part_of(Z,r123).

fail

Tracing Datalog proofs in AIspace

- You can trace the example from the last slide in the AIspace Deduction Applet at <http://aispace.org/deduction/> using file <http://cs.ubc.ca/~mack/CS322/in-part-of.pl>



- Question 4 of assignment 3 asks you to use this applet

Datalog: queries with variables

```
in(alan, r123).  
part_of(r123,cs_building).  
in(X,Y) ← part_of(Z,Y) & in(X,Z).
```

Query: in(alan, X1).
yes(X1) ← in(alan, X1).

What would the answer(s) be?

Datalog: queries with variables

```
in(alan, r123).  
part_of(r123,cs_building).  
in(X,Y) ← part_of(Z,Y) & in(X,Z).
```

Query: in(alan, X1).
yes(X1) ← in(alan, X1).

What would the answer(s) be?

```
yes(r123).  
yes(cs_building).
```

You can trace the SLD derivation for this query in the AIspace Deduction Applet, using file <http://cs.ubc.ca/~mack/CS322/in-part-of.pl>



One important Datalog detail

- In its SLD resolution proof, Datalog always chooses the first clause with a matching head it finds in KB
- What does that mean for recursive function definitions?
 - The clause(s) defining your base case(s) have to appear first in KB
 - Otherwise, you can get infinite recursions
 - This is similar to recursion in imperative and functional programming languages
- Datalog is a subset of Prolog (**P**rogramming in **L**ogic)

Learning Goals For Logic

- PDCL syntax & semantics
 - Verify whether a logical statement belongs to the language of propositional definite clauses
 - Verify whether an **interpretation** is a **model** of a PDCL KB.
 - Verify when a conjunction of atoms is a **logical consequence** of a KB
- Bottom-up proof procedure
 - Define/read/write/trace/debug the Bottom Up (**BU**) proof procedure
 - Prove that the BU proof procedure is **sound and complete**
- Top-down proof procedure
 - Define/read/write/trace/debug the Top-down (**SLD**) proof procedure (as a search problem)
- Datalog
 - Represent simple domains in Datalog
 - Apply the **Top-down** proof procedure in Datalog

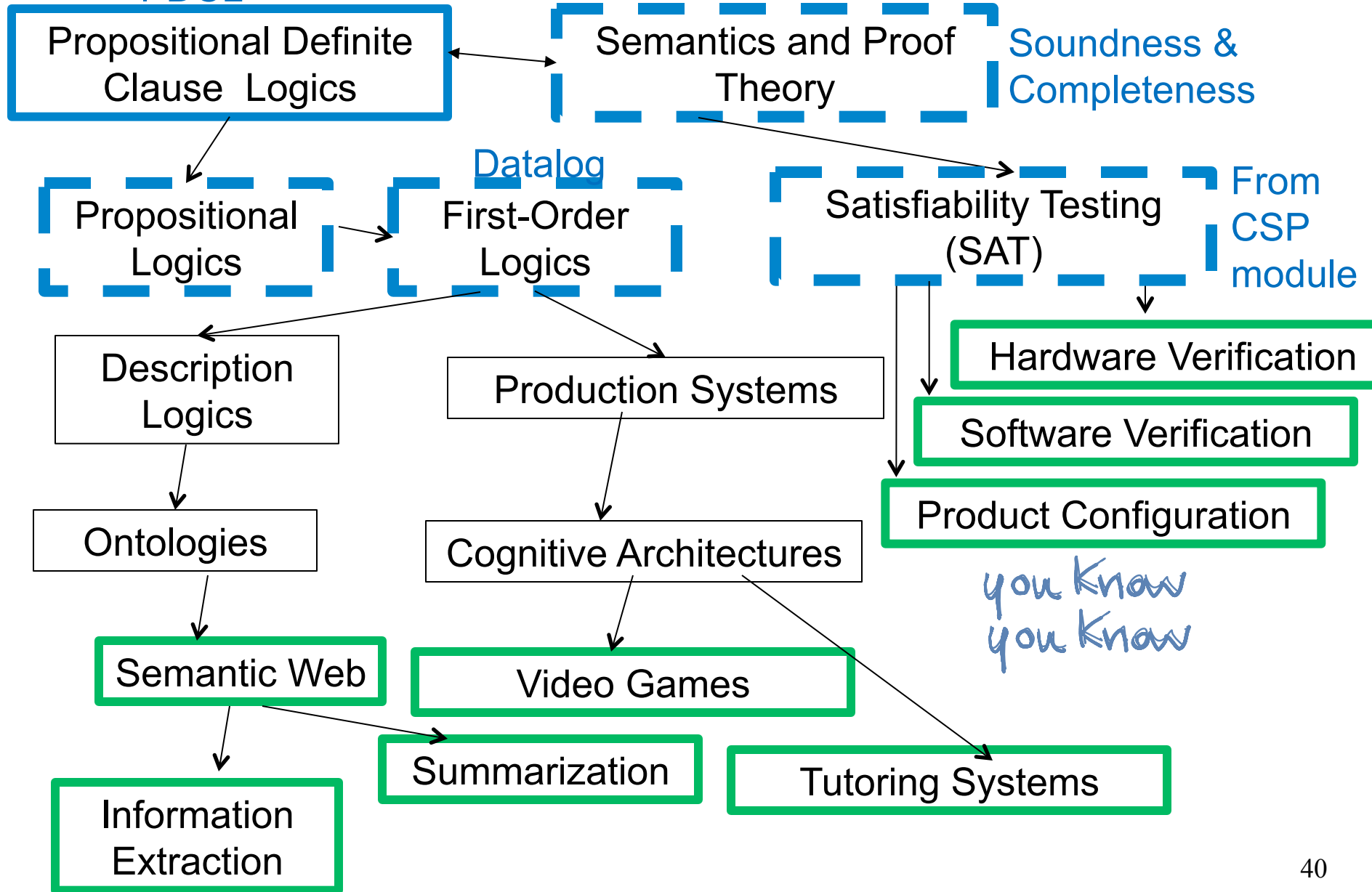
Lecture Overview

- Recap: Bottom-up proof procedure is sound and complete
- Top-down Proof Procedure
- Datalog

 Logics: Big Picture

Logics: Big picture

PDCL



Logics: Big picture

- We only covered rather simple logics
 - There are much more powerful representation and reasoning systems based on logics e.g. full first order logic (with negation, disjunction and function symbols), second-order logics, non-monotonic logics, modal logics, ...
- There are many important applications of logic
 - For example, software agents roaming the web on our behalf
 - Based on a more structured representation: the semantic web
 - This is just one example for how logics are used

Example problem: automated travel agent

- Examples for typical queries
 - How much is a typical flight to Mexico for a given date?
 - What's the cheapest vacation package to some place in the Caribbean in a given week?
 - Plus, the hotel should have a white sandy beach and scuba diving
- If webpages are based on basic HTML
 - Humans need to scout for the information and integrate it
 - Computers are not reliable enough (yet?)
 - Natural language processing can be powerful (see Watson and Siri!)
 - But some information may be in pictures (beach), or implicit in the text, so simple approaches like Watson and Siri still don't get all the info.

More structured representation: the Semantic Web

- Beyond HTML pages only made for humans
- Languages and formalisms **based on logics** that allow websites to include information in a more structured format
 - Goal: software agents that can roam the web and carry out sophisticated tasks on our behalf.
 - This is different than searching content for keywords and popularity!
- For further material, see P&M text, Chapter 13 and the Introduction to the Semantic Web tutorial given at **2011 Semantic Technology Conference**
<http://www.w3.org/People/Ivan/CorePresentations/SWTutorial/>
(This is the best technical intro; Herman keeps it up to date.)

Examples of ontologies for the Semantic Web

- “Ontology”: logic-based representation of the world
- eClassOwl: eBusiness ontology
 - for products and services
 - 75,000 classes (types of individuals) and 5,500 properties
- National Cancer Institute’s ontology: 58,000 classes
- Open Biomedical Ontologies Foundry: several ontologies
 - including the Gene Ontology to describe
 - gene and gene product attributes in any organism or protein sequence
 - annotation terminology and data
- OpenCyc project: a 150,000-concept ontology including
 - Top-level ontology
 - describes general concepts such as numbers, time, space, etc
 - Hierarchical composition: superclasses and subclasses
 - Many specific concepts such as “OLED display”, “iPhone”

Course Overview

Course Module

Environment

Deterministic

Stochastic

Representation

Reasoning
Technique

Problem Type

Constraint
Satisfaction

Logic

Planning

	<p>Arc Consistency</p> <p><i>Variables + Search</i></p> <p><i>Constraints</i></p>	
Static	<p><i>Logics</i></p> <p>Search</p>	<p><i>Bayesian Networks</i></p> <p>Variable Elimination</p>
Sequential	<p><i>STRIPS</i></p> <p>Search</p> <p>As CSP (using arc consistency)</p>	<p><i>Decision Networks</i></p> <p>Variable Elimination</p> <p><i>Markov Processes</i></p> <p>Value Iteration</p>

Uncertainty

Decision
Theory

This concludes
the logic module