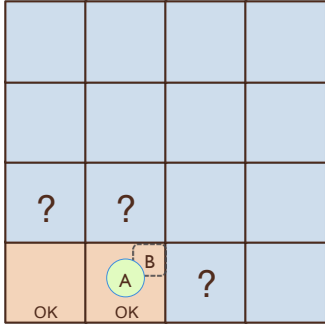




Lecture 09:  
Propositional Logic, II

Artificial Intelligence (CS 131)

1

### Logic in the Wumpus World



 : Agent  
 : Breeze

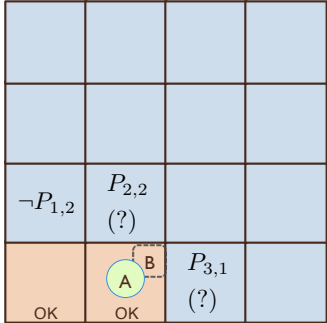
We can use PL to express the situation of the agent:



- Let  $P_{i,j}$  be true if there is a pit in location  $[i, j]$
- Let  $B_{i,j}$  be true if there is a breeze in location  $[i, j]$
- The relevant KB is:
 
$$\{\neg P_{1,1}, \neg P_{2,1}, \neg B_{1,1}, B_{2,1}\}$$
- PL can also express Wumpus World rules like "Pits cause breezes in adjacent squares":
 
$$B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$$

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

2

### Inference in the Wumpus World



 : Agent  
 : Breeze

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

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

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

Based upon the knowledge base of basic propositional facts, and logical rules about how the world functions, we want to infer *new knowledge* about that world:

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

3

### Inference Using Truth-Tables

| Proposition Symbols |           |           |           |           |           |           | Sentences in the Knowledge Base |                |           |  |   |
|---------------------|-----------|-----------|-----------|-----------|-----------|-----------|---------------------------------|----------------|-----------|--|---|
| $B_{1,1}$           | $B_{2,1}$ | $P_{1,1}$ | $P_{1,2}$ | $P_{2,1}$ | $P_{2,2}$ | $P_{3,1}$ | $\neg P_{1,1}$                  | $\neg B_{1,1}$ | $B_{2,1}$ | $B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})$ | $B_{2,1} \Leftrightarrow (P_{1,1} \vee (P_{2,2} \vee P_{3,1}))$ |
| T                   | T         | T         | T         | T         | T         | T         | F                               | F              | T         | T  | T   |
| T                   | T         | T         | T         | T         | T         | F         | F                               | F              | T         | T  | T   |
| T                   | T         | T         | T         | T         | F         | T         | F                               | F              | T         | T  | T   |
| T                   | T         | T         | T         | T         | F         | F         | F                               | F              | T         | T  | T   |
| :                   | :         | :         | :         | :         | :         | :         | :                               | :              | :         | :  | :   |
| F                   | T         | F         | F         | F         | T         | T         | T                               | T              | T         | T  | T   |
| F                   | T         | F         | F         | F         | T         | F         | T                               | T              | T         | T  | T   |
| F                   | T         | F         | F         | F         | F         | T         | T                               | T              | T         | T  | T   |
| :                   | :         | :         | :         | :         | :         | :         | :                               | :              | :         | :  | :   |
| F                   | F         | F         | F         | F         | F         | T         | T                               | T              | F         | T  | F   |
| F                   | F         | F         | F         | F         | F         | F         | T                               | T              | F         | T  | T   |

We can use the full set of possible truth-values to check whether a sentence  $\alpha$  is entailed by our knowledge base

4

## Inference Using Truth-Tables

| Proposition Symbols   | Sentences in the Knowledge Base  |
|---|--|
| $B_{1,1} \ B_{2,1} \ P_{1,1} \ P_{1,2} \ P_{2,1} \ P_{2,2} \ P_{3,1}$ | $\neg P_{1,1} \ \neg B_{1,1} \ B_{2,1} \ B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1}) \ B_{2,1} \Leftrightarrow (P_{1,1} \vee (P_{2,2} \vee P_{3,1}))$ |
| T T T T T T T   | F F T T T  |
| T T T T T T F   | F F T T T  |
| T T T T T F T   | F F T T T  |
| T T T T T F F   | F F T T T  |
| ⋮   | ⋮  |
| F T F F F T T   | T T T T T  |
| F T F F F T F   | T T T T T  |
| F T F F F F T   | T T T T T  |
| ⋮   | ⋮  |
| F F F F F F T   | T T F T F  |
| F F F F F F F   | T T F T T  |

- Given the full set of truth values, we consider models/assignments for which *everything we know already* is true

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5

5

## Inference Using Truth-Tables

| Proposition Symbols   | Sentences in the Knowledge Base  |
|---|--|
| $B_{1,1} \ B_{2,1} \ P_{1,1} \ P_{1,2} \ P_{2,1} \ P_{2,2} \ P_{3,1}$ | $\neg P_{1,1} \ \neg B_{1,1} \ B_{2,1} \ B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1}) \ B_{2,1} \Leftrightarrow (P_{1,1} \vee (P_{2,2} \vee P_{3,1}))$ |
| T T T T T T T   | F F T T T  |
| T T T T T T F   | F F T T T  |
| T T T T T F T   | F F T T T  |
| T T T T T F F   | F F T T T  |
| ⋮   | ⋮  |
| F T F F F T T   | T T T T T  |
| F T F F F T F   | T T T T T  |
| F T F F F F T   | T T T T T  |
| ⋮   | ⋮  |
| F F F F F F T   | T T F T F  |
| F F F F F F F   | T T F T T  |

- Now, we can see that there *cannot* be a Pit at location (1,2), since that sentence is false in every model consistent with what we already know

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6

6

## Inference Using Truth-Tables

| Proposition Symbols   | Sentences in the Knowledge Base  |
|---|--|
| $B_{1,1} \ B_{2,1} \ P_{1,1} \ P_{1,2} \ P_{2,1} \ P_{2,2} \ P_{3,1}$ | $\neg P_{1,1} \ \neg B_{1,1} \ B_{2,1} \ B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1}) \ B_{2,1} \Leftrightarrow (P_{1,1} \vee (P_{2,2} \vee P_{3,1}))$ |
| T T T T T T T   | F F T T T  |
| T T T T T T F   | F F T T T  |
| T T T T T F T   | F F T T T  |
| T T T T T F F   | F F T T T  |
| ⋮   | ⋮  |
| F T F F F T T   | T T T T T  |
| F T F F F T F   | T T T T T  |
| F T F F F F T   | T T T T T  |
| ⋮   | ⋮  |
| F F F F F F T   | T T F T F  |
| F F F F F F F   | T T F T T  |

- Similarly, we can see that there is a Pit *either* at (2,2) *or* (3,1), but we don't know which, since either one could be false (although *not both*)

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7

7

## Inference Techniques

- Inference is the procedure of **deriving** (i.e., generating) new sentences based on our existing knowledge base

$KB \vdash_P \alpha \equiv \alpha$  can be derived from  $KB$  using procedure  $P$

- Procedure  $P$  is **sound** if it only ever leads to sentences that are entailed by the knowledge base:

$$KB \vdash_P \alpha \Rightarrow KB \models \alpha$$

- Procedure  $P$  is **complete** if it can generate *all* sentences that are entailed by the knowledge base:

$$KB \models \alpha \Rightarrow KB \vdash_P \alpha$$

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8

8

## Effective Inference Procedures

- Enumerating all truth tables to derive new sentences can be automated (algorithm: Figure 7.10, page 221 of R&N text)
- This procedure is sound and complete; however, it is ineffective due to complexity:
  - For  $n$  propositional symbols, the full truth-table has  $2^n$  rows
  - If it took only a nanosecond to check each row, then checking:

|             |   |   |
|-------------|---|---|
| 10 symbols  | = | 0.000001s                                     |
| 20 symbols  | = | 0.0015s                                       |
| 30 symbols  | = | 1.1s  |
| 40 symbols  | = | 1100s = 18.3 minutes                          |
| 50 symbols  | = | 1,125,900s = 13 days                          |
| 60 symbols  | = | $12 \times 10^9$ s = 36.6 years               |
| ⋮           |   |   |
| 100 symbols | = | $1.27 \times 10^{21}$ s = 40.2 trillion years |

- For Breeze and Pit only, (4 × 4) WumpusWorld has 32 symbols
- For Breeze, Stench, Glitter, Pit, Gold, Wumpus, we already get 96!

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9

9

## A Special Case: Definite/Horn Form

- In some cases, we can represent KB in a special *restricted* PL syntax
- A **definite clause**  $\alpha$  is either (1) a **literal**: either  $P_i$  or  $\neg P_i$ , for some propositional symbol  $P_i$ , or (2) a sentence of the special form:



- Definite clauses are equivalent to **disjunctions** with *exactly one* positive symbol
 
$$P_1 \wedge P_2 \wedge \dots \wedge P_j \Rightarrow P_k \equiv \neg(P_1 \wedge P_2 \wedge \dots \wedge P_j) \vee P_k$$

$$\equiv \neg P_1 \vee \neg P_2 \vee \dots \vee \neg P_j \vee P_k$$
- A slightly more relaxed definition: **Horn clause**, which has *at most one* positive symbol; both of the following are Horn clauses, but only the *first* is definite

$$\neg P_1 \vee \neg P_2 \vee \neg P_3 \vee P_4$$

$$\neg P_1 \vee \neg P_2 \vee \neg P_3 \vee \neg P_4$$

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10

10

## A Special Case: Definite/Horn Form

- If our knowledge base is made up only of definite clauses, then we can repeatedly apply the rule of **Modus Ponens (MP)**:

$$\frac{\alpha_1 \quad \alpha_2 \quad \dots \quad \alpha_n \quad (\alpha_1 \wedge \alpha_2 \wedge \dots \wedge \alpha_n) \Rightarrow \beta}{\beta}$$

- This can be written in disjunctive form as well:

$$\frac{\alpha_1 \quad \alpha_2 \quad \dots \quad \alpha_n \quad \neg \alpha_1 \vee \neg \alpha_2 \vee \dots \vee \neg \alpha_n \vee \beta}{\beta}$$

- This rule gives us a **sound and complete** inference procedure that can be employed in simple **linear-time** algorithms

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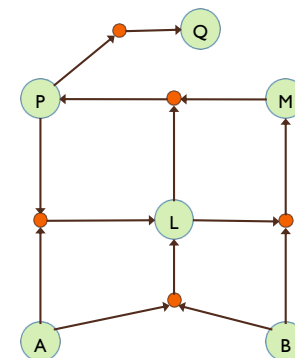
11

11

## Forward Chaining with Definite KBs

- When we have a set of definite clauses, we can represent the structure of our knowledge base graphically

| KB           |                 |
|--------------|-----------------|
| $P$          | $\Rightarrow Q$ |
| $L \wedge M$ | $\Rightarrow P$ |
| $B \wedge L$ | $\Rightarrow M$ |
| $A \wedge P$ | $\Rightarrow L$ |
| $A \wedge B$ | $\Rightarrow L$ |
| $A$          |                 |
| $B$          |                 |



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12

12

## Forward Chaining with Definite KBs

Image source: Russell & Norvig (2021)

```

function PL-FC-ENTAILS?(KB, q) returns true or false
  inputs: KB, the knowledge base, a set of propositional definite clauses
         q, the query, a proposition symbol
  count ← a table, where count[c] is initially the number of symbols in clause c's premise
  inferred ← a table, where inferred[s] is initially false for all symbols
  queue ← a queue of symbols, initially symbols known to be true in KB

  while queue is not empty do
    p ← POP(queue)
    if p = q then return true
    if inferred[p] = false then
      inferred[p] ← true
      for each clause c in KB where p is in c.PREMISE do
        decrement count[c]
        if count[c] = 0 then add c.CONCLUSION to queue
  return false
  
```

For definite clauses that are equivalent to conditionals, count how many propositions make up its tail

- For example, the following clause has  $count = 3$ :  
 $\neg P_1 \vee \neg P_2 \vee \neg P_3 \vee P_4 \equiv (P_1 \wedge P_2 \wedge P_3) \Rightarrow P_4$

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13

## Forward Chaining with Definite KBs

```

function PL-FC-ENTAILS?(KB, q) returns true or false
  inputs: KB, the knowledge base, a set of propositional definite clauses
         q, the query, a proposition symbol
  count ← a table, where count[c] is initially the number of symbols in clause c's premise
  inferred ← a table, where inferred[s] is initially false for all symbols
  queue ← a queue of symbols, initially symbols known to be true in KB

  while queue is not empty do
    p ← POP(queue)
    if p = q then return true
    if inferred[p] = false then
      inferred[p] ← true
      for each clause c in KB where p is in c.PREMISE do
        decrement count[c]
        if count[c] = 0 then add c.CONCLUSION to queue
  return false
  
```

Premise elimination: if we see part of the tail, decrement count

Modus Ponens: if we have seen all parts of tail, infer head

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14

## Backwards Chaining

- A similar algorithm works *backward* from the thing we want to prove,  $q$ , checking if it is already in KB
  - If not, then we check for a Horn clause that has  $q$  as its **head**
  - If so, we repeat the process for all premises in the **tail**
  - Terminate (and fail) if any premise can't be proved, else succeed
- Both FC and BC are linear-time algorithms:
  - By keeping track of whether we have used a premise before (either to "fire" a clause in FC or as an object of proof in BC), we visit each propositional symbol at most once
  - For  $n$  sentences in our KB, the algorithms run in  $O(n)$
  - BC can often do even better, avoiding unnecessary work

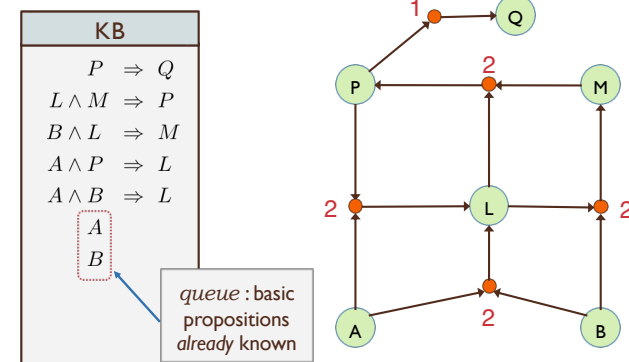
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15

## Forward Chaining with Definite KBs, 01

- Initial  $count$  set to # premises in tail of conditionals



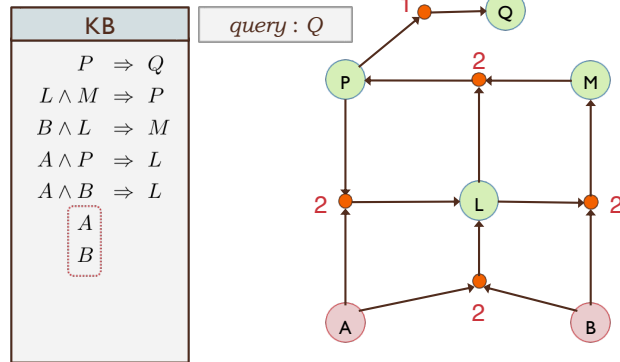
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16

## Forward Chaining with Definite KBs, 02

- Suppose we are trying to infer  $Q$ : we begin to “fire” clauses using Modus Ponens rules, decrementing *count*



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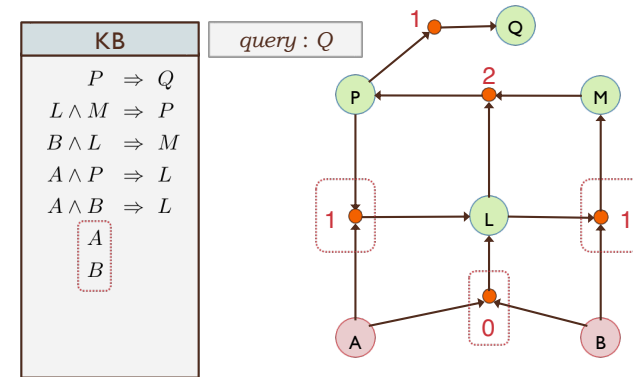
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17

17

## Forward Chaining with Definite KBs, 03

- Known facts in the queue lead to decremented *count*



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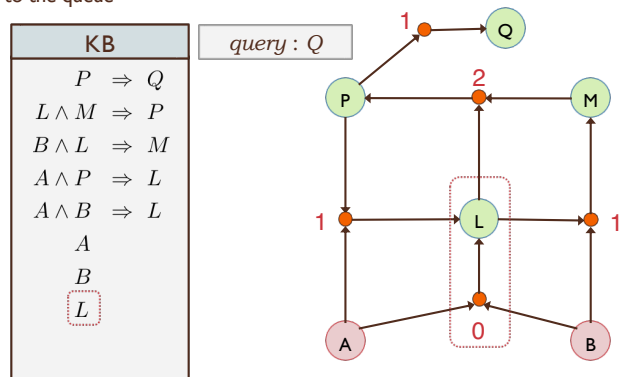
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18

18

## Forward Chaining with Definite KBs, 04

- Whenever *count* reaches 0, we add the associated *head* of the conditional to the queue



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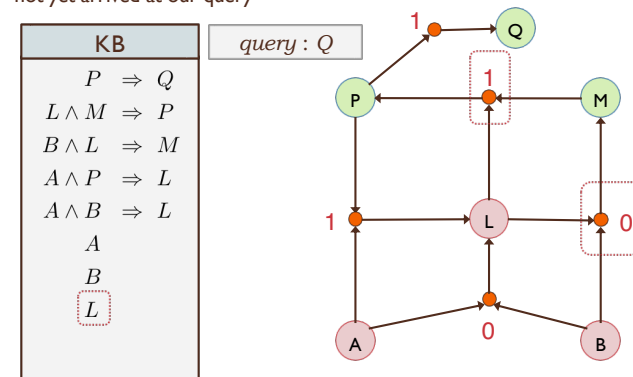
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19

19

## Forward Chaining with Definite KBs, 05

- We repeat the process so long as something is in the queue, and we have not yet arrived at our query



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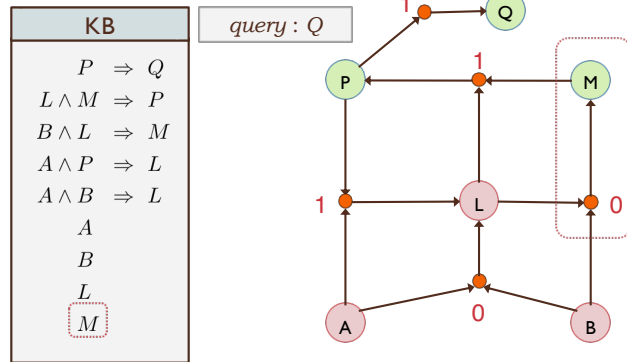
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20

20

## Forward Chaining with Definite KBs, 06

- ▶ We repeat the process so long as something is in the queue, and we have not yet arrived at our query



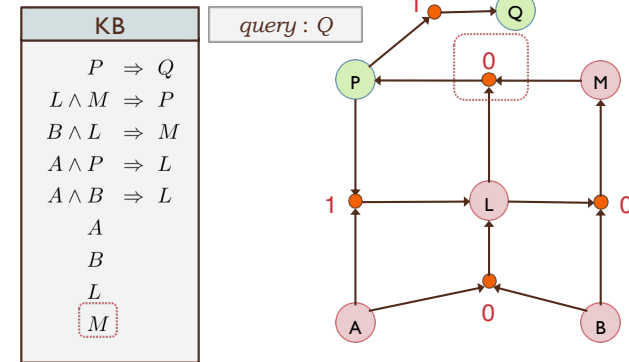
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21

## Forward Chaining with Definite KBs, 07

- ▶ We repeat the process so long as something is in the queue, and we have not yet arrived at our query



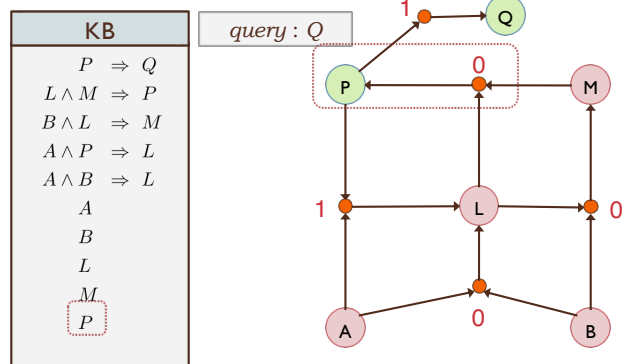
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22

## Forward Chaining with Definite KBs, 08

- ▶ We repeat the process so long as something is in the queue, and we have not yet arrived at our query



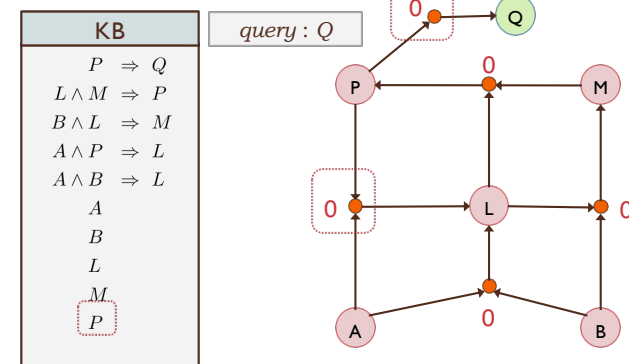
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23

## Forward Chaining with Definite KBs, 09

- ▶ We repeat the process so long as something is in the queue, and we have not yet arrived at our query



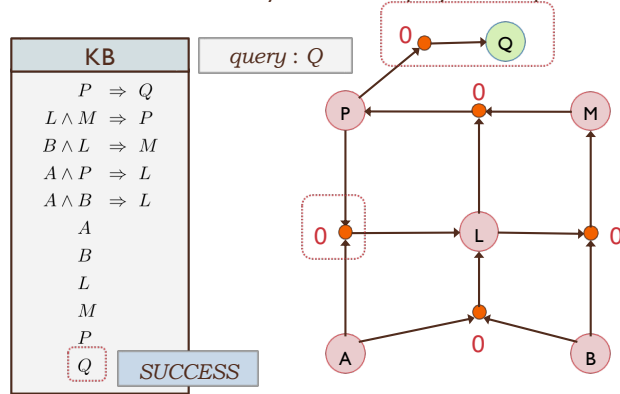
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24

## Forward Chaining with Definite KBs, 10

- ▶ We conclude when either every count is 0, or query is in the queue



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25

25

## A General Algorithm: Resolution

- ▶ While definite/Horn clauses cannot express everything in PL, we can put any sentence into **Conjunctive Normal Form** (CNF): a conjunction where each **clause** is a disjunction of positive/negative literals, e.g.:

$$(\neg P_1 \vee P_2) \wedge (P_3 \vee P_4 \vee \neg P_5) \wedge P_6$$

- ▶ For disjunctive clauses, **resolution** is a sound & complete inference rule:

$$\frac{l_1 \vee \dots \vee l_k \quad m_1 \vee \dots \vee m_n}{l_1 \vee \dots \vee l_{i-1} \vee l_{i+1} \vee \dots \vee l_k \vee m_1 \vee \dots \vee m_{j-1} \vee m_{j+1} \vee \dots \vee m_n}$$

where  $l_i$  and  $m_j$  are **complementary**—i.e., one is the negation of the other:

$$\frac{P_{1,3} \vee P_{2,2} \quad \neg P_{2,2}}{P_{1,3}} \quad \frac{\neg P_{2,4} \vee P_{1,1} \quad P_{3,3} \vee P_{2,4}}{P_{1,1} \vee P_{3,3}}$$

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26

26

## Converting to CNF

- ▶ Every sentence of PL can be written in CNF by a simple set of conversions. For example:

$$S_{1,1} \Leftrightarrow (\neg P_{1,2} \vee \neg P_{2,1})$$

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

$$(S_{1,1} \Rightarrow (\neg P_{1,2} \vee \neg P_{2,1})) \wedge ((\neg P_{1,2} \vee \neg P_{2,1}) \Rightarrow S_{1,1})$$
2. Eliminate  $\Rightarrow$ , replacing  $\alpha \Rightarrow \beta$  with  $\neg \alpha \vee \beta$ :  

$$(\neg S_{1,1} \vee \neg P_{1,2} \vee \neg P_{2,1}) \wedge (\neg(\neg P_{1,2} \vee \neg P_{2,1}) \vee S_{1,1})$$
3. Convert  $\neg(\alpha \vee \beta)$  to  $(\neg \alpha \wedge \neg \beta)$ , by de Morgan's rules:  

$$(\neg S_{1,1} \vee \neg P_{1,2} \vee \neg P_{2,1}) \wedge ((\neg \neg P_{1,2} \wedge \neg \neg P_{2,1}) \vee S_{1,1})$$
4. Eliminate any double negations:  

$$(\neg S_{1,1} \vee \neg P_{1,2} \vee \neg P_{2,1}) \wedge ((P_{1,2} \wedge P_{2,1}) \vee S_{1,1})$$
5. Apply distributive laws ( $\vee$  over  $\wedge$ ):  

$$(\neg S_{1,1} \vee \neg P_{1,2} \vee \neg P_{2,1}) \wedge (P_{1,2} \vee S_{1,1}) \wedge (P_{2,1} \vee S_{1,1})$$

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27

27

## The Resolution Algorithm

Image source: [Russell & Norvig \(2021\)](#)

- ▶ We can use repeated applications of resolution to show  $(KB \wedge \neg \alpha)$  is **unsatisfiable**, where  $\alpha$  is some query we are interested in (NB: **negated**)
  - ▶ If achieve **empty set** of clauses, this shows that the KB and the query  $\neg \alpha$  cannot both be true at the same time; by definition, this means that KB **entails**  $\alpha$
  - ▶ A form of **proof by contradiction**

**function** PL-RESOLUTION( $KB, \alpha$ ) **returns** true or false  
**inputs:**  $KB$ , the knowledge base, a sentence in propositional logic  
 $\alpha$ , the query, a sentence in propositional logic

$clauses \leftarrow$  the set of clauses in the CNF representation of  $KB \wedge \neg \alpha$   
 $new \leftarrow \{\}$   
**while** true **do**  
   **for each** pair of clauses  $C_i, C_j$  **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  
    $clauses \leftarrow clauses \cup new$

The basic resolution rule, applied as many times as appropriate

No entailment if we ever run out of new things to resolve

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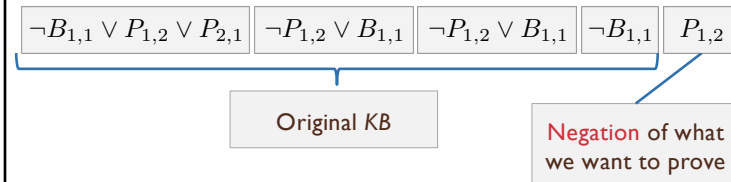
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28

28

## Using Resolution

Suppose we want to prove:  $\neg P_{1,2}$  using  $KB: (B_{1,1} \Leftrightarrow (P_{1,2} \vee P_{2,1})) \wedge \neg B_{1,1}$   
 Convert  $KB$  to CNF:  $(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}) \wedge (\neg P_{1,2} \vee B_{1,1}) \wedge (\neg P_{1,2} \vee B_{1,1}) \wedge \neg B_{1,1}$   
 Equivalent to clauses:  $\{(\neg B_{1,1} \vee P_{1,2} \vee P_{2,1}), (\neg P_{1,2} \vee B_{1,1}), (\neg P_{1,2} \vee B_{1,1}), \neg B_{1,1}\}$   
 Add in the negation of what we want to prove:  $\neg \neg P_{1,2} \equiv P_{1,2}$



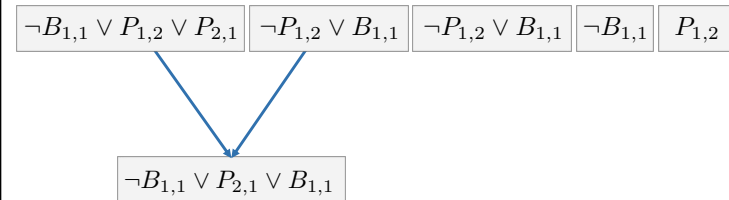
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29

## Using Resolution

- Resolution rule will not *always* be *productive*: it can sometimes generate consequences that are not really helpful to our proof:



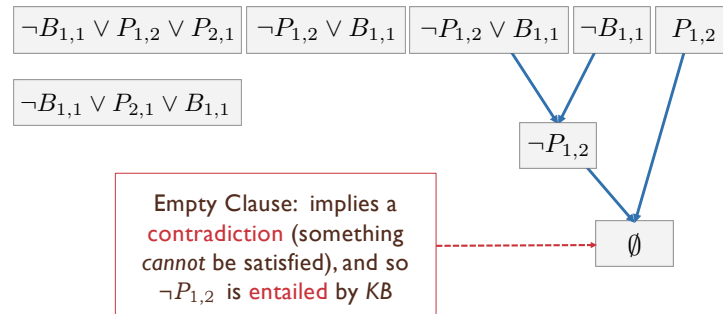
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30

## Using Resolution

- However, it will also eventually generate *everything* that is entailed by  $KB$  (a **complete** procedure):



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Artificial Intelligence (CS 452) 31

31

## Resolution as an Algorithm

- Resolution is a **sound** and **complete** inference method
- However, while *often* effective, we must live with the fact that in the worst case it can run in **exponential time**
  - Given  $n$  propositional symbols, we may need  $2^n$  applications of basic resolution rule to reach a stopping condition
- This seems to be a necessary fact for any general-purpose PL inference procedure: it is well-known that the PL entailment/provability question is **NP-complete**

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Artificial Intelligence (CS 452) 32

32