# FORWARD CHAINING

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# Inference Chaining

- Generalized Modus Ponens gives us a natural, intuitive, and reasonably powerful tool to use for inference.
- There are two types of inference that differ simply in direction.
- Forward Chaining starts with new premises and tries to generate all new conclusions.
- Backward Chaining begins from a desired conclusion and attempts to realize the necessary implications and premises required to arrive at it.

# Inference Procedures

### Forward Chaining

- Data-driven
- Triggered by adding a new fact
- Premises  $\Rightarrow$  Consequent(s)
- Renaming
- Composition:

Subst(Compose( $\theta_1$ ,  $\theta_2$ ), p) = Subst( $\theta_2$ , Subst( $\theta_1$ , p))

### **Backward Chaining**

- Goal-driven
- Triggered by a query, i.e. Ask
- · Premises ← Consequent
- · Build up the unifier as it goes

#### Which method to use?

#### FORWARD CHAINING

**BACKWARD CHAINING** 

If the 'average' rule has more conditions than conclusions, that is the typical hypothesis or goal (the conclusions) can lead to many more questions (the conditions)

The average rule has more conclusions than conditions such that each fact may fan out into a large number of new facts or actions

# Forward Chaining Algorithm

```
function FOL-FC-Ask(KB, \alpha) returns a substitution or false
repeat until new is empty
      new \leftarrow \{\}
      for each sentence r in KB do
            (p_1 \land \ldots \land p_n \Rightarrow q) \leftarrow \text{STANDARDIZE-APART}(r)
            for each \theta such that (p_1 \land \ldots \land p_n)\theta = (p'_1 \land \ldots \land p'_n)\theta
                             for some p'_1, \ldots, p'_n in KB
                 q' \leftarrow \text{SUBST}(\theta, q)
                if q' is not a renaming of a sentence already in KB or new then do
                       add q' to new
                       \phi \leftarrow \text{UNIFY}(q', \alpha)
                       if \phi is not fail then return \phi
      add new to KB
return false
```

... it is a crime for an American to sell weapons to hostile nations:

 $American(x) \land Weapon(y) \land Sells(x, y, z) \land Hostile(z) \Rightarrow Criminal(x)$ Nono ... has some missiles, i.e.,  $\exists x \ Owns(Nono, x) \land Missile(x)$ :

 $Owns(Nono, M_1)$  and  $Missile(M_1)$ 

... all of its missiles were sold to it by Colonel West

 $\forall x \; Missile(x) \land Owns(Nono, x) \Rightarrow Sells(West, x, Nono)$ 

Missiles are weapons:

 $Missile(x) \Rightarrow Weapon(x)$ 

An enemy of America counts as "hostile":

 $Enemy(x, America) \Rightarrow Hostile(x)$ 

West, who is American ...

American(West)

The country Nono, an enemy of America ...

Enemy(Nono, America)

## Properties of forward chaining

Sound and complete for first-order definite clauses (proof similar to propositional proof)

Datalog = first-order definite clauses + no functions (e.g., crime KB) FC terminates for Datalog in poly iterations: at most  $p \cdot n^k$  literals

May not terminate in general if  $\alpha$  is not entailed

This is unavoidable: entailment with definite clauses is semidecidable

### Efficiency of forward chaining

Simple observation: no need to match a rule on iteration k if a premise wasn't added on iteration k-1

⇒ match each rule whose premise contains a newly added literal

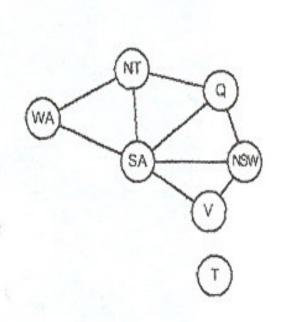
Matching itself can be expensive

Database indexing allows O(1) retrieval of known facts e.g., query Missile(x) retrieves  $Missile(M_1)$ 

Matching conjunctive premises against known facts is NP-hard

Forward chaining is widely used in deductive databases

## Hard matching example



 $Diff(wa, nt) \land Diff(wa, sa) \land$   $Diff(nt, q)Diff(nt, sa) \land$   $Diff(q, nsw) \land Diff(q, sa) \land$   $Diff(nsw, v) \land Diff(nsw, sa) \land$   $Diff(v, sa) \Rightarrow Colorable()$   $Diff(Red, Blue) \quad Diff(Red, Green)$   $Diff(Green, Red) \quad Diff(Green, Blue)$   $Diff(Blue, Red) \quad Diff(Blue, Green)$ 

Colorable() is inferred iff the CSP has a solution CSPs include 3SAT as a special case, hence matching is NP-hard