

# Feature Extraction with Description Logics Functional Subsumption

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Dan Roth

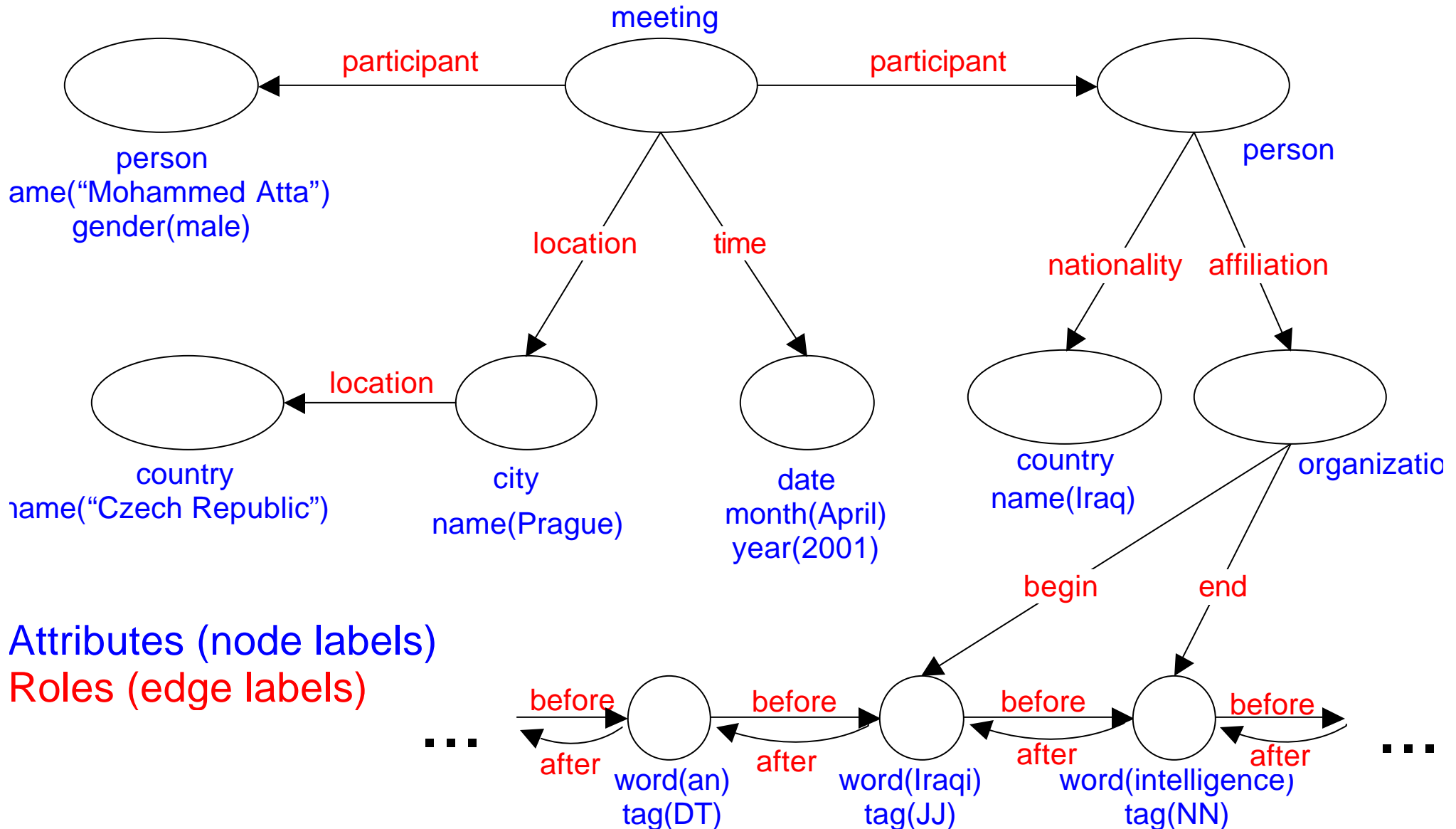
University of Illinois at Urbana-Champaign

# A conflict

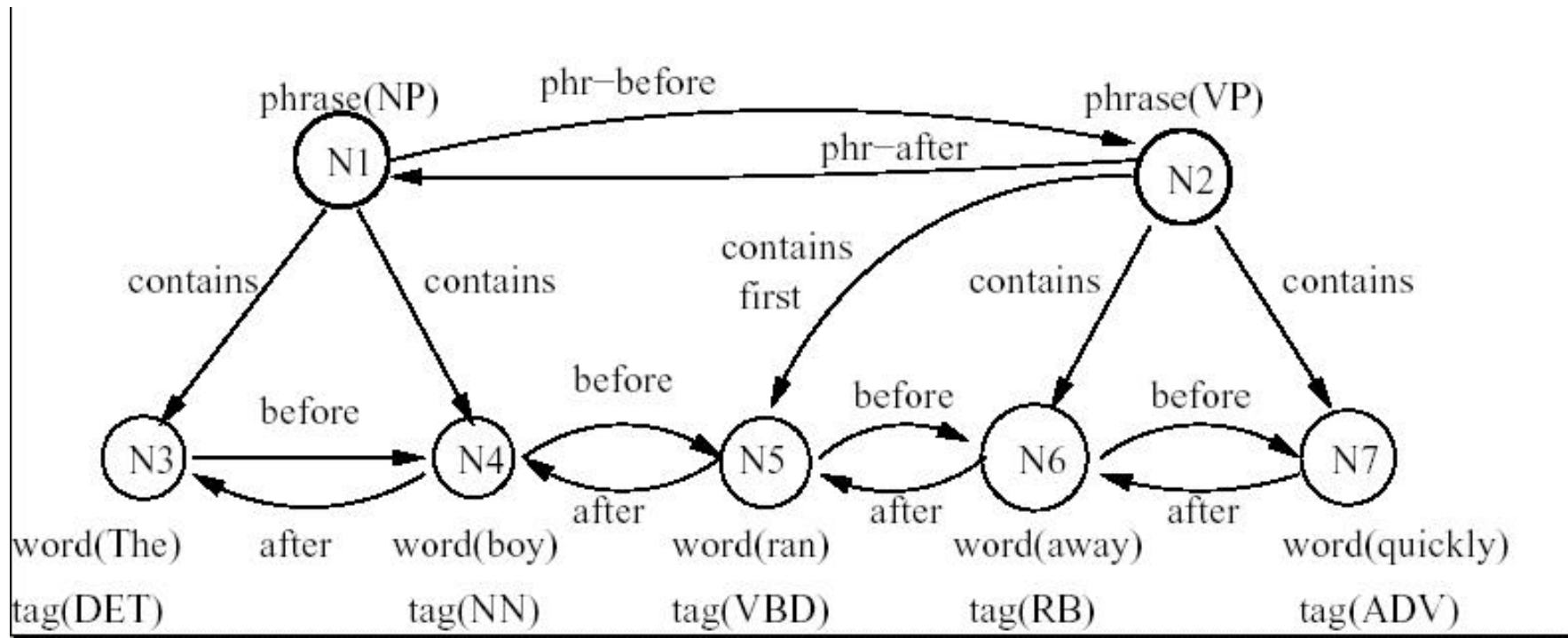
- ? Most machine learning algorithms use **feature vectors** as inputs.
- ? Most data is best represented as **structured data**.
- ? **Feature extraction** is the conversion from one to the other (and may be most of the work).

# Structured data – I

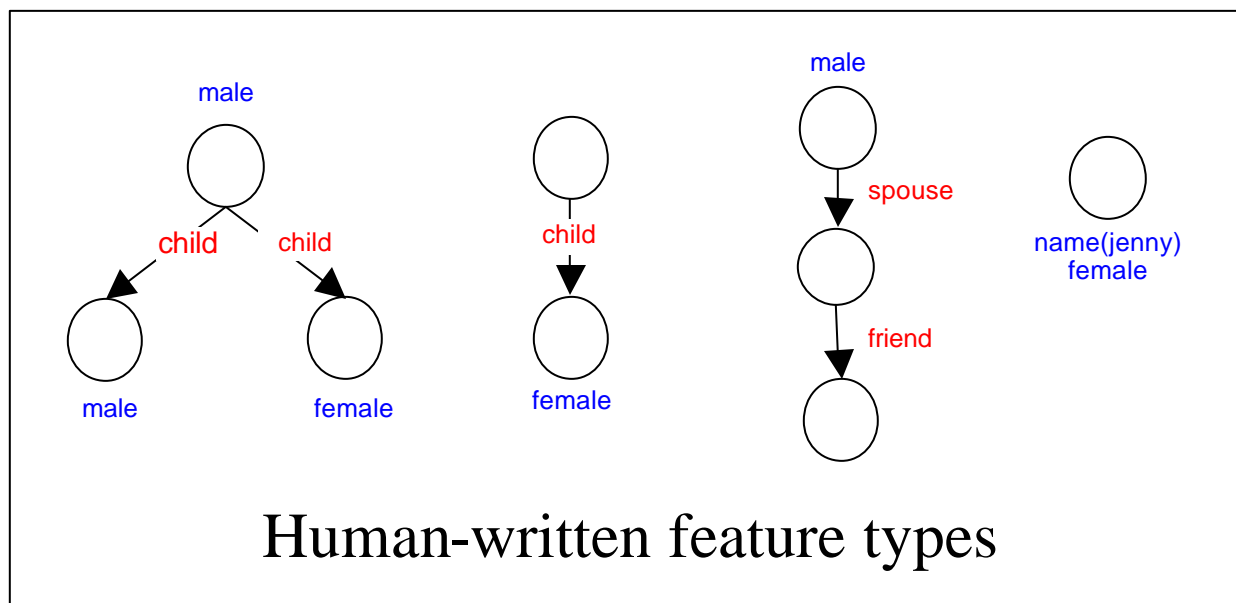
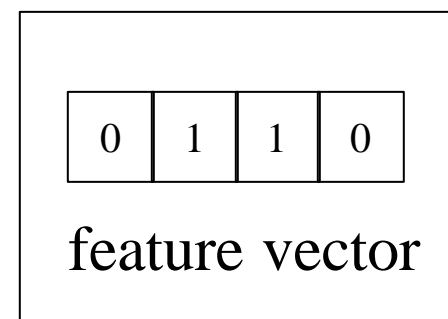
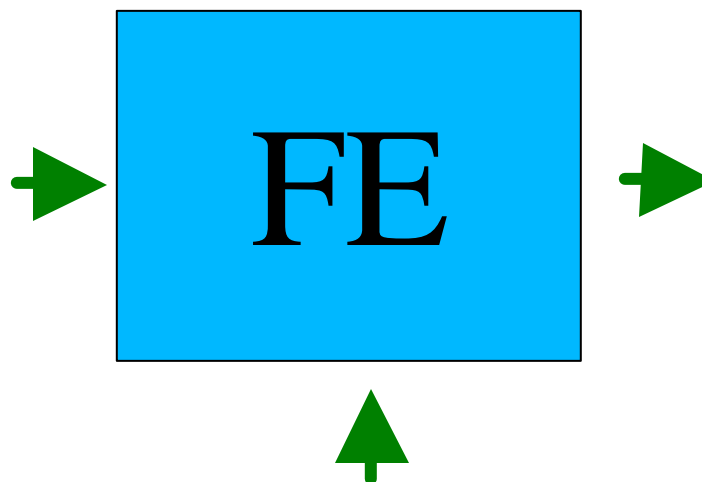
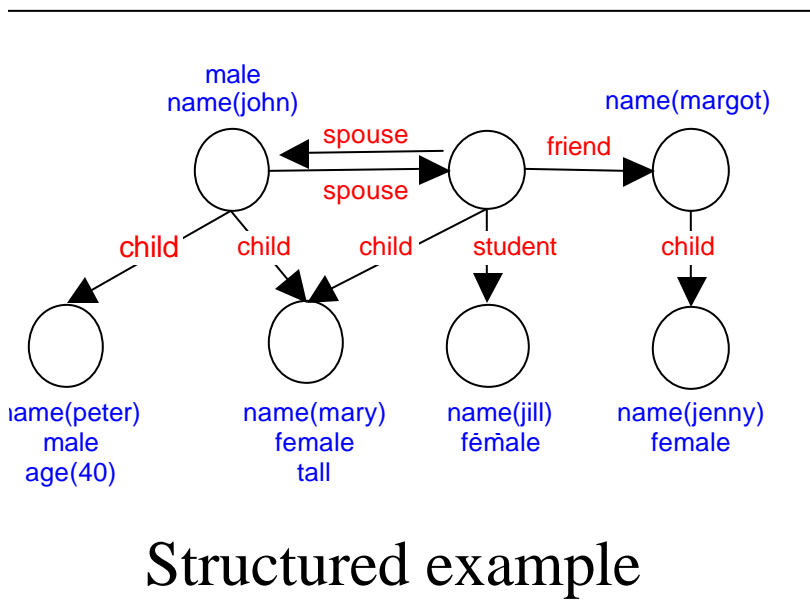
Mohammed Atta met with an Iraqi intelligence agent in Prague in April 2001



# Structured data – II



# Feature Extraction

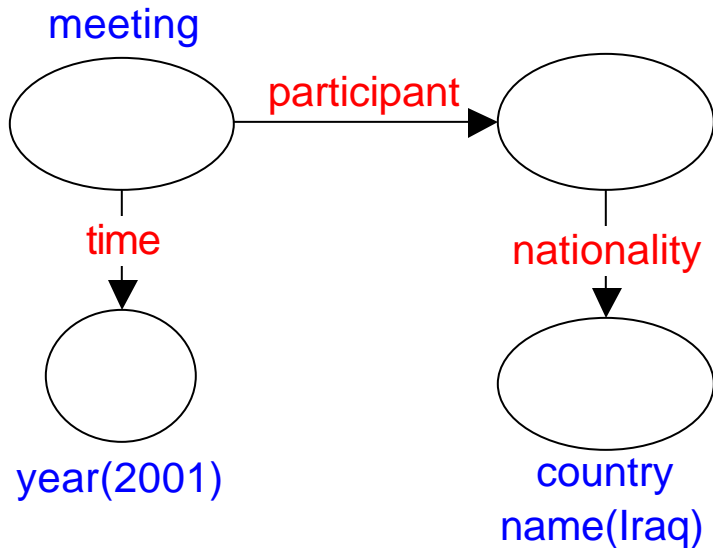


# Feature Extraction

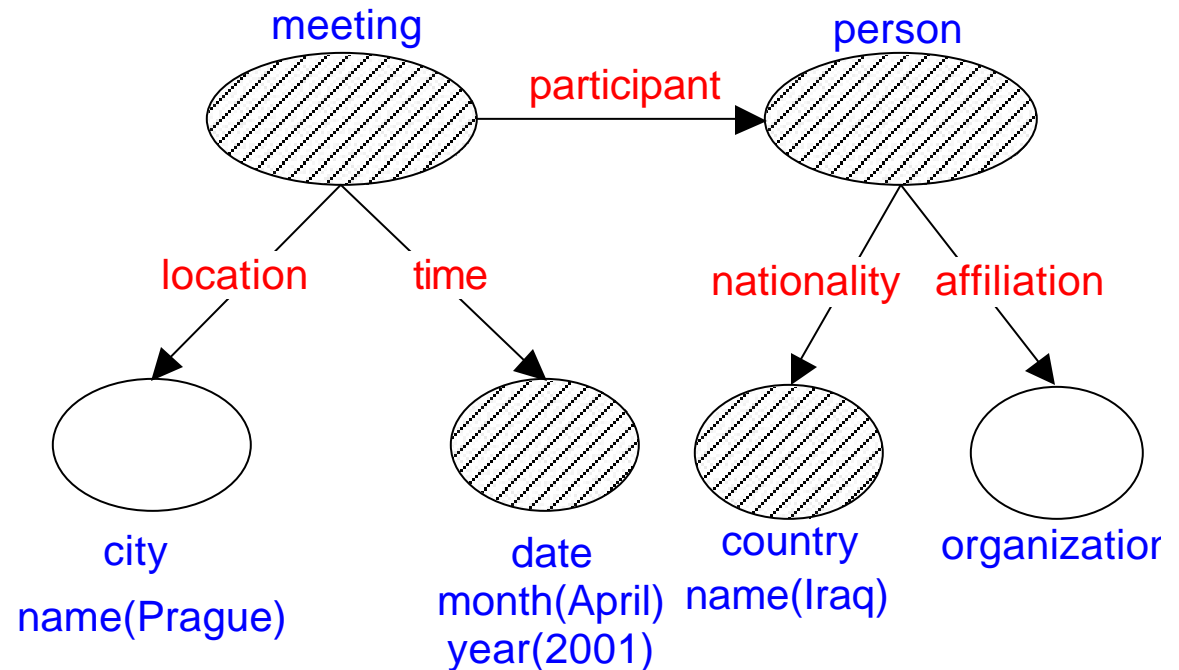
- ? *Typically done in ad hoc fashion:*
  - ? *Prevents general analysis;*
  - ? *Prevents Feature Extraction/Learning unified analysis (e.g. kernels).*
- ? *Using a language is tricky*
  - ? *Type of **inference**.*
  - ? *May be **intractable** if not careful.*

# A language for declaring which features to generate

Feature type specifications by  
**directed trees**

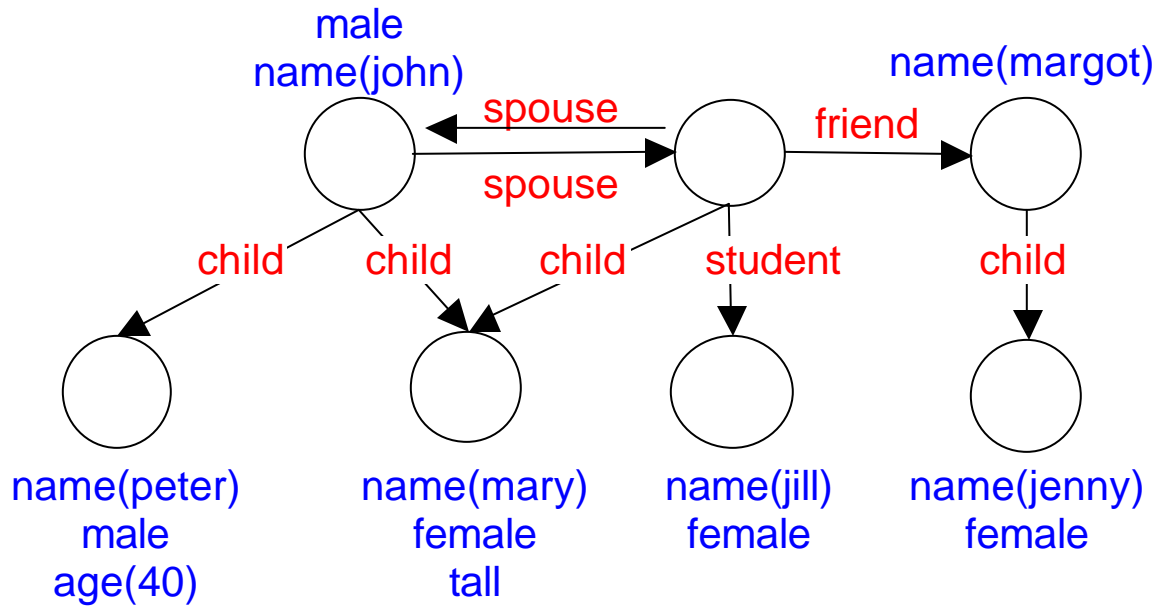


Example segment

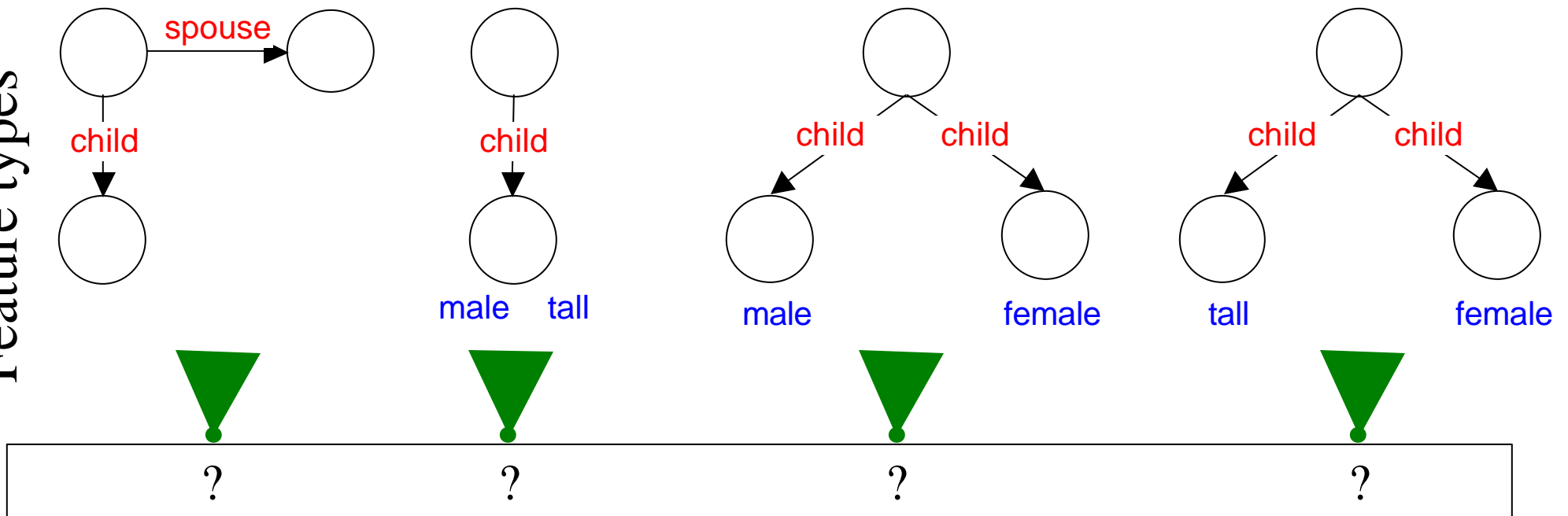


# Generating feature vectors

Example



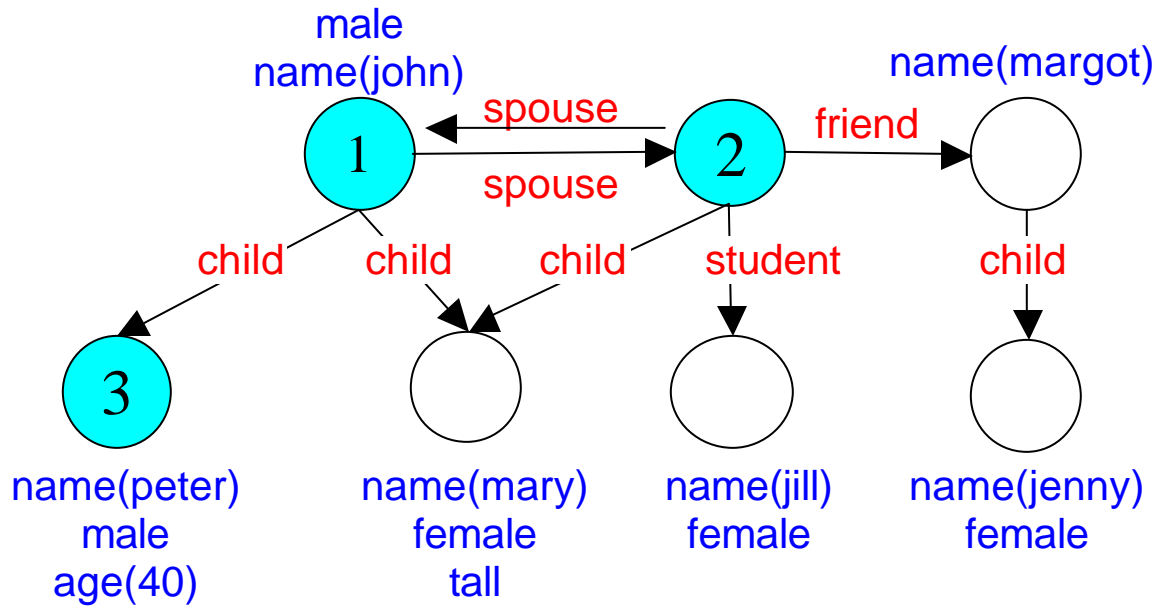
Feature types



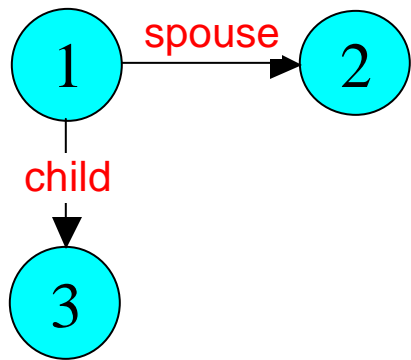


# Generating feature vectors

Example



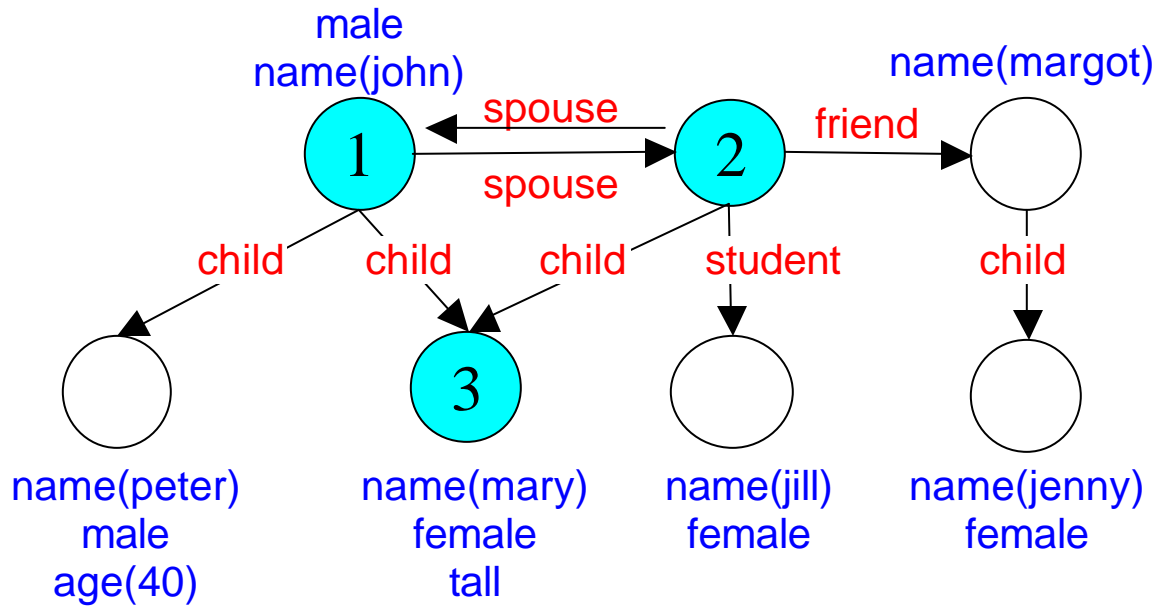
Feature types



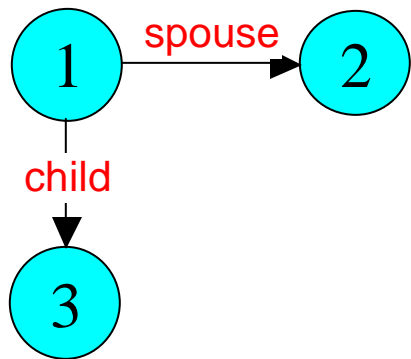
1

# Generating feature vectors

Example



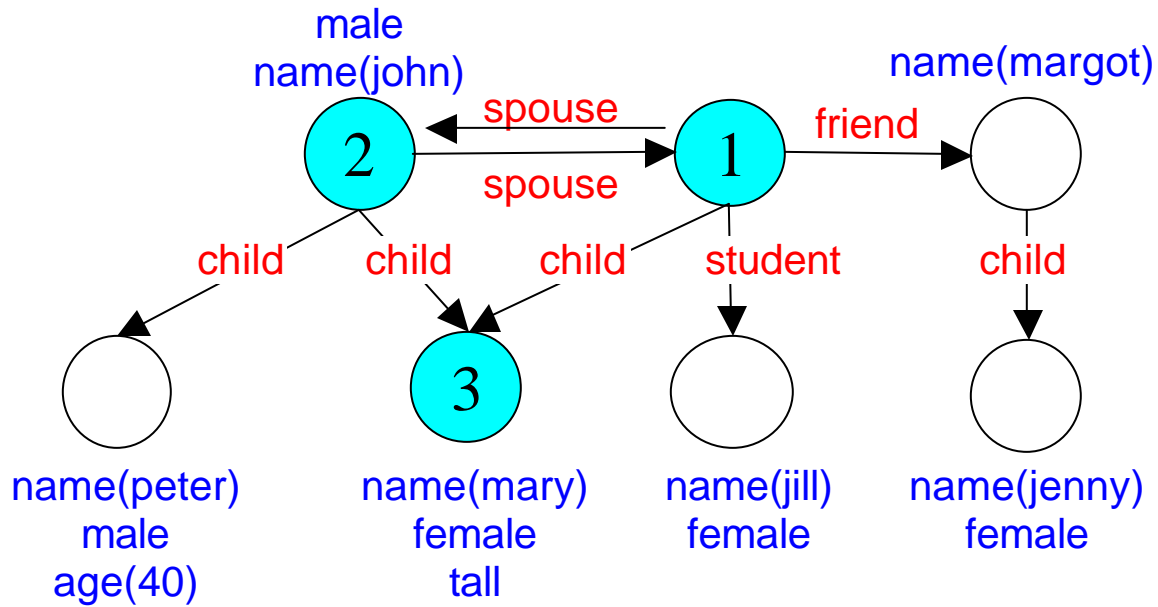
Feature types



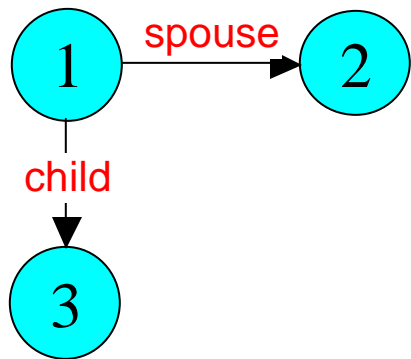
1

# Generating feature vectors

Example



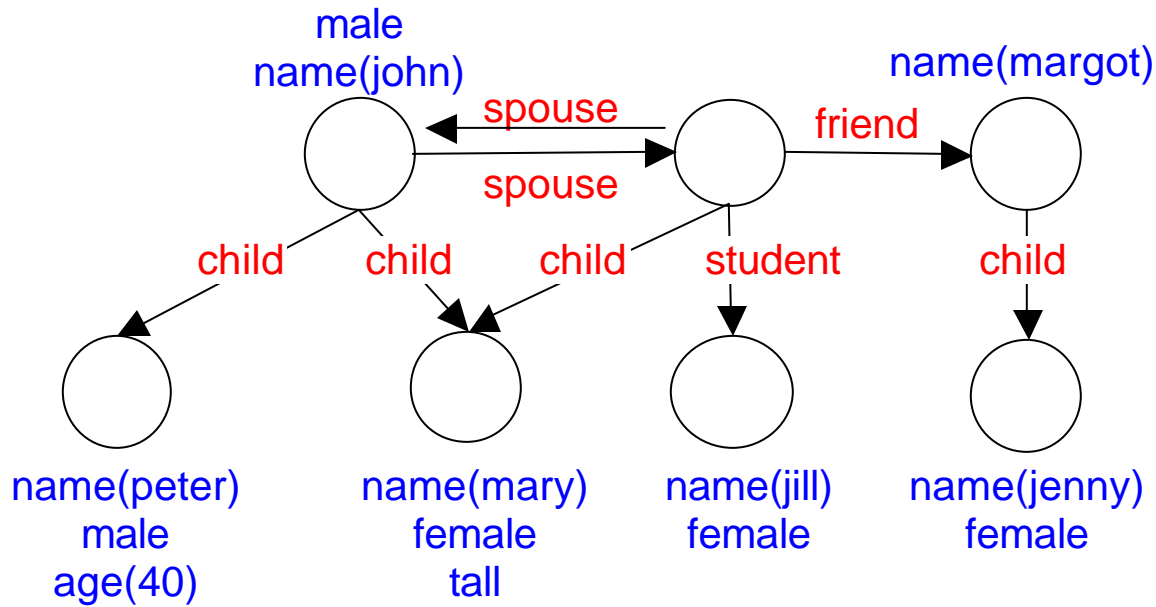
Feature types



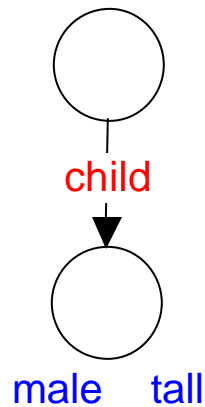
1

# Generating feature vectors

Example



Feature types

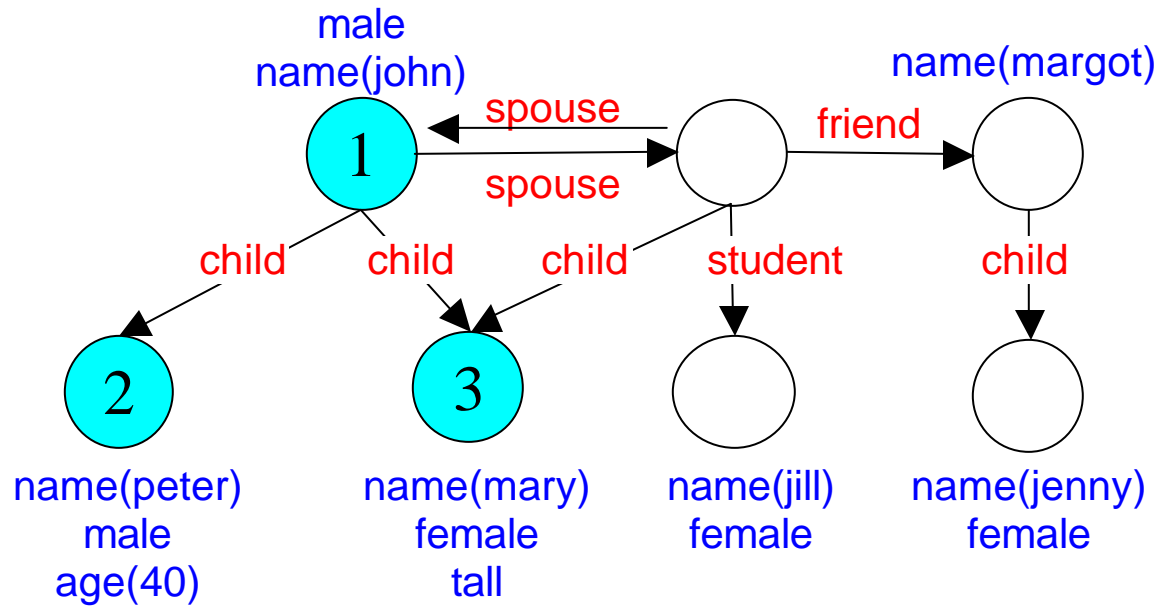


Nothing like this  
in the example!

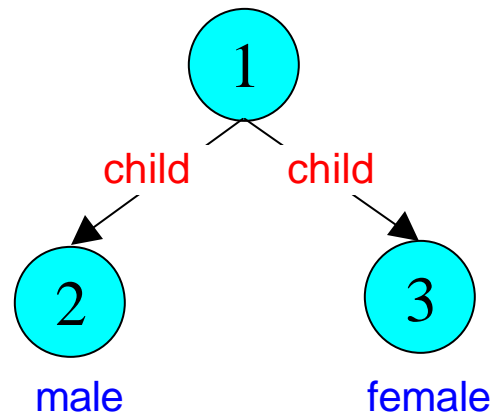
0

# Generating feature vectors

Example



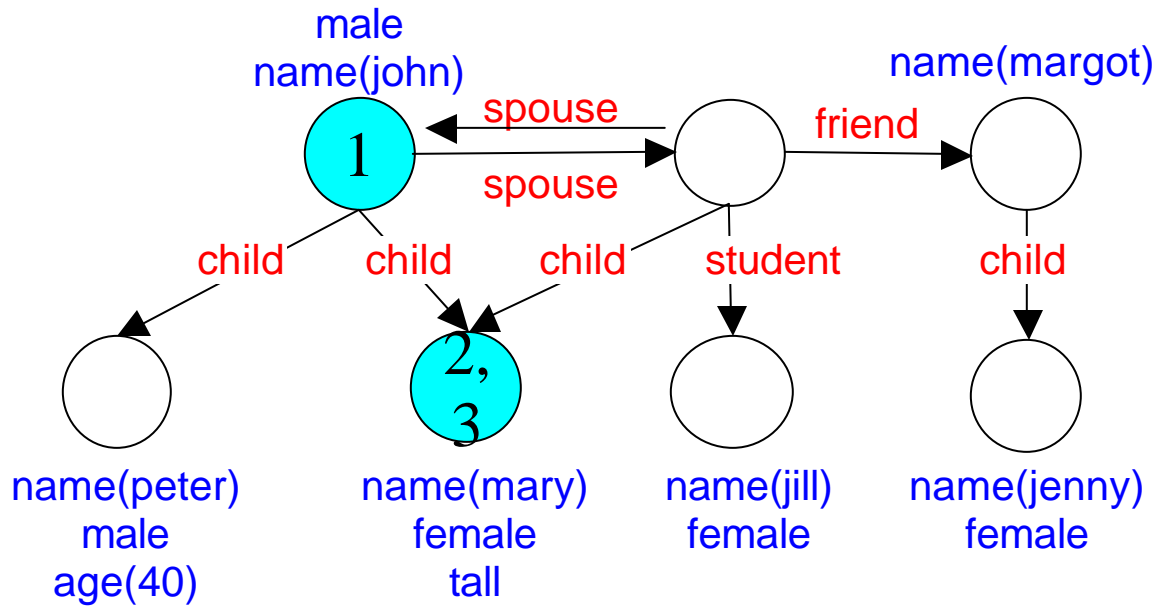
Feature types



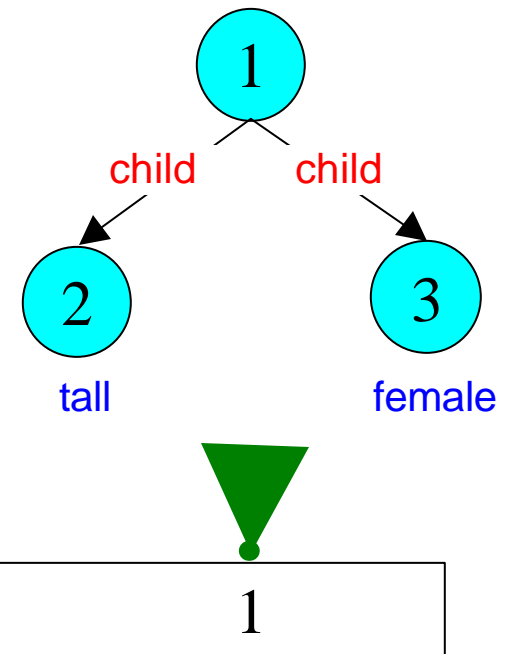
1

# Generating feature vectors

Example

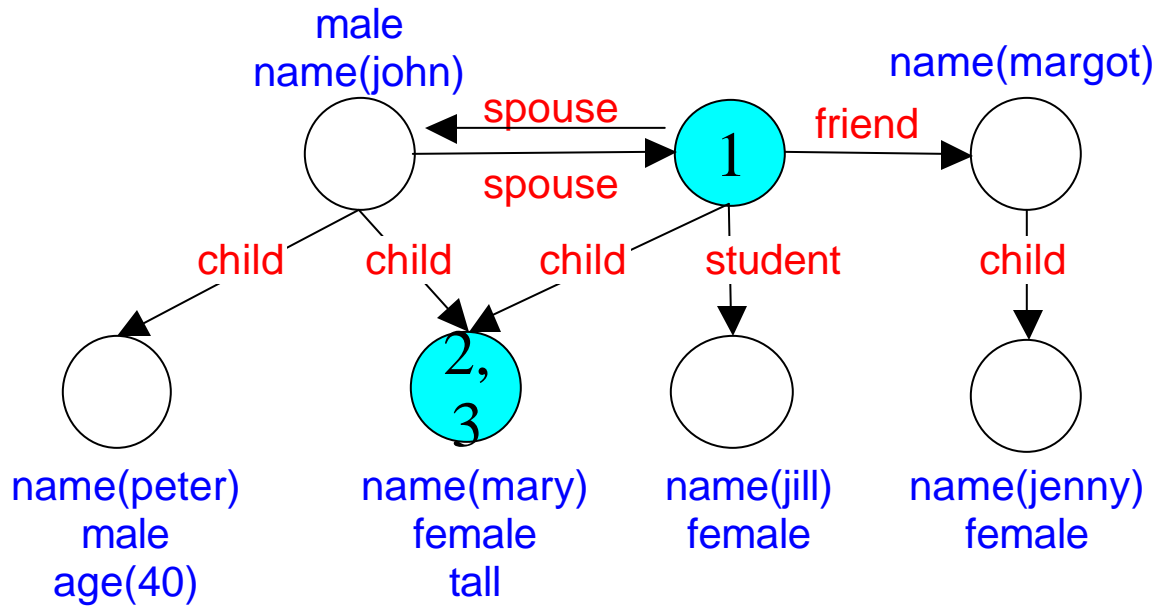


Feature types

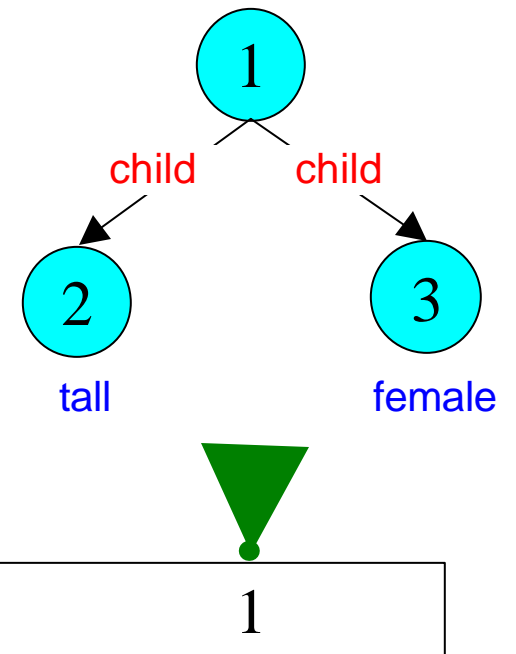


# Generating feature vectors

Example

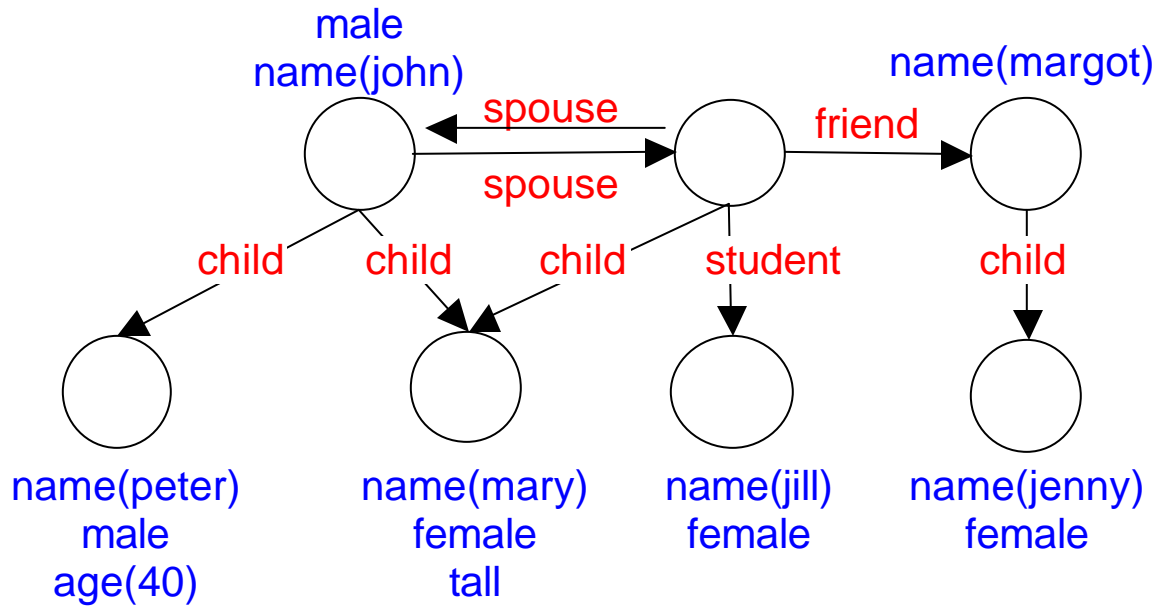


Feature types

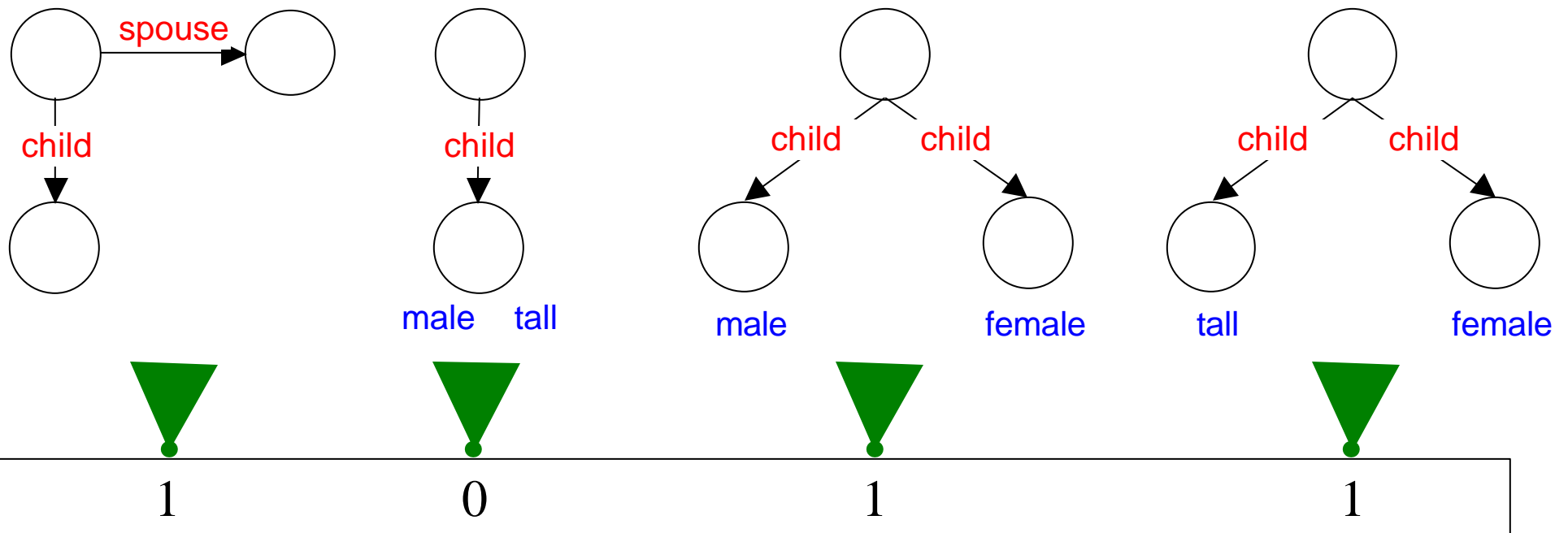


# Generating feature vectors

Example

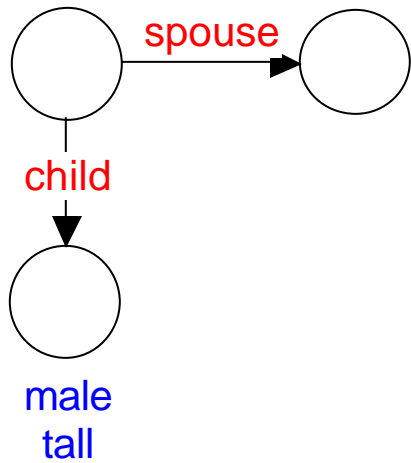


Feature types

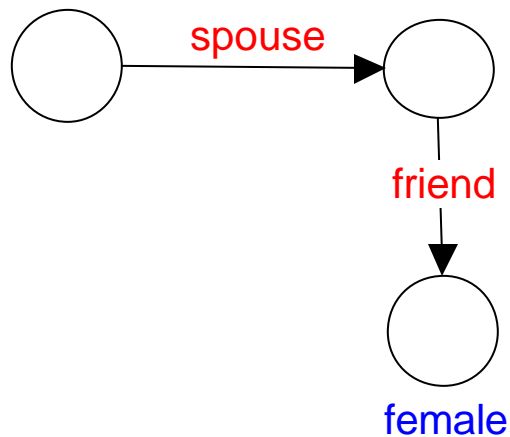




# Feature Description Logics



(AND  
(SOME spouse ANY)  
(SOME child (AND male tall))))



(SOME spouse  
(SOME friend female))

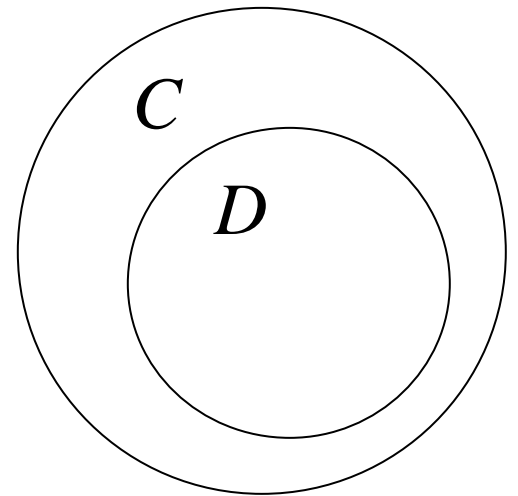
# Subsumption

?A description  $C$  *subsumes* ( $\supseteq$ ) a description  $D$  if every individual in  $D$  must be in  $C$ , no matter the interpretation.

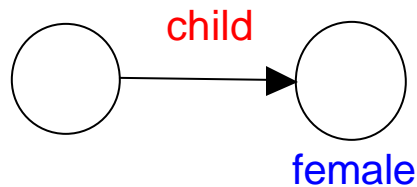
?Subsumption is tractable.

$C = (\text{AND}$   
    (SOME spouse ANY)  
    (SOME child male))

$D = (\text{AND}$   
    (SOME spouse (SOME student ANY))  
    (SOME child (AND tall male))  
    (SOME child female))



# Feature extraction as subsumption



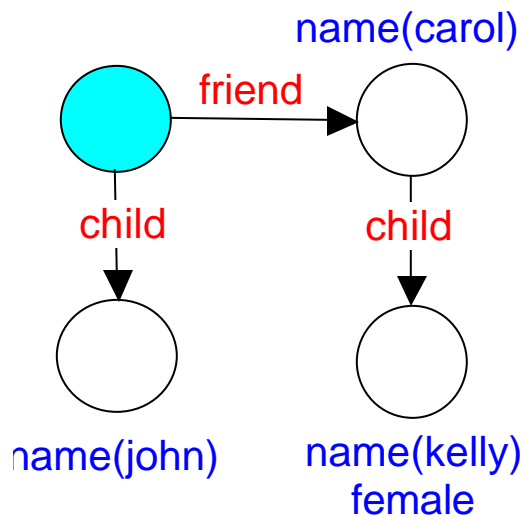
(SOME **child** female)

Feature type

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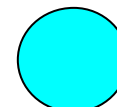
$\neq$

Example

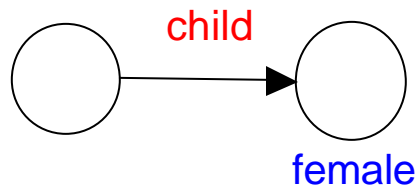


(AND  
SOME **friend**  
(AND  
name(carol)  
SOME **child** (AND name(kelly) female))  
SOME **child** name(john))

Description of node



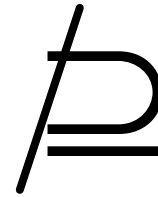
# Feature extraction as subsumption



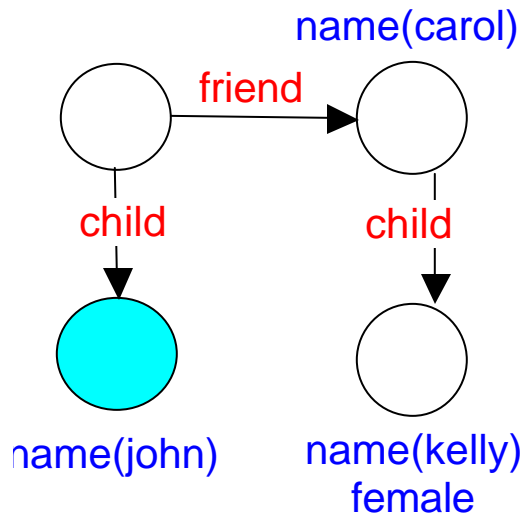
(SOME **child** female)

Feature type

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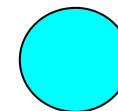


Example

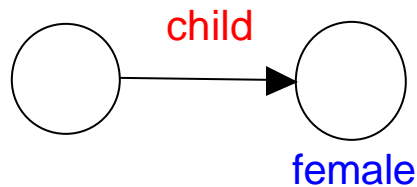


name(john)

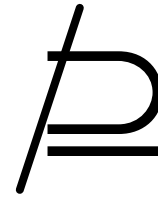
Description of node



# Feature extraction as subsumption



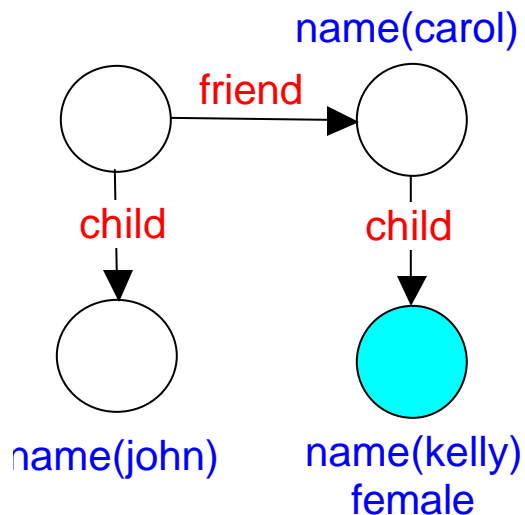
(SOME **child** female)



Feature type

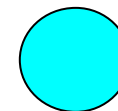
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Example

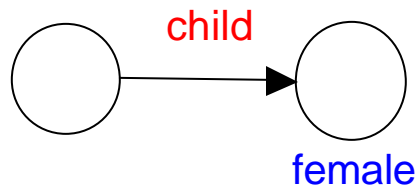


(AND **name(kelly)** female)

Description of node

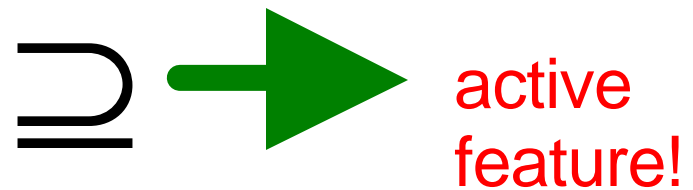


# Feature extraction as subsumption

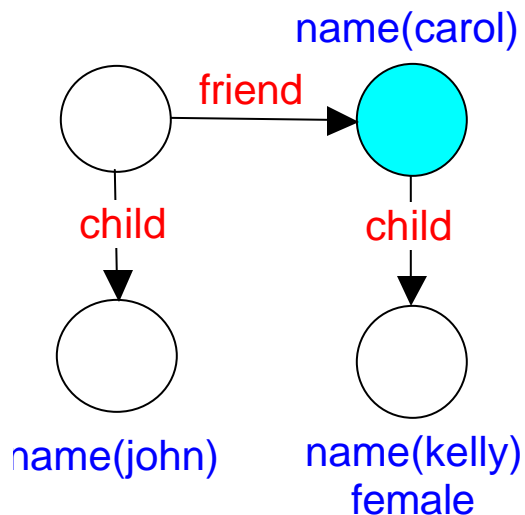


Feature type

(SOME **child** female)



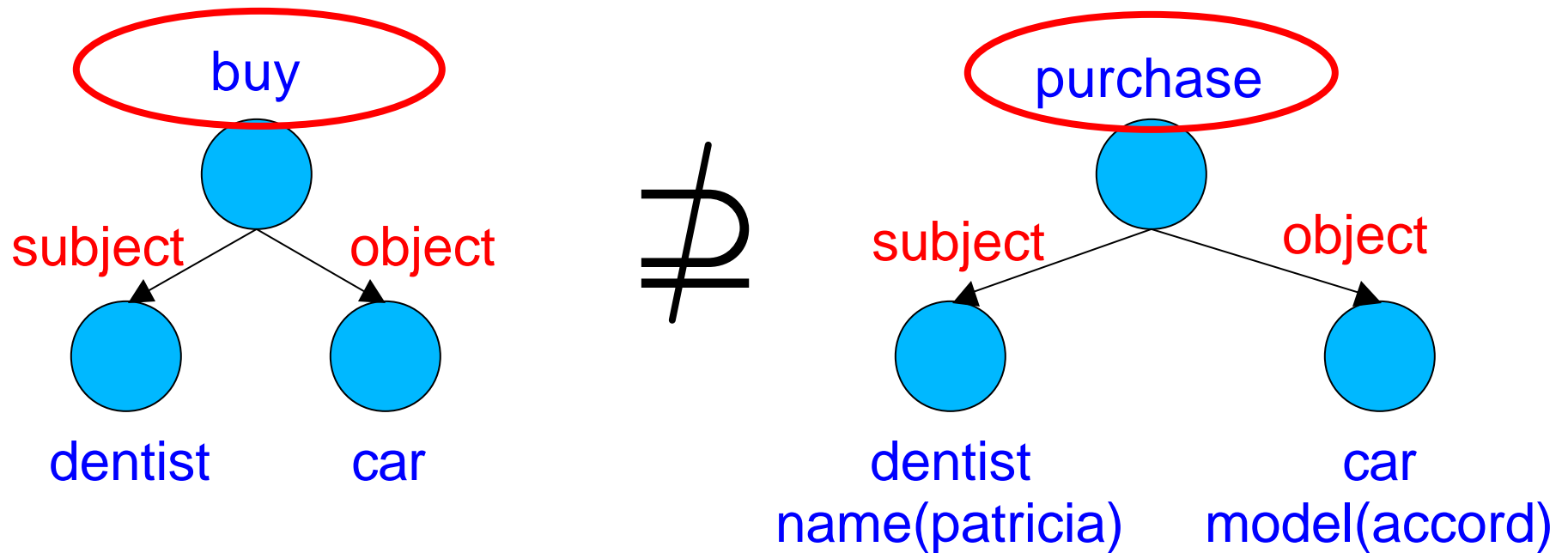
Example



(AND  
name(carol)  
SOME **child** (AND name(kelly) female))

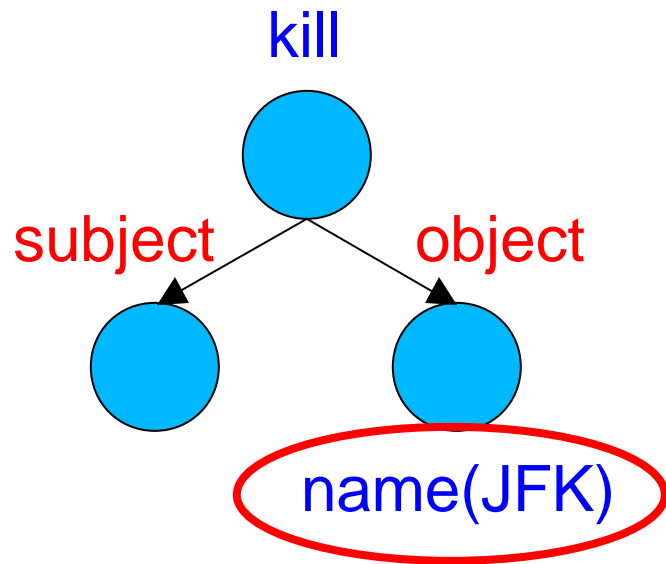
Description of node 

# A problem in practice

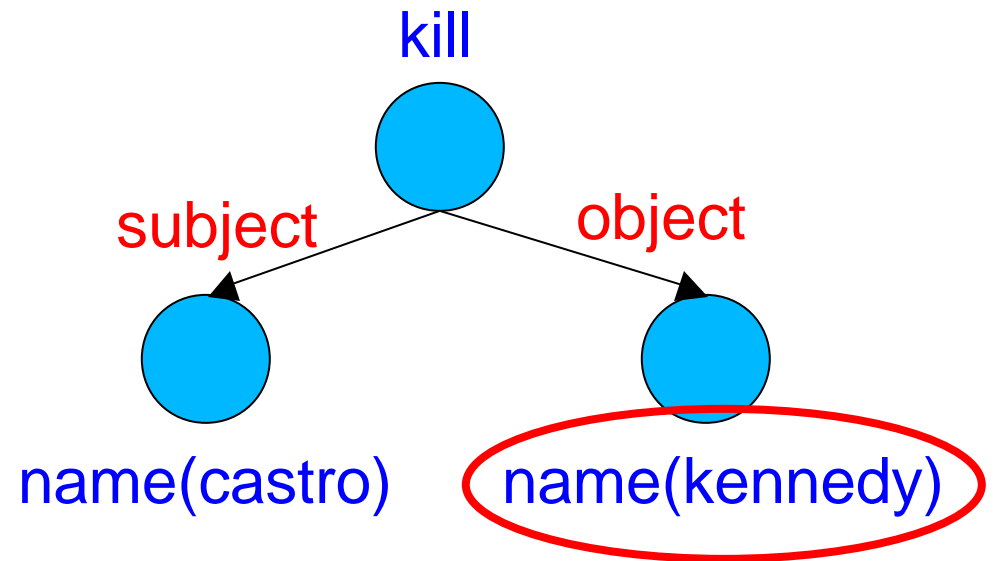


Subsumption would be natural in this case but does not occur

# A problem in practice



$\neq$





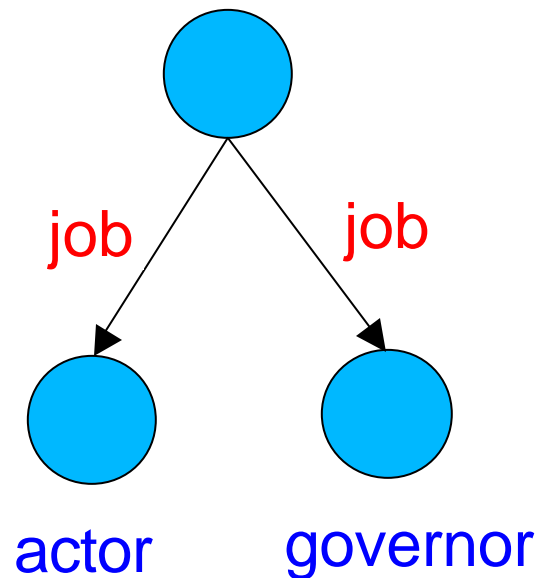
# A problem in practice

name(schwarzenegger)



$\neq$

name(schwarzneger)



# Make comparison more flexible

? At core of subsumption algorithm is the comparison of attributes:

... if (`attr1 == attr2`) ...

? We simply make that a function call:

... if (`f (attr1, attr2) == 1`) ...

Is this just a hack?

What about the nice DL **semantics**?

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in fact, equivalent to “shallow OR” (*tractable*).

# Is this just a hack?

What about the nice DL semantics?

in fact, equivalent to “shallow OR” (*tractable*).

Replace any *attr* by (OR  $a_1$   $a_2$  ...  $a_n$ )

where  $f(\text{attr}, a_i) = 1$ .

AND *kill*

(SOME *object* JFK))

AND (OR *kill* *murder*  *assassinate*)

(SOME *object* (OR

JFK *kennedy* “John F. Kennedy” ...))

# Why not just use shallow OR then?

- ? Function is an implicit representation.
- ? We may incorporate procedural knowledge:
  - ? Typos;
  - ? Similar sounding words;
  - ? Context-sensitive knowledge.

# Take home message

- ? Feature Description Logics provides an expressive way to deal with structured examples.
- ? Syntax choices render it tractable.
- ? Allows for FE-learning integrated approaches like kernels (Cumby & Roth 2003).
- ? Can be made even more expressive with little extra cost by functional subsumption.

The End