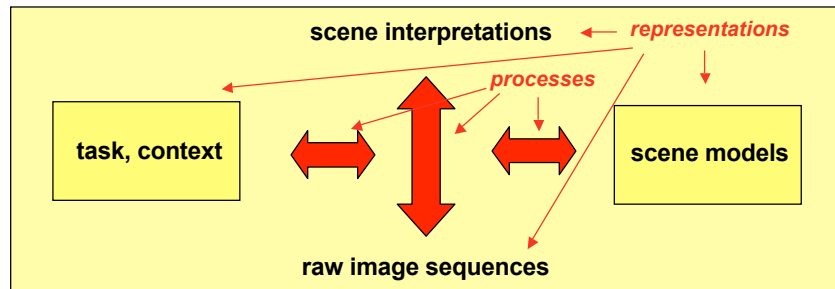


Representations and Processes in Knowledge-based Systems



Characteristics of ideal knowledge-based systems:

- Problems are specified by background and task knowledge using a declarative knowledge representation language
- Problems are solved using standard inference procedures

Knowledge representation formalisms must support representations and processes (inferences)!

1

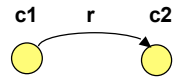
Basic Knowledge Representation Formalisms

- Semantic Networks
- Frames
- Constraints
- Relational Structures

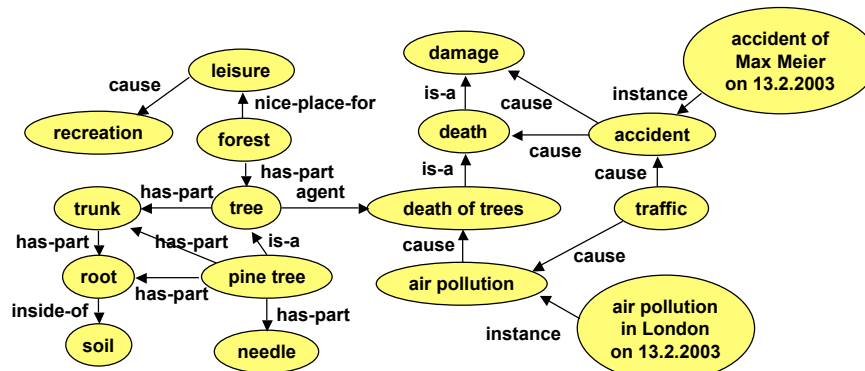
2

Semantic Networks

Graphical representation of binary relations:
 labelled nodes = concepts or individuals
 directed labelled edges = binary relations

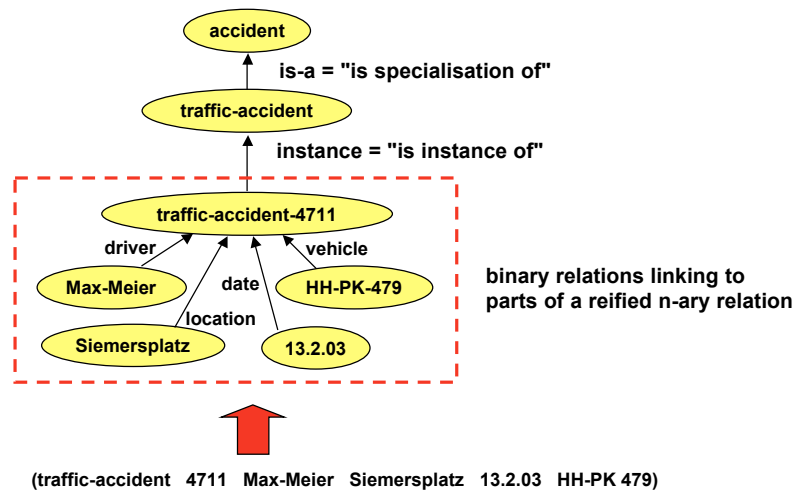


Semantic networks were originally developed to model associations between concepts in the human mind.



3

Basic Relations in Semantic Networks



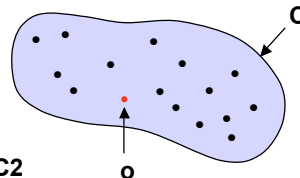
4

Concepts and Instances

Nodes of a semantic network describe concepts and individuals.

A concept denotes a set of objects.

An individual denotes a single object.



C1 is-a C2 specifies that C1 is a subset of C2

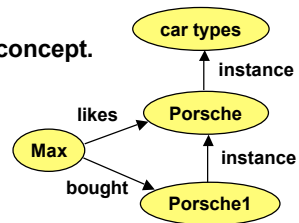
o instance C specifies that o is a member of C

A node may represent both, an individual and a concept.

Example:

Max likes a Porsche.

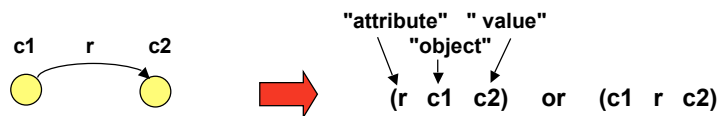
Max bought a Porsche at the car dealer.



5

Attribute-Object-Value Triplets

In knowledge representation languages and programming languages, a semantic network can be represented by a set of triplets:



The accident example:

(is-a traffic-accident accident)
 (instance traffic-accident-4711 traffic-accident)
 (driver traffic-accident-4711 Max-Meier)
 (location traffic-accident-4711 Siemersplatz)
 (date traffic-accident-4711 13.2.03)
 (vehicle traffic-accident-4711 HH-PK-479)

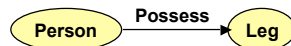
Note:

- notions of attribute, object and value do not always seem fitting
- notation is not object centered

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Unclear Semantics of Semantic Webs

For nodes C1 and C2 representing concepts, the meaning of (r C1 C2) is not clear. Example:



1. All persons possess at least one leg
2. All persons possess exactly one leg
3. All persons possess zero to infinity legs
4. All objects which a person possesses are legs
5. Possessing means that a person has a least one leg
6. Possessing means that a person has exactly one leg
7. Possessing means that a person has zero to infinity legs
8. There is at least one person which possesses at least one leg
9. There is at least one person which possesses exactly one leg
10. There is at least one person which possesses zero to infinity legs
11. Every leg is possessed by at least one person
12. Every leg is possessed by exactly one person
13. Every leg is possessed by zero to infinity persons
- ...

7

Physical Object Descriptions

Scene interpretations are based on information about physical objects. Hence concepts and relations about the "physical reality" are important.

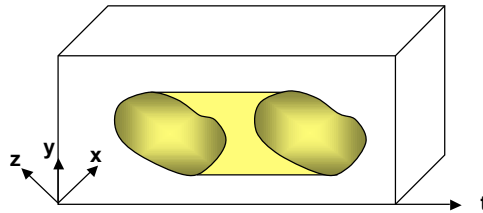
Characteristics of physical object descriptions:

- 1) **Individuals**
 - A description is valid for an absolute time point or period
 - Nodes denote individual physical objects (or object parts)
 - Objects have a spatial extent
 - Objects often have a shape and appearance
 - Objects are often described in terms of location and orientation
 - Objects obey physical laws
- 2) **Concepts**
 - Concepts define equivalent classes by abstracting from individual properties
 - Abstractions may be defined in terms of qualitative properties
 - Abstractions may involve relations to other objects

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Objects in Space-Time

An epistemologically well-founded way of defining an individual physical object is in terms of a subspace of 4-dimensional space-time.



Example: The potatoe lying on the Max Meier's table from 11:45 until 12:10 on August 8, 2003.

Individual physical objects which keep their identity over time constitute a common kind of abstraction.

Example: Max Meier is considered an individual in spite of his changes over time.

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Frames

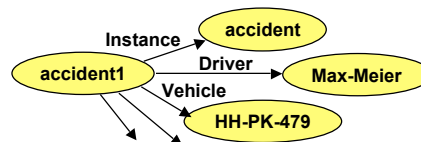
Frames have been proposed as knowledge representation structures for representing interrelated knowledge in larger units.

Marvin Minsky: "A Framework for Representing Knowledge", 1975

Simple frame structure for the individual "accident1":

| | |
|----------------|----------------|
| ID: | accident1 |
| Instance: | accident |
| Driver: | Max-Meier |
| Vehicle: | HH-PK-479 |
| Location: | Siemersplatz |
| Date: | 13.2.03 |
| Damage: | 5000-EUR |
| Police Report: | HH-2003-AX4711 |
| Witness: | Karl-Kruse |

- Slots represent binary relations:



- Slot fillers may be primitives or frames
- Inheritance and other inference services may be provided

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Facets and Procedural Attachment

- **Slot values are typed by "meta-types" and organized into facets.**
 - actual value
 - default (preset) value
 - value obtainable by procedure call
 - value obtainable by user inquiry
- **Attached procedures provide knowledge services as "demons" (without explicit invocation).**
 - for value computation
 - for checks before value change
 - for activities after value changes

Treat frames as plane frames but get more for value for your money!

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Frame Representation Language FRL

- **Facet names specify different slot filler "metatypes":**
 - \$DATA** normal data
 - \$DEFAULT** default values
 - \$IF-ADDED** write-access triggers specified demon procedures
 - \$IF-NEEDED** read-access triggers specified demon procedures
 - \$REQUIRE** demon procedures check conditions which must be met by slot fillers
- **Built-in inference services enriched by demon procedures**

Example:

| | |
|--------------|---|
| ID: | (\$DATA Person007) |
| Is-a: | (\$DATA Person) |
| Name: | (\$DATA Max-Meier) |
| Age: | (\$REQUIRE Agetest) (\$DATA 27) |
| Nationality: | (\$DEFAULT German) |
| Hobbies: | (\$DATA Eating, Sleeping, Singing) (\$IF-ADDED Singing Notify-Uni-Choir) |
| Phone: | (\$IF-NEEDED Directory-Retrieval-Service) |
| Address: | (\$DATA Address4711) |

Values are retrieved

1. from \$DATA facet
2. by inheritance from parent \$DATA facets
3. from \$DEFAULT facet
4. by inheritance from parent \$DEFAULT facets
5. by \$IF-NEEDED demon procedures

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Semantics of Frames

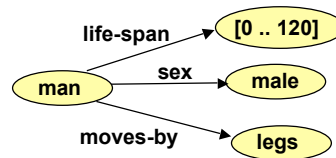
Frames as concept descriptions

| | |
|------------|------------|
| man | |
| moves-by: | legs |
| life-span: | [0 .. 120] |
| sex: | male |

semantics as



(man moves-by legs)
 (man life-span [0 .. 120])
 (man sex male)



Do we mean:

"Any object moving by legs, life-span 0 to 120 and sex male is a man"

or

"If an object is a man, it moves by legs, has life-span 0 to 120 and sex male"

The semantics of frames is not well-defined!

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Constraints

Constraints express restrictions on the values of variables.

Given variables and constraints, a constraint satisfaction problem (CSP) is the task of assigning values to the variables such the constraints are satisfied.

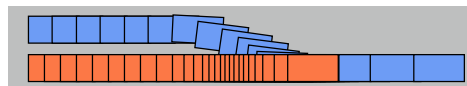
Constraints are useful for knowledge representation and inferencing:

- Constraints may provide a compact representation for n-ary relations
 Example: $\text{sum}(\text{action1.duration}, \text{action2.duration}, \text{action3.duration}) \leq 120 \text{ sec}$
- Spatial and temporal constraints are important for scene interpretation

Is this a cover?



Is this a forced brake ("Ausbremsen")?



- There exist efficient algorithms for solving special CSPs
- Constraints support flexible interpretation strategies

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Constraints for Scene Interpretation (1)

Constraints may model conceptual knowledge:

```
properly-parked-car-group
parts:    f-car is-a car
          b-car is-a car
          r-car is-a car
constraints: behind (b-car, f-car)
             behind (r-car, b-car)
             distance (f-car, b-car) ≥ 30cm
             distance (r-car, b-car) ≥ 30cm
```

Constraints may express concrete knowledge about a scene:

```
length(car1) ≤ 300cm
behind (car1, car2)
behind (car3, car2)
distance (car1, car2) = 42cm
distance (car3, car2) ≥ 400cm
```



Constraints may express inferred knowledge about a scene:

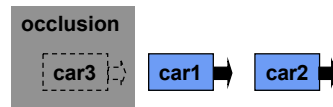
```
behind (car3, car1)
distance (car3, car1) ≥ 100cm
```

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Constraints for Scene Interpretation (2)

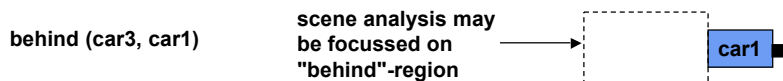
Constraints may restrict hypotheses:

```
properly-parked-car-group1
parts:    car2 is-a car
          car1 is-a car
          car3 is-a car
constraints: behind (car1, car2)
             behind (car3, car1)
             distance (car1, car2) = 42cm
             distance (car3, car1) ≥ 30cm
```



car3 is not visible but
may be hypothesized

Constraints may help to focus processing:



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Hard and Soft Constraints

- Hard constraints must be satisfied. A violated constraint prohibits a solution. The CSP is a satisfiability problem.
- Soft constraints should be satisfied. A violated constraint impairs the quality of a solution. The CSP is an optimization problem.

Constraints relevant for scene interpretation may have different origin:

Constraints arising from logics

- Examples:
- to be "relatives" persons must have a common ancestor
 - "same-object-as" requires that two objects are identical
 - "touches" implies "near"

Constraints arising from physical laws

- Examples:
- an object may not be at different places at the same time
 - different solid objects may not occupy the same place at the same time
 - "holding" requires that the holder is physically connected to the held object

Constraints arising from conventions or goal-directed behavior

- Examples:
- spatial constraints for a "cover" on a table
 - temporal constraints for a typical "overtake"
 - actions for inserting a CD into a CD-player

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Relational Models

Relational models describe objects (object classes) based on parts (components) and relations between the parts

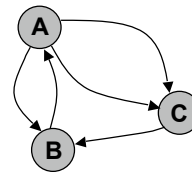
Relational model can be represented as structure with nodes and edges:

nodes: parts with properties

e.g.  The diagram shows the components of the letter 'A': a diagonal line, a horizontal bar, another diagonal line, and the full letter 'A'. A double-headed orange arrow points from the components to the letter 'A', indicating that these parts define the whole.

edges: relations between parts

- e.g.
- obtuse-angle
 - 2cm-distance
 - touches
 - surrounds
 - left-of
 - after



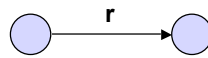
18

Relations between Components

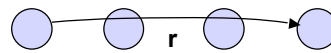
unary relation: property
 n-ary relation: relation, constraint

Graphical representation

binary relation:



n-ary relation:



"hypergraph"

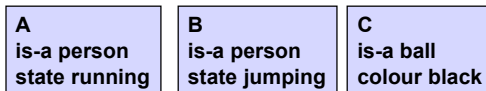
19

Relational Models for High-level Vision

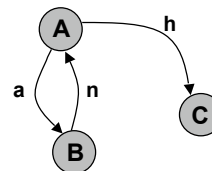
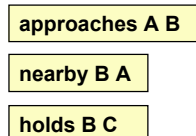
Relational models describe objects (object classes) based on parts (components) and relations between the parts

A relational model can be represented as a structure with nodes and edges:

Nodes: parts with properties



Edges: relations between parts




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Representing N-ary Relations

Awkward graphical representation:



N-ary relations can be transformed into binary relations:

(BETWEEN A B C)  (INSTANCE BETW1 BETWEEN)
 (BETWEEN-ARG1 BETW1 A)
 (BETWEEN-ARG2 BETW1 B)
 (BETWEEN-ARG3 BETW1 C)

(OVERTAKE VEH1 VEH2 23 46)  (INSTANCE OT1 OVERTAKE)
 (OVERTAKER OT1 VEH1)
 (OVERTAKEE OT1 VEH2)
 (TBEG OT1 23)
 (TEND OT1 42)

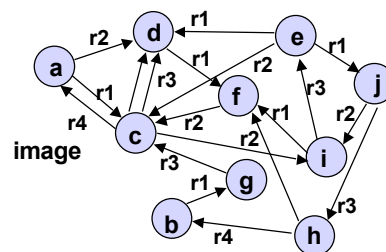
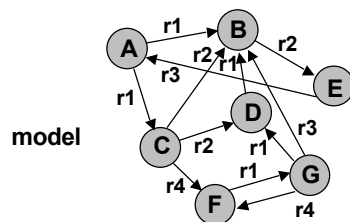
Transforming a relation into an object is called reification.

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Recognition by Relational Matching

Principle:

- construct relational model(s) for object class(es)
- construct relational image description
- compute morphism (best partial match) between image and model(s)



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Compatibility of Relational Structures

Different from graphs, nodes and edges of relational structures may represent entities with rich distinctive descriptions.

Example: nodes = image regions with diverse properties
edges = spatial relations

1. Compatibility of nodes

An image node is compatible with a model node, if the properties of the nodes match.

2. Compatibility of edges

An image edge is compatible with a model edge, if the edge types match.

3. Compatibility of structures

A relational image description B is compatible with a relational model M, if there exists a bijective mapping of nodes of a partial structure B' of B onto nodes of a partial structure M' of M such that

- corresponding nodes and edges are compatible
- M is described by M' with sufficient completeness

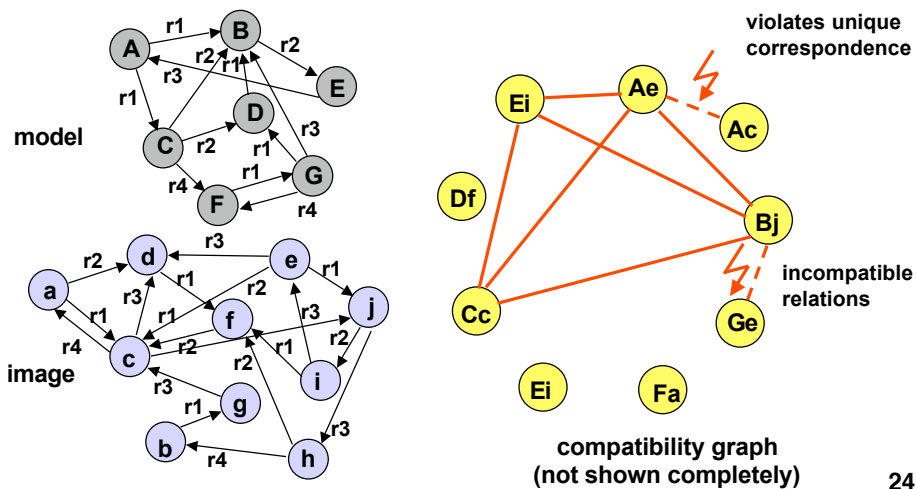
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Relational Matching with a Compatibility Graph

nodes of compatibility graph = pairs with compatible properties

edges of compatibility graph = compatible pairs

cliques in compatibility graph = compatible partial structures



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Finding Maximal Cliques

Algorithms are available in the literature, e.g.

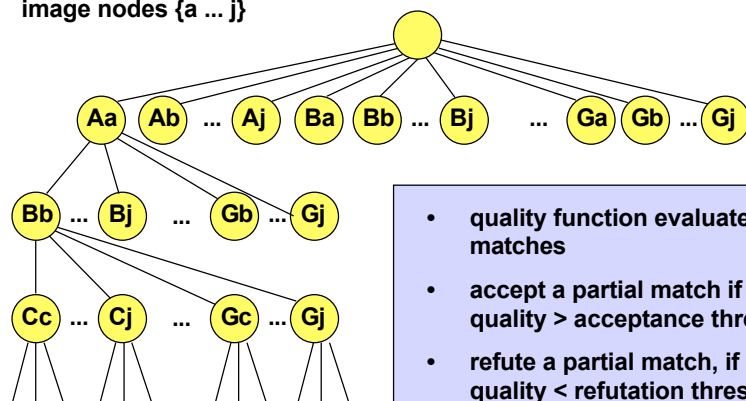
Bron & Kerbusch, Finding all Cliques of an Undirected Graph, Communications of the ACM, Vol. 16, Nr. 9, S. 575 - 577, 1973.

- Complexity is exponential relative to number of nodes of compatibility graph
- Efficient (suboptimal) solutions based on heuristic search

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Relational Matching with Heuristic Search

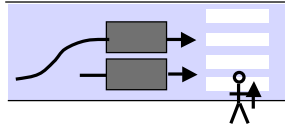
Stepwise correspondence search between model nodes {A ... G} and image nodes {a ... j}



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Example for Relational Event Recognition (1)

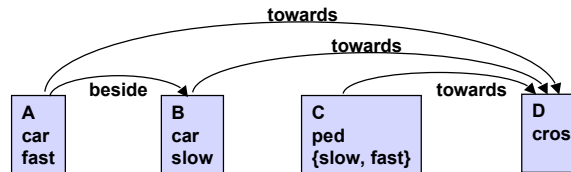
Suppose we want to recognise dangerous overtake events at pedestrian crossings:



Nodes: type: {car, pedestrian, plain_road, crossing}
speed: {zero, slow, fast}

Relations: {beside, behind, before, towards, away}

Model for dangerous overtake events:



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Example for Relational Event Recognition (2)

Is there a dangerous overtake event in the following exciting traffic scene?

| | |
|--------------|-------------|
| a car fast | a behind b |
| b car fast | b before a |
| c car slow | b towards k |
| d car slow | c away k |
| e ped fast | d towards k |
| f ped slow | b beside d |
| g ped slow | f towards l |
| h ped slow | g towards k |
| i ped slow | e beside h |
| k cros | i away k |
| l plain_road | h towards l |

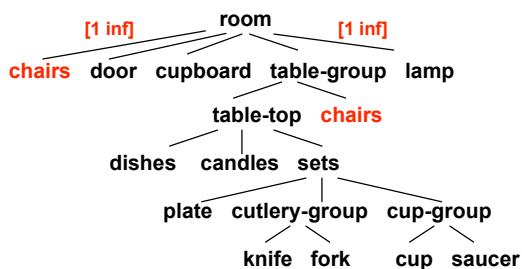
- Apply heuristic search!
- What heuristic may be useful?
- How can the approach be improved?

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Shortcomings of Relational Matching for Scene Interpretation (1)

Natural hierarchical structures and groupings are not well represented by flat relational structures.

Example: Modelling dining room views



Symbolic hierarchical models allow unique representations for repeated substructures, cardinality information for part-of relations and other features not available in flat relational models.

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Shortcomings of Relational Matching for Scene Interpretation (2)

Node compatibility is not clearly defined



Edge compatibility is not clearly defined

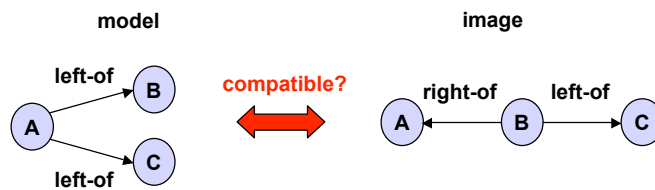


Logical relations between different node descriptions and different edge labels must be represented.

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Shortcomings of Relational Matching for Scene Interpretation (3)

Implicit information about the semantics of relations is not considered



Reasoning may be required to determine compatibility

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How Useful is Relational Matching?

- **Relational structure captures basic high-level notions**
 - **Graceful degradation w.r.t. completeness and degree of match**
 - **Well-understood computational procedures**
 - finding maximal cliques in compatibility graphs
 - heuristic search
 - constraint satisfaction
 - neural network implementations
 - **Improvement by hierarchical matching**
-
- **Multi-level aggregate structure required**
 - **Differentiated compatibility measure required**
 - fuzziness
 - compatibility vs. consistency
 - probabilities
 - **Reasoning about temporal, spatial, physical relations**
 - **Uncertainty management required**

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