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Lecture 5: Case-Based Reasoning and Its Relation with Fuzzy Systems

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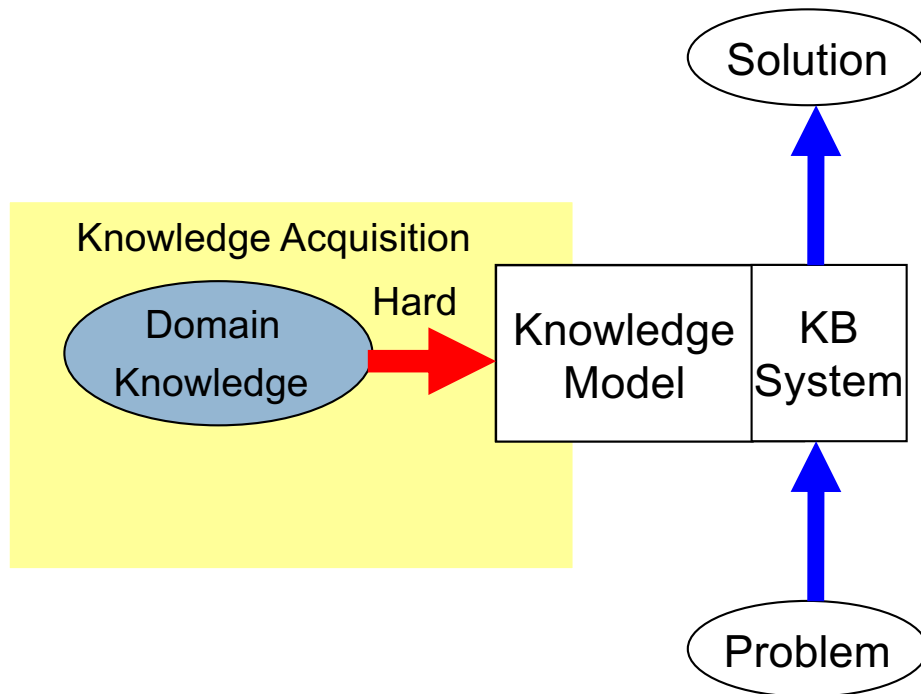
Agenda

- What is case-based reasoning (CBR) ?
- Practical techniques in CBR
- The relation of CBR with fuzzy systems

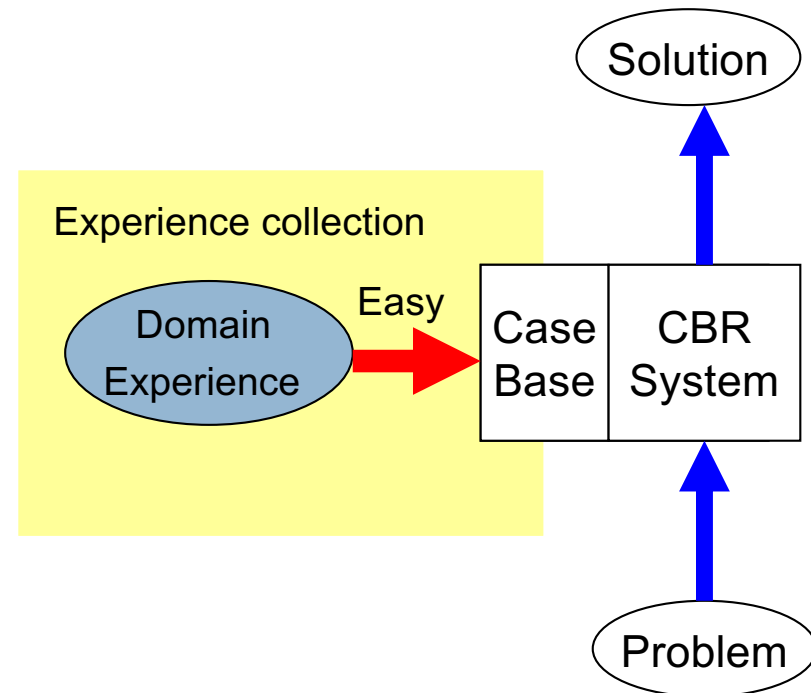
What is Case-Based Reasoning

What is Case-Based Reasoning (CBR)

Traditional Knowledge-Based systems



Case-Based Reasoning Systems

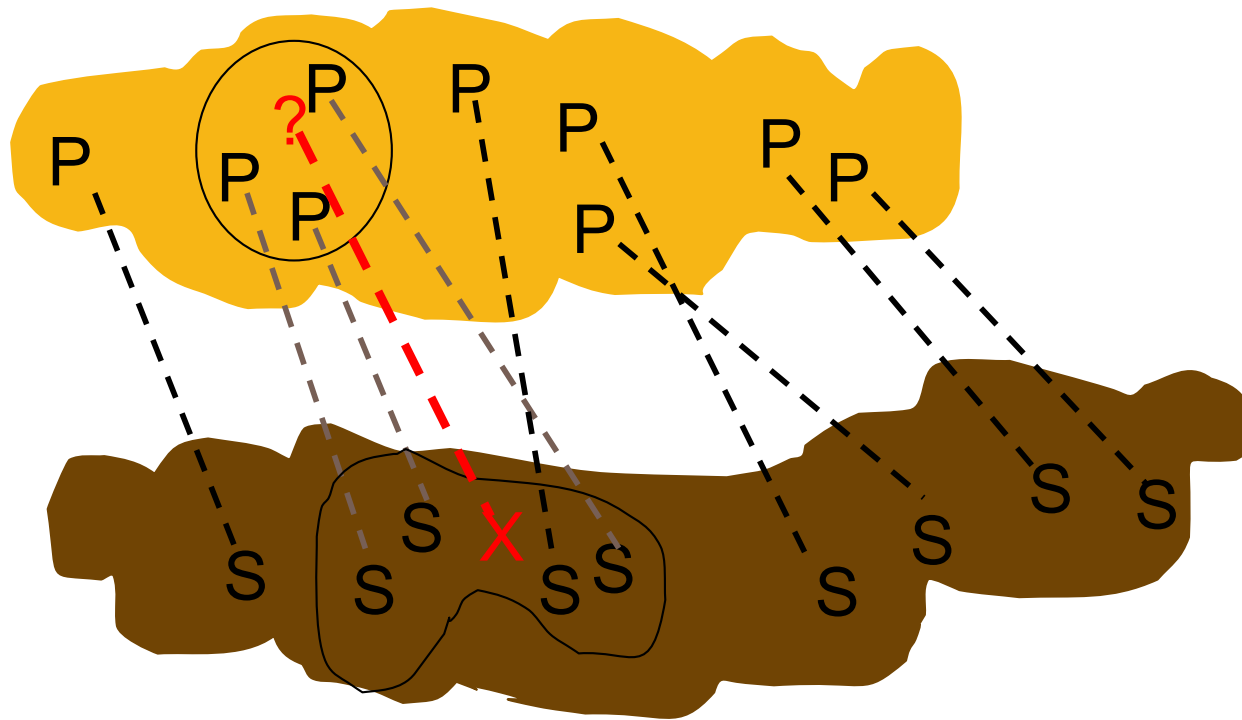


Humans often use available experiences in solving new problems

Problem: $999 \times 9 = ?$
Known case: $1000 \times 9 = 9000$
Modication: $999 \times 9 = 9000 - 9$

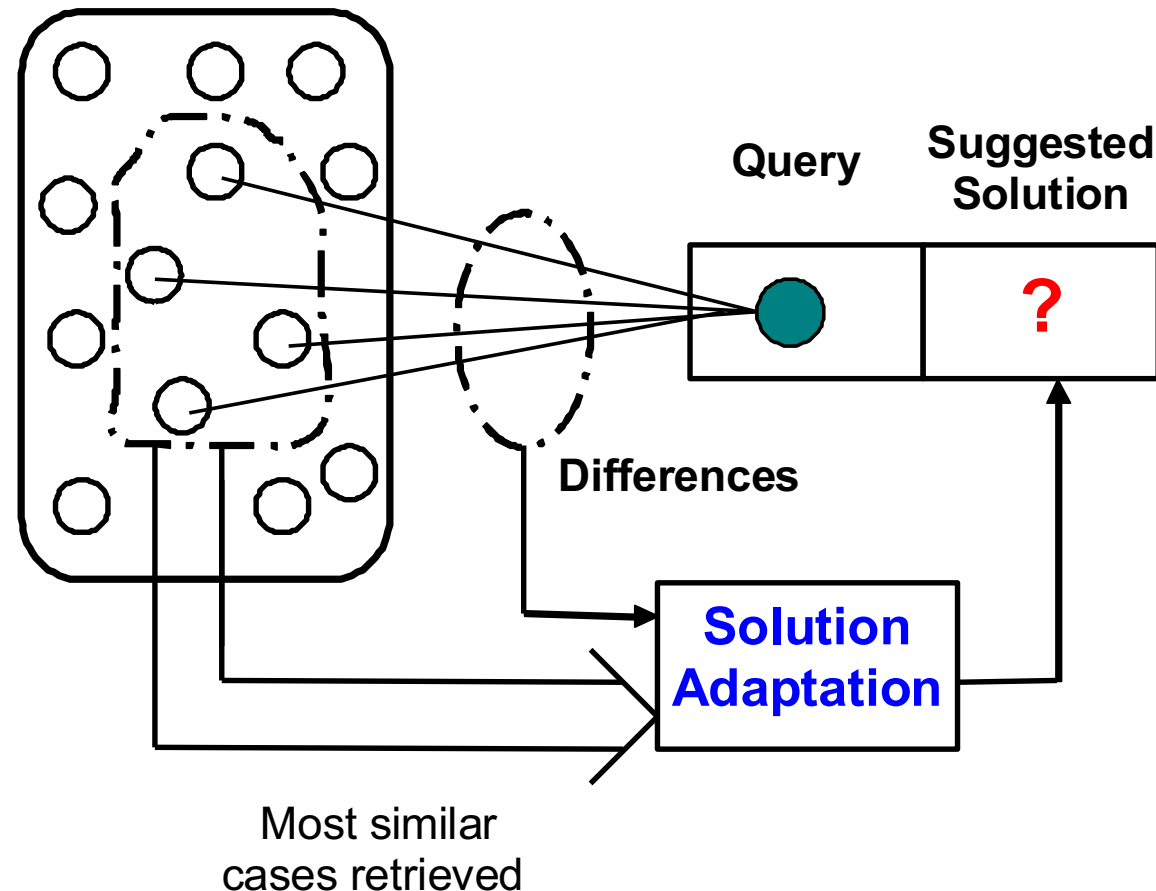
Fundamental Principle of CBR

Similar problems have similar solutions



How is CBR Working

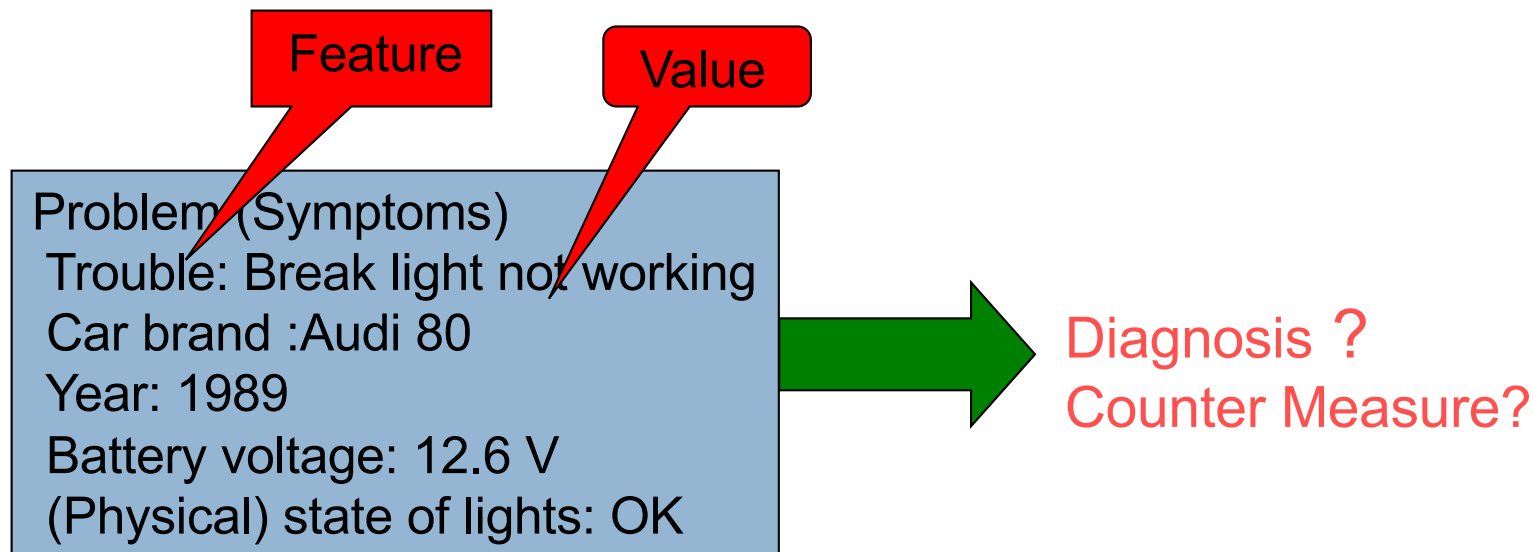
Case Base



- A case in the case base contains a set of values of features and the associated solution
- A problem in query has only feature values

A Car Diagnosis Problem

Current states of the car define a problem in query:



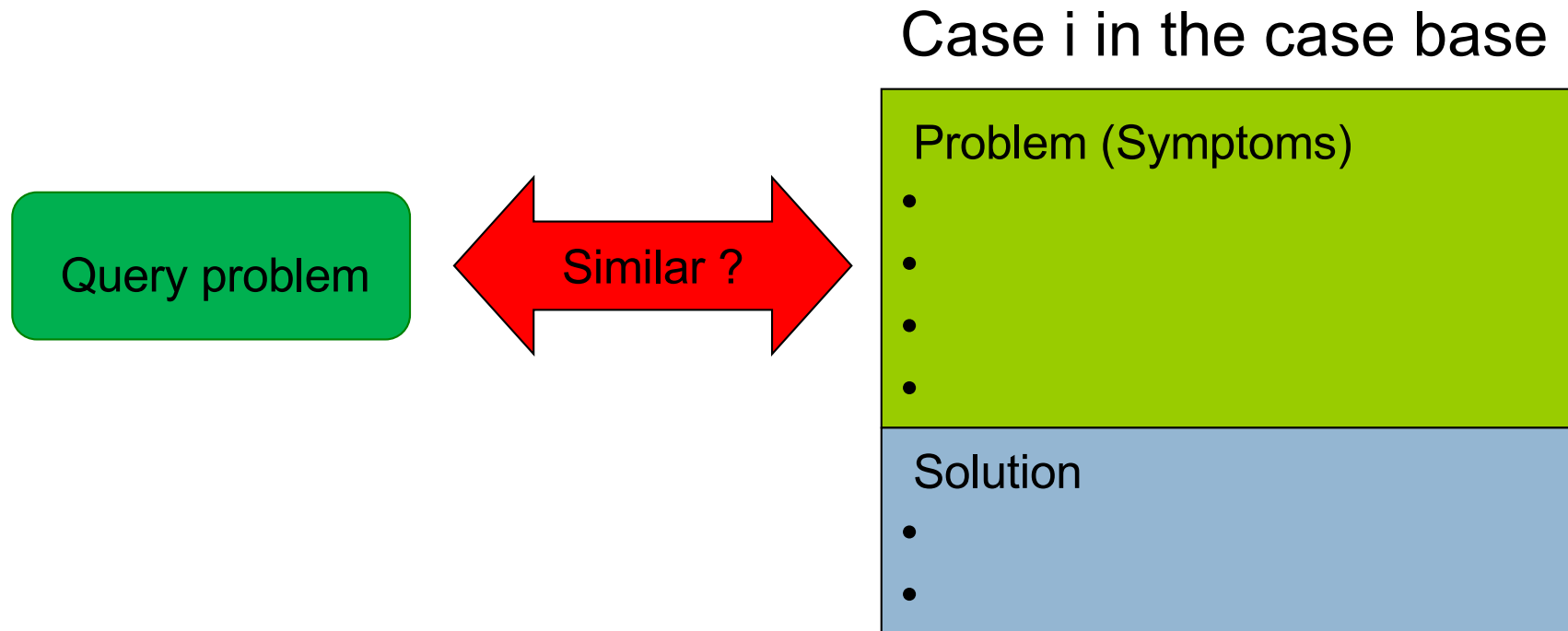
A query problem is a case without solution part

A Case Base with Two Cases

- Each case describes experience in one particular situation
- All cases are independent from each other

C A S E	1	<div>Problem (Symptoms)<ul style="list-style-type: none">• Trouble: Front light not working• Car: VW Golf II, 1.6 L• Year 1993• Battery voltage: 13.6 V• State of lights: OK</div> <div>Solution: Diagnosis: Front light fuse defect Repair: Replace the front light fuse</div>
	2	<div>Problem (Symptoms)<ul style="list-style-type: none">• Trouble: Front light not working• Car: Honda 97• Year 1997• Battery voltage: 15.6 V• State of lights: Surface damaged</div> <div>Solution: Diagnosis: Bulb defect Repair: Replace front light</div>

Compare between Query and Known cases

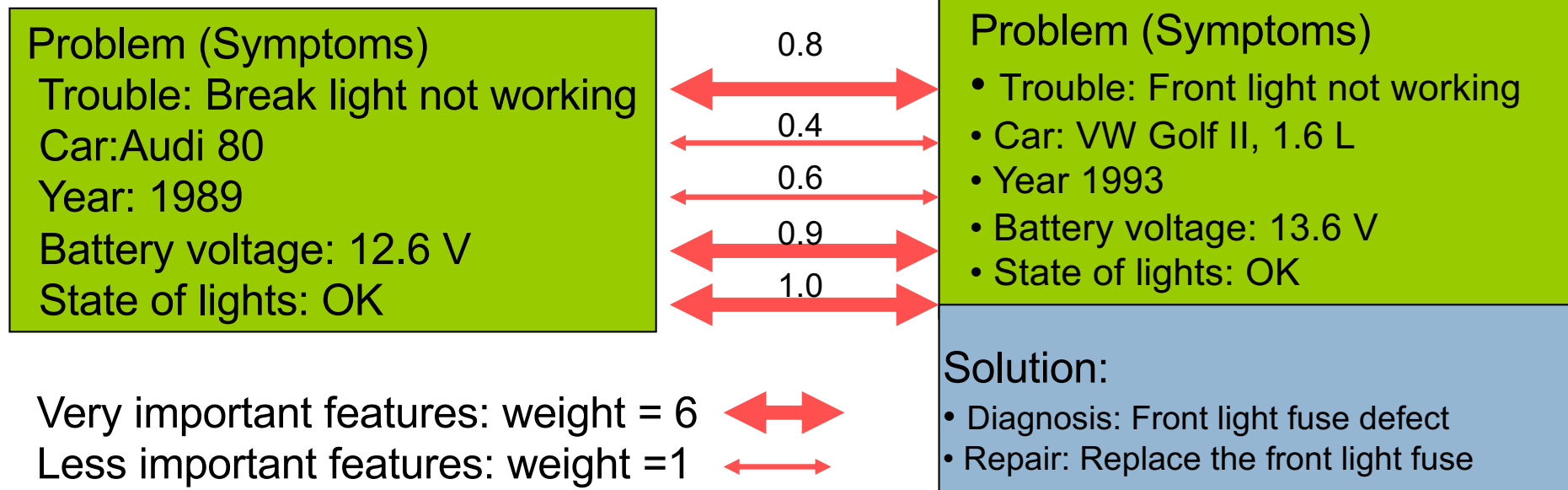


- Similarity is one of the most important concepts in case-based reasoning
- Similarity is assessed based on the difference on each feature
- But different features may have different importance

Similarity Computation

1. Assignment of local similarity degrees on features with a number between 0 and 1 (0: not similar at all; 1: very similar)

Case 1

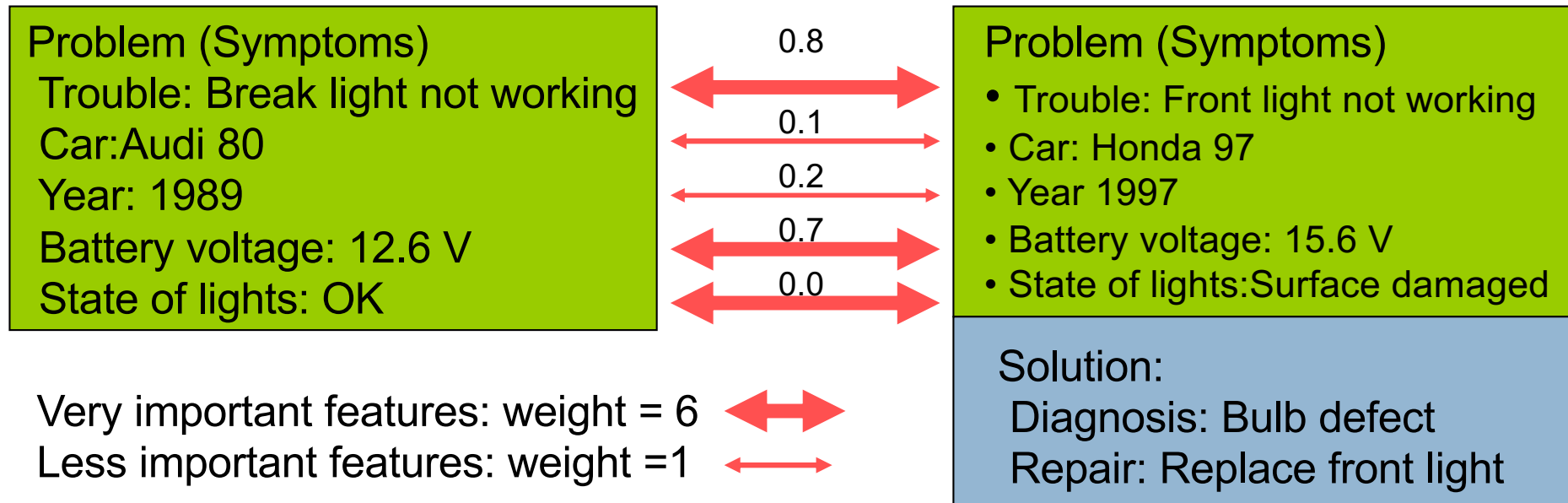


2. Global similarity calculated as a weighted average of local similarity degrees on features:

$$Sim(query, case \ 1) = \frac{6 * 0.8 + 1 * 0.4 + 1 * 0.6 + 6 * 0.9 + 6 * 1.0}{6 + 1 + 1 + 6 + 6} = 0.86$$

Compare Query Problem with Case 2

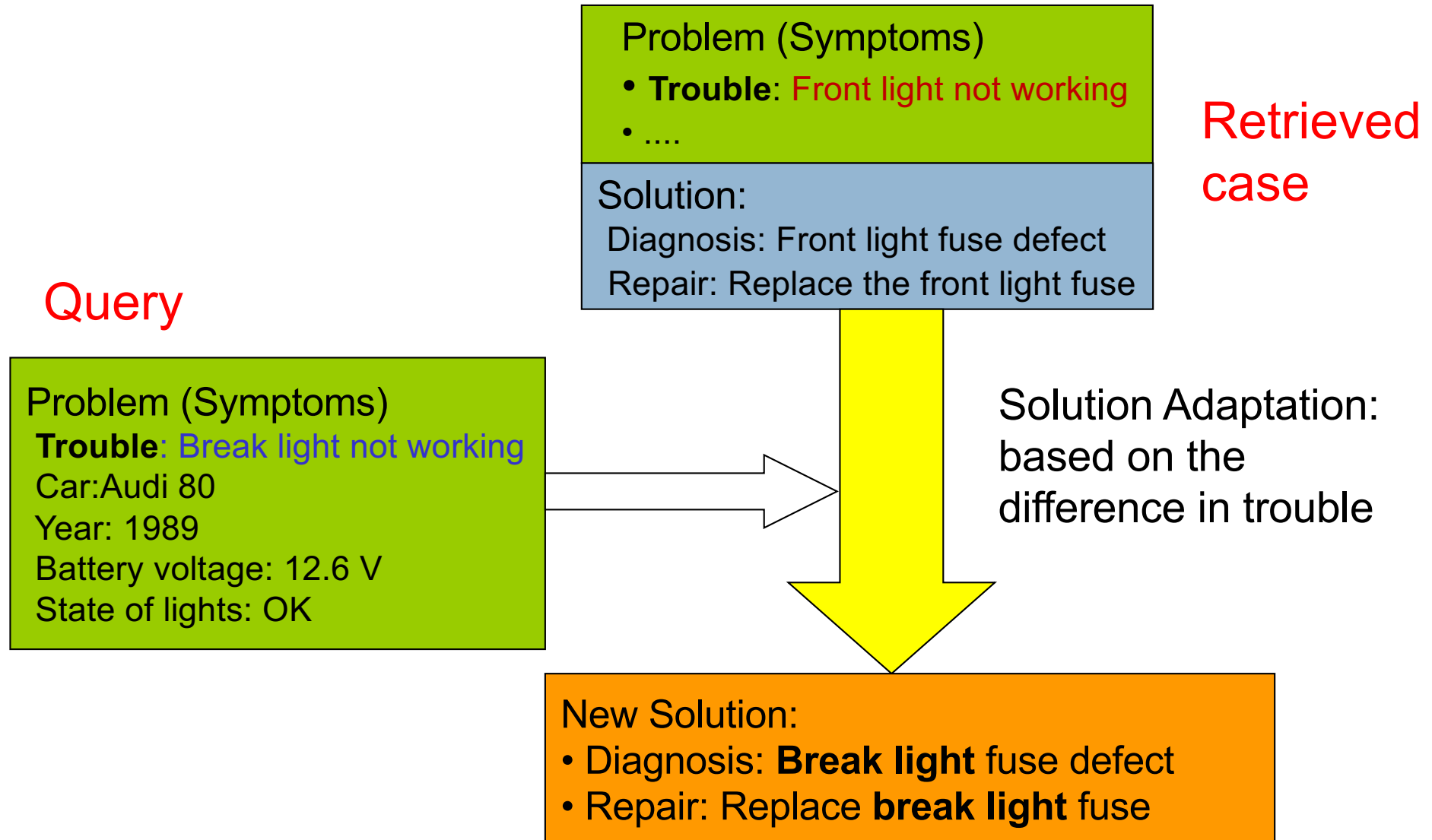
Case 2



The case similarity calculated as a weighted average of local similarity degrees on features:

$$Sim(query, case\ 2) = \frac{6 * 0.8 + 1 * 0.1 + 1 * 0.2 + 6 * 0.7 + 6 * 0}{6 + 1 + 1 + 6 + 6} = 0.465$$

Case 1 Retrived for Adaptation



Store the New Experience

If the solution is correct, store the new case in the case base
(retain of new experience)

Case 3

Problem (Symptoms)

Trouble: Break light not working

Car brand :Audi 80

Year: 1989

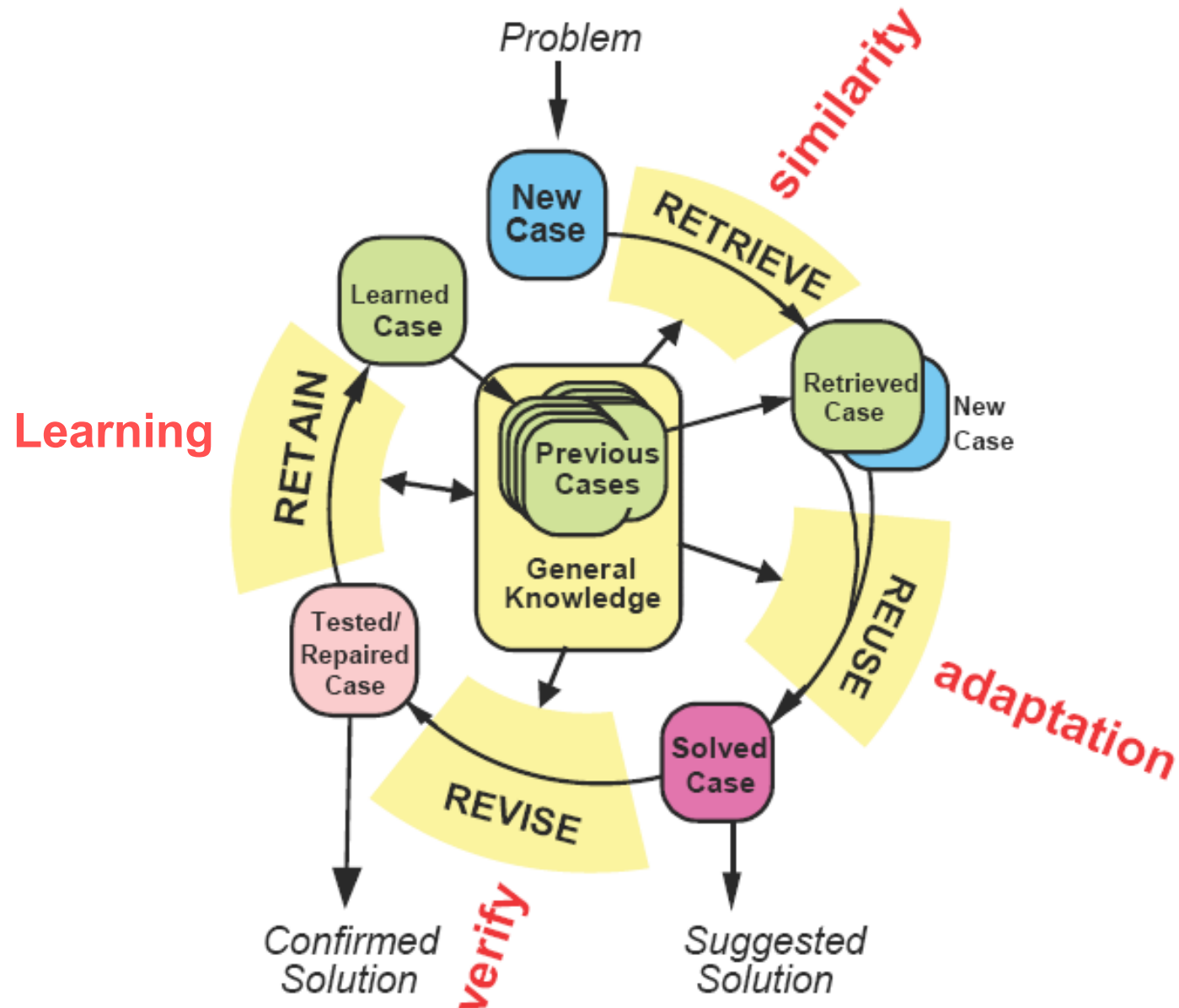
Battery voltage: 12.6 V

(Physical) state of lights: OK

Solution:

- Diagnosis: Break light fuse defect
- Repair: Replace break light fuse

CBR Cycle



Advantages of CBR

Advantages:

- Domains not need to be completely understood
- Ease the knowledge acquisition bottleneck
- Simplify problem solving procedure, i. e., derivation from scratch is avoided
- Enable a sort of incremental learning (case retain)
- Increasing applications of CBR in many areas such as:
fault diagnosis, scheduling, planning, pattern classification

Limitation:

- Difficulty in case adaptation

Practical Techniques in CBR

Similarity Metrics

Similarity metrics play a central role in CBR for retrieval of similar cases given a new problem (query case).

Usually similarity can be assessed by using one of the metrics as follows:

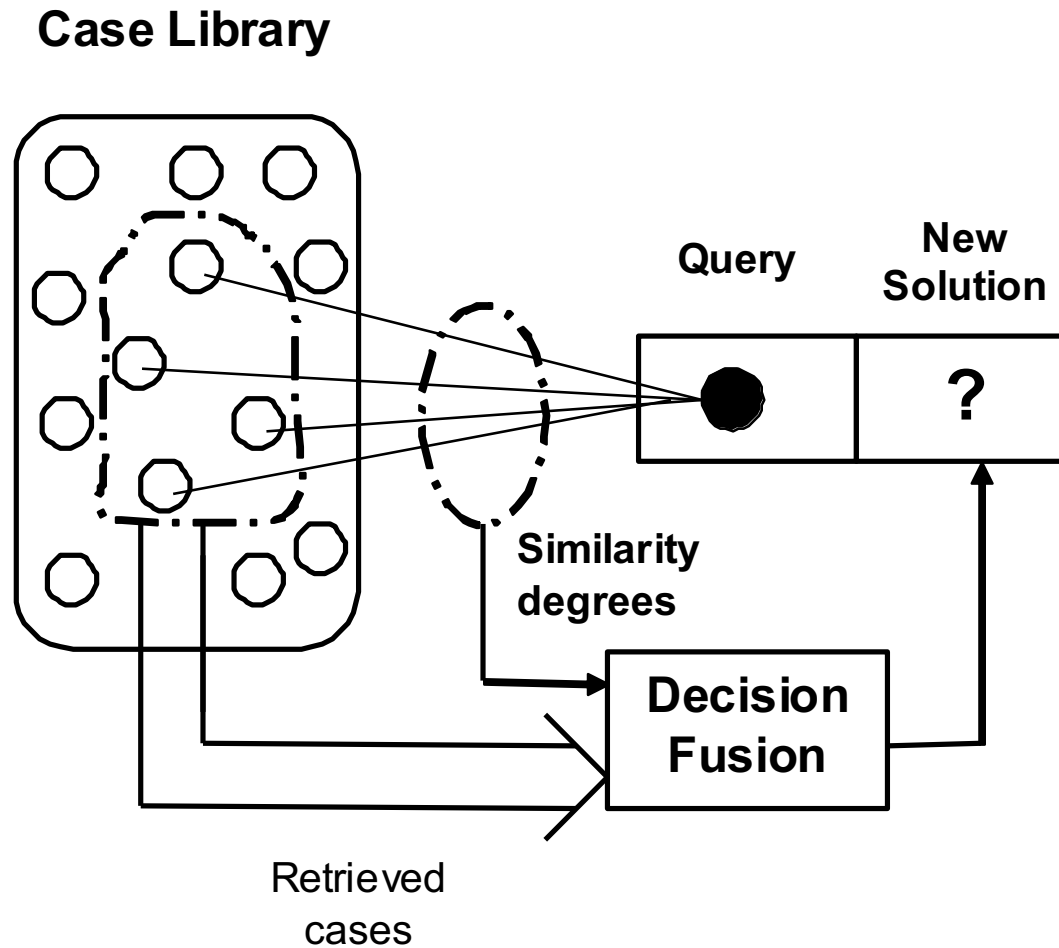
Weighted Euclidean distance: $DIST(X,Y) = \sqrt{\sum_{i=1 \dots n} w_i (x_i - y_i)^2}$

Weighted Manhattan distance: $DIST(X,Y) = \sum_{i=1 \dots n} w_i |x_i - y_i|$

Weighted average on local similarity values on attributes:

Weights reflect the importance of features

Case Reuse by Decision Fusion



Ways of Decision Fusion

Classification: a) majority class

b) Voting strength for class(class with maximum VS)

$$VS(B) = \sum_{C \in Retrieved} \begin{cases} Sim(C, Q), & \text{if } conseq(C) = B \\ 0, & \text{otherwise} \end{cases}$$

Numerical prediction:

$$y = \frac{\sum_{C_i \in Retrieved} y_i \cdot Sim(C_i, Q)}{\sum_{C_i \in Retrieved} Sim(C_i, Q)}$$

Relation between CBR and Fuzzy Systems

CBR and Fuzzy Reasoning

Case-Based Reasoning

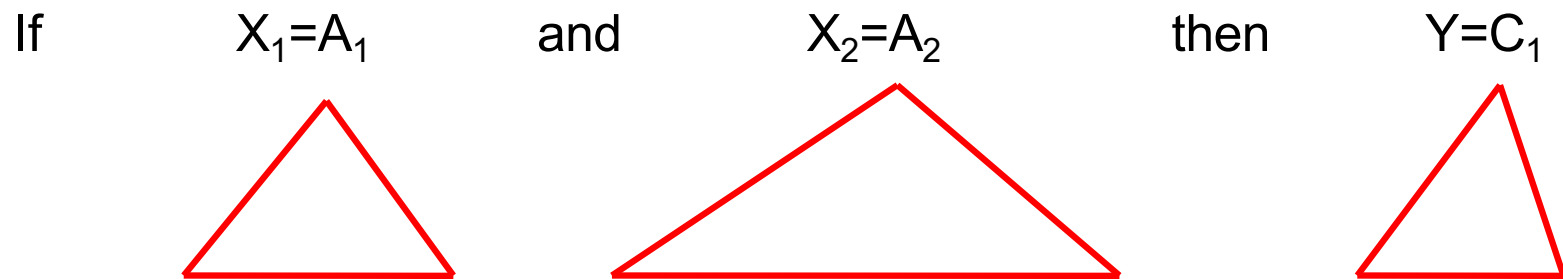
- A memory of known cases
- Matching a query problem to cases; similarity as the measure
- No exact matching is required, but the degrees of similarity are extensively used
- **A case is a specific form of fuzzy rule**; regarded as a fuzzy rule with all membership functions being singletons

Fuzzy Rule-Based Reasoning

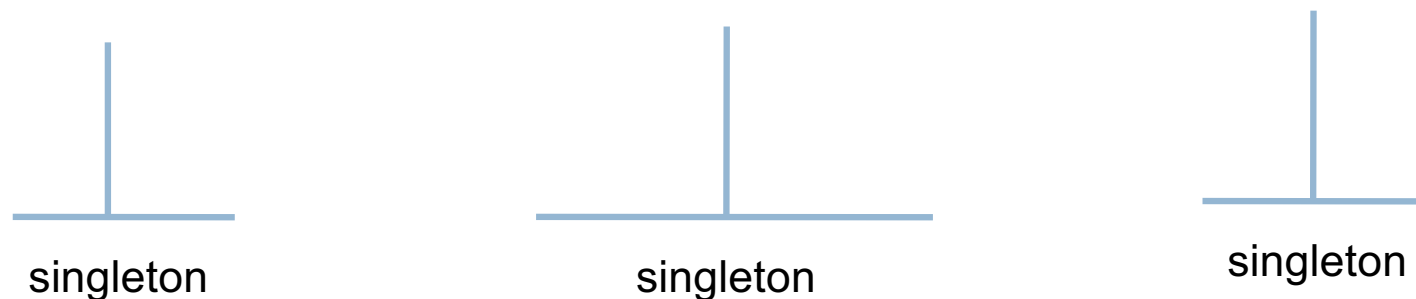
- A set of if-then rules: if A_i and B_i then C_i
- Matching a query problem to rules; firing strength as the measure
- No exact matching is required, but the strengths of firing are extensively used
- **A fuzzy rule can be considered as a generalized case**, covering many concrete cases

Cases and Fuzzy Rules

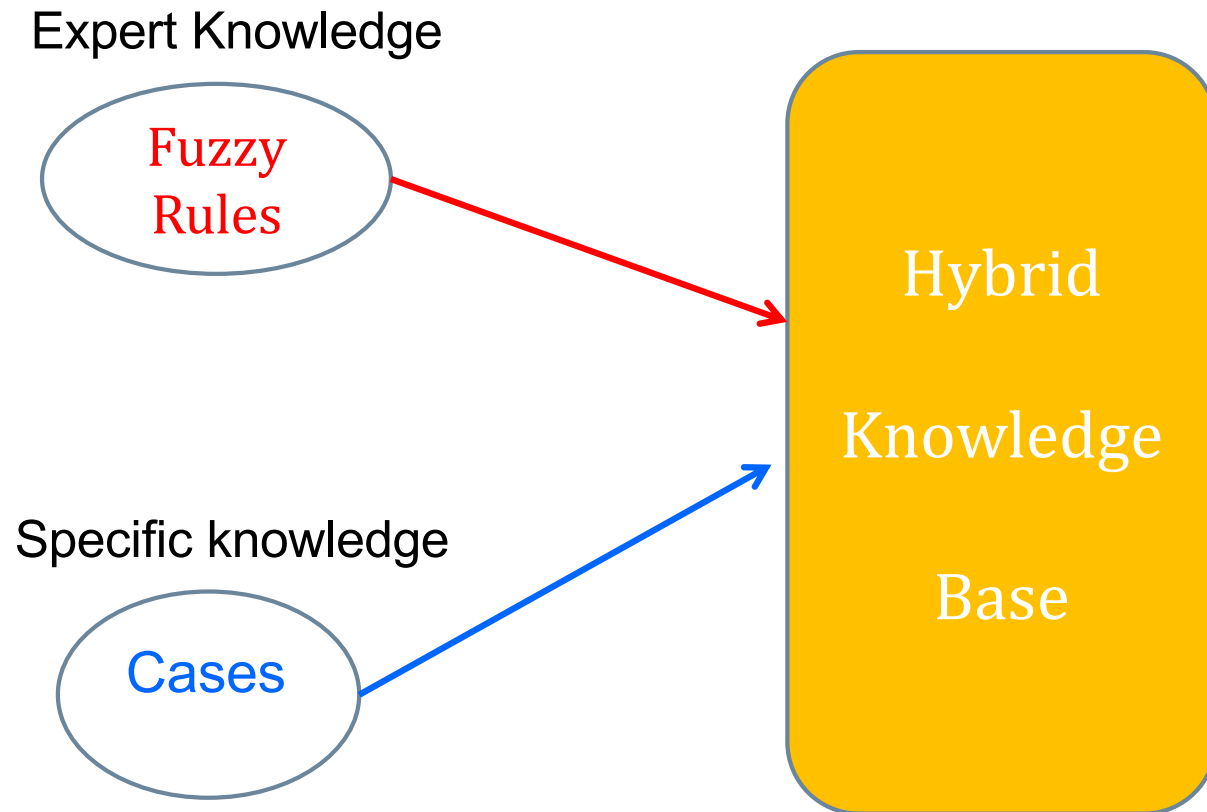
Fuzzy rule as generalized case:



Case as special form of fuzzy rules:



Integrating Cases and Fuzzy Knowledge



Possibility for unified case-based and fuzzy rule based reasoning

CBR in Fuzzy View

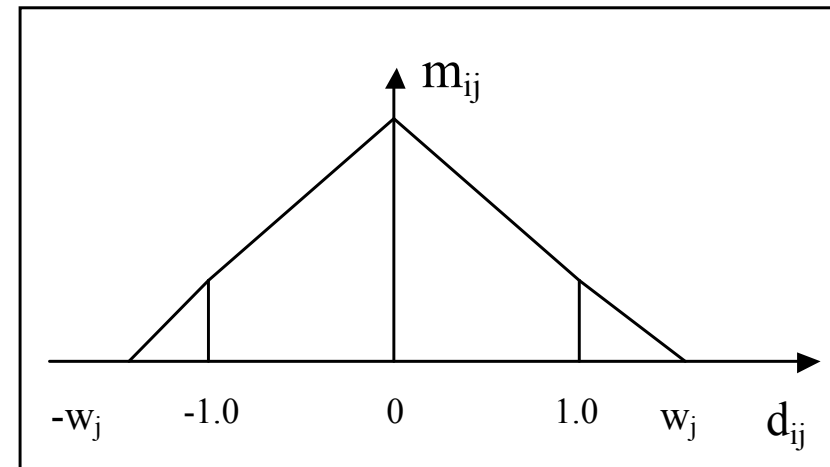
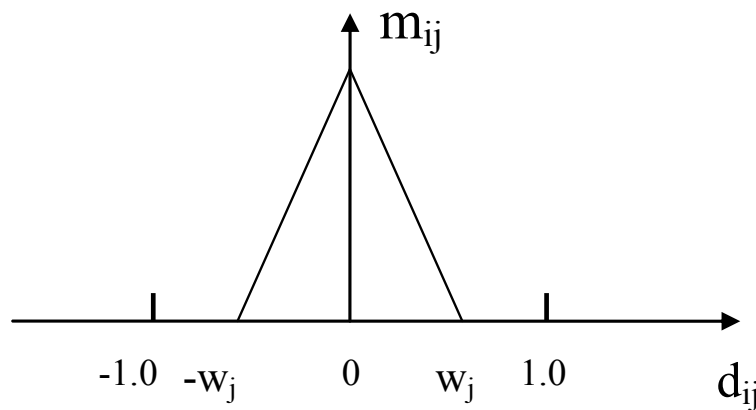
- The fundamental principle of CBR "similar problems have similar solutions" can be formulated as fuzzy rule:

*If problem A is similar to problem B,
then the solution of A is similar to that of B*

- Define the concept "similar" in fuzzy view as a fuzzy subset

Local Similarity

- The local similarity can be considered as a fuzzy subset on the feature difference, whose membership function can be defined as shown in the following figures:



- The parameter w_j reflects the importance of the feature
- The meaning of similarity for different features is different

Global Similarity

- The criterion for similarity between a library case C_i and query Q can be formulated with the following fuzzy rule:

If (feature 1 of C_i is **similar** to that of Q) and and (feature n of C_i is **similar** to that of Q) then C_i is **similar** to Q

Hence similarity degree between C_i and Q is equal to the firing strength of the rule:

$$Sim(C_i, Q) = \min(m_{i1}, m_{i2}, \dots, m_{in})$$

- A new similarity measurement without feature weighting
- More flexible to exclude a remote case in the reasoning procedure