

Idea: experiences themselves are stored. These are called cases.

Given a new example, the most appropriate case(s) in the knowledge base are found and these are used to predict properties of the new example.

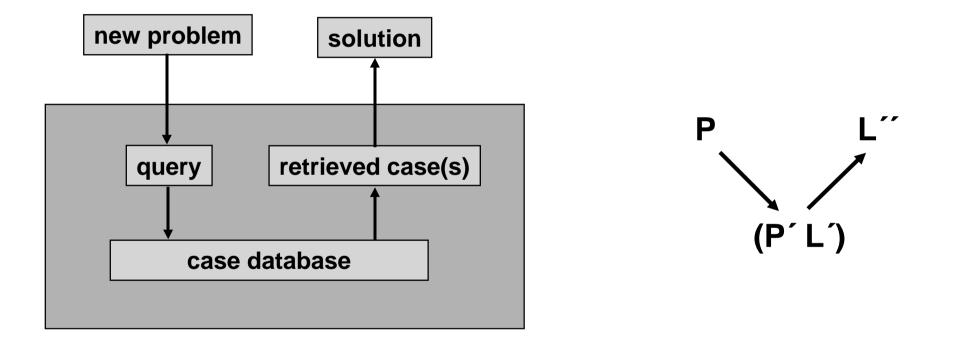


Extremes of Case-based Reasoning

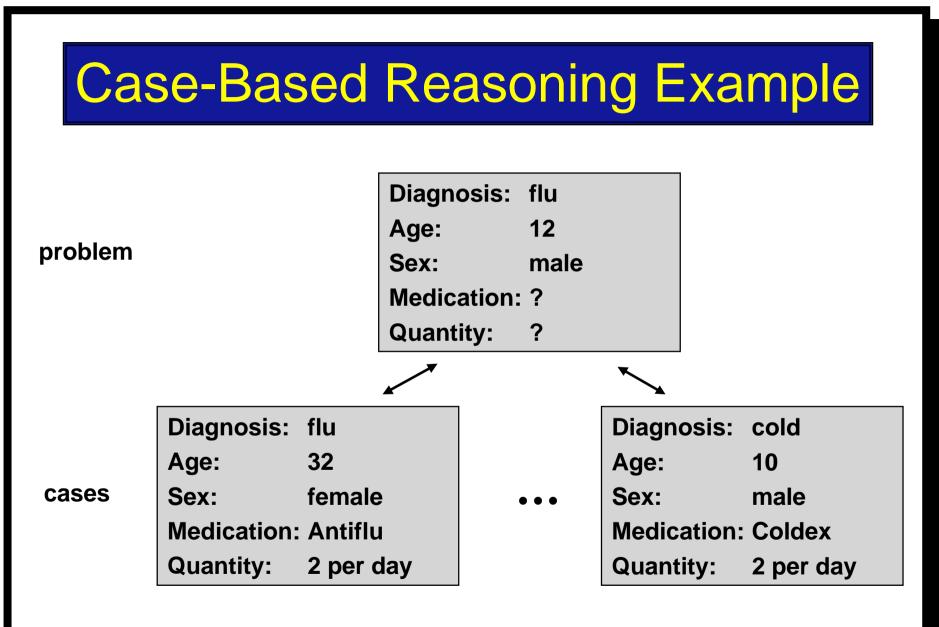
- The cases are simple and for each new example the agent has seen many identical instances. Use the statistics of the cases.
- The cases are simple but there are few exact matches.
 Use a distance metric to find the closest cases.
- The cases are complex, there are no matches. You need sophisticated reasoning to determine why an old case is like the new case.

Examples: legal reasoning, case-based planning.

Principle of Case-based Problem Solving



- store problems together with their solutions in case database
- query database with new problems to find matching cases
- adapt solution of retrieved case to new problem



k-nearest Neighbors

- > Need a distance metric between examples.
- Given a new example, find the k nearest neighbors of that example.
- Predict the classification by using the mode, median, or interpolating between the neighbors.
- Often want k > 1 because there can be errors in the case base.



Euclidean Distance

- Define a metric for each dimension (convert the values to a numerical scale).
- \blacktriangleright The Euclidean distance between examples x and y is:

$$d(x, y) = \sqrt{\sum_{A} w_A (x_A - y_A)^2}$$

x_A is the numerical value of attribute *A* for example *x w_A* is a nonnegative real-valued parameter that specifies the relative weight of attribute *A*.





Like a decision tree, but examples are stored at the leaves.

- The aim is to build a balanced tree; so a particular example can be found in log *n* time when there are *n* examples.
- ▶ Not all leaves will be an exact match for a new example.
- Any exact match can be found in $d = \log n$ time

All examples that miss on just one attribute can be found in $O(d^2)$ time.

