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# 1 Problems of deliberative architectures

- deliberation requires **accurate world models**
- hard to offer real-time guarantees on performance:
- solving any given problem takes longer than an **equivalent reactive implementation**
- solving different problems takes different amounts of time

## 1.1 Deliberation is inherently serial

- generation of each alternative is **inherently serial**, since each step relies on the state **produced by the previous step**
- the number of courses of action grows exponentially with the length of the solution, rather than linearly in the number of percept-action pairs
- the number of alternatives an agent can pursue in parallel is bounded by the number of **processing units available**
- consideration of some alternatives **must be deferred**, i.e., processed serially

## 1.2 Deliberation takes more time

- in a reactive architecture, if we know which action(s) to perform in a given situation we can perform them **immediately**
- all other things being equal, **deliberation will take more steps** than simply reacting, since we have to (serially) generate and compare alternatives
- if the agent doesn't learn, it **must re-solve a problem each time** it encounters it, and will always be slower than a reactive agent
- in a dynamic environment a deliberative agent may take **too long to select an action**, e.g., if the situation changes before deliberation is complete

## 1.3 Deliberation takes unpredictable time

- there is no one best problem-solving method (No Free Lunch Theorem)
- different problems are more or less difficult **for different techniques**
- it is difficult to tell how hard a problem is **just by looking at it**: we have to try and solve it
- we don't know how long it will take to find a solution to a problem **in advance**
- time required by a deliberative agent **will vary** from problem instance to problem instance

### 1.3.1 Approaches to deliberation time

- **ignore it** and hope for the best, e.g., classical planning
- make deliberation run faster, e.g., by using non-optimal algorithms which **sacrifice solution quality for speed**
- try to **predict** how long deliberation will take to solve the current problem, e.g., empirical AI
- **adapt** the amount of deliberation performed **to the time available**, e.g., anytime algorithms
- accept that deliberation may occasionally **be too slow** for some environments or some problems, e.g., monitor the environment and replan if the environment changes

# 2 Accurate world models

Agent must be able to:

- **update its model** of the current environment based on its percepts
- generate the **counterfactual part of the model** by correctly predicting how the world will change, e.g., as a result of its actions

if the world model is incorrect, correct deliberation may **select the wrong course of action**

## 2.1 Representing the real part

it is difficult to create and maintain the veridical model of the current state of the world required for deliberation:

- the information available to the agent may be **incomplete**, e.g., if the environment is only partially observable
- the information available to the agent may be **inaccurate**, e.g., if the environment has changed since the representation was last updated or the agents sensors give incorrect information

## 2.2 Representing the counterfactual part

To generate hypothetical future states which are useful, the agent must be able to accurately predict how the world will change.

- as a result of the **agents actions**: agents actions are often assumed to be infallible;
- as a result of **exogenous changes**: actions of other agents or environmental changes are assumed to be predictable

## 2.3 Approaches to errors in representation

- **ignore them** and hope for the best, e.g., classical planning
- try harder to make the representation **fit reality**, e.g., by using more and better sensors, or by trying to ensure that actions really are infallible, e.g., by implementing them as robust reactive behaviours
- explicitly **represent and reason about uncertainty**, e.g., probabilistic representations, decision theoretic approaches
- **interleave planning and acting**: plan only a little way ahead and keep checking that things are going according to plan; if not, replan.

## 3 Empirical AI

- difficult to tell how hard a problem is for a given problem-solving technique **without trying to solve it**
- empirical AI tries to characterise which problems are hard or easy for a given technique
- most of this work has been in the areas of optimisation, planning and constraint satisfaction problems
- concerned with which problems are soluble at all

## 4 Anytime algorithms

- interruptible algorithm which produces a sequence of solutions of monotonically increasing quality
- anytime algorithms can be used in two ways:
  - to compute the optimum amount of time to spend deliberating to **maximise the utility of the outcome**, e.g., decision theoretic approaches
  - **deliberate for as long as possible** before acting, on the assumption that deliberating for longer will produce a better solution

## 5 Planning in dynamic environments

## Reference section

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