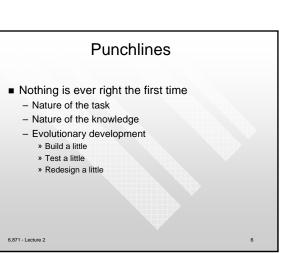
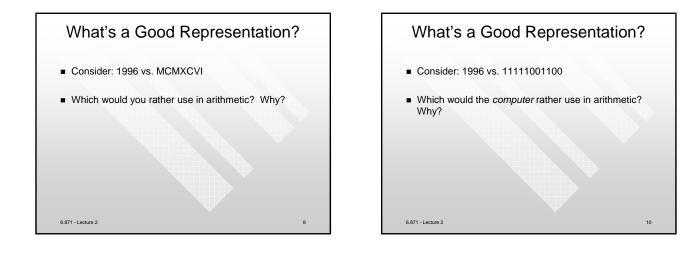




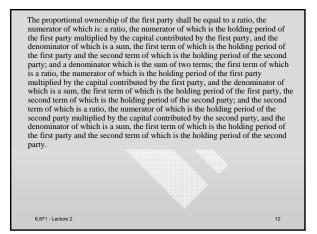
- One payoff: multiple uses of the same knowledge.
- Performance is only the beginning
 - Solving the problem is only (a small) part of the job Explanation
 - Learning
 - Tutoring
- Suppressing detail helps
- Build a custom language

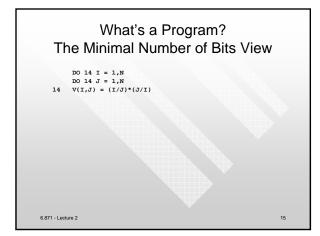
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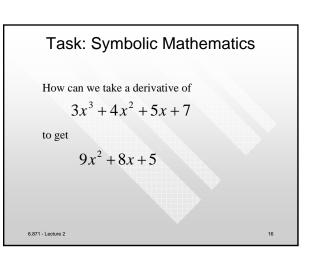








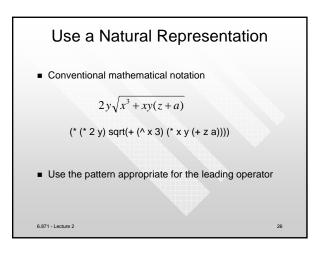




Observations about the knowledge

- It's organized around the operators.
- It's organized around nested sub-expressions
- Top-down tree descent is the natural approach
- The representation should reflect that.
- The representation should facilitate that.

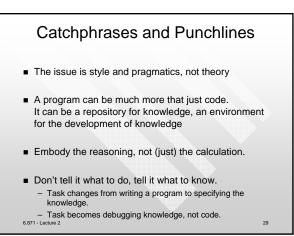
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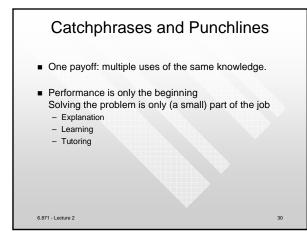


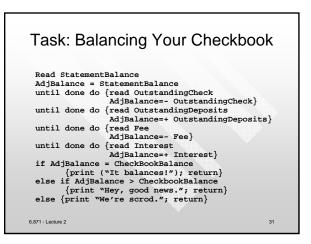
A Small Language In effect we've built a *language* with the right *abstractions:*Expression tree Dispatching on leading operator

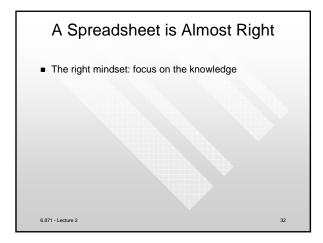
- Recursive descent through the expression tree
- Operators are independent, modular chunks of "mathematical knowledge"
- Operators can be added incrementally
- There is an *indexing mechanism* for finding relevant operators given the structure of the current representational focus

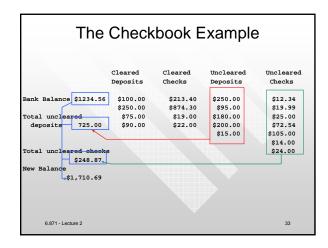
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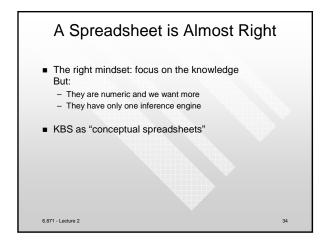


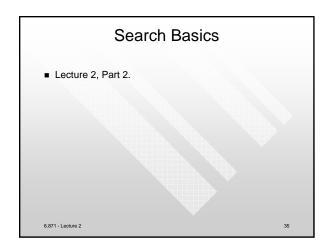


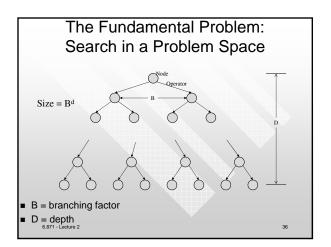


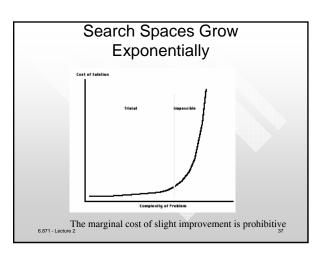


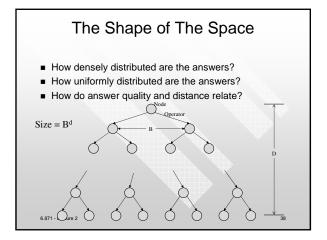


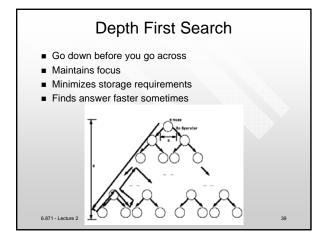


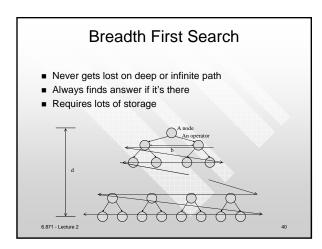


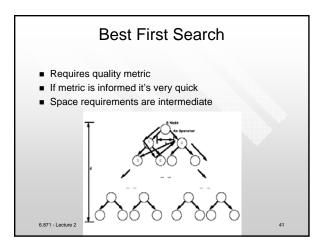


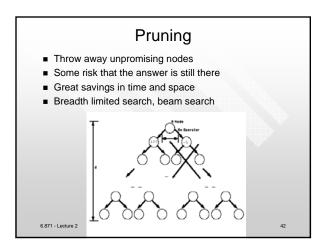


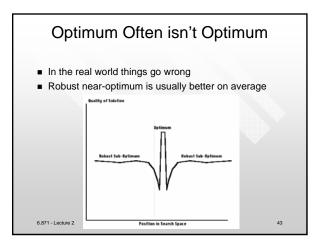


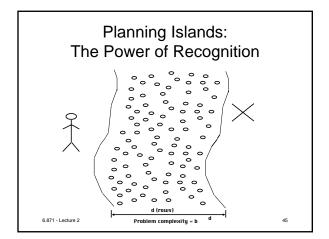


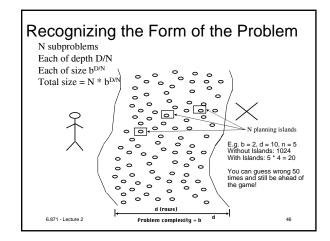












Summary All problem solving problems involve search spaces Search space grow intractably Many common algorithms for search are known In the Knowledge Lies the Power Knowledge of a heuristic metric Knowledge of planning islands Knowledge of relevant abstractions Build representations that capture these sources of power

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