03 Establish datum point at bullseye (0.25, 1.00) 004 B VMC1 0.10 0.34 01 Install 0.15-diameter side-milling tool

> 02 Rough side-mill pocket at (-0.25, 1.25) length 0.40, width 0.30, depth 0.50

03 Finish side-mill pocket at (-0.25, 1.25)

## May All Your Plans Succeed! (or have a high expected utility)

#### Dana S. Nau

# MARYLAND

005 D EC1 30.00 20.00 01 Setup

02 Etching of copper 005 T EC1 90.00 54.77 01 Total time on EC1

 
 006 A
 MC1 30.00
 4.57 01 Setup 02 Prepare board for soldering

 006 B
 MC1 30.00
 0.29 01 Setup

 005 C
 MC1 30.00
 0.29 01 Setup

 006 C
 MC1 30.01 0.29 01 Setup
 0.20 01 Setup

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# Image: Search search?q=plan Image: Search search search?q=plan Image: Search se

#### plan n.

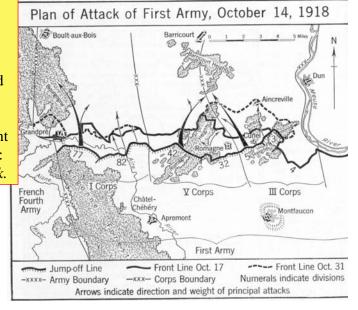
- 1. A scheme, program, or method worked out beforehand for the accomplishment of an objective: *a plan of attack*.
- 2. A proposed or tentative project or course of action: *had no plans for the evening*.
- 3. A systematic arrangement of elements or important parts; a configuration or outline: *a seating plan; the plan of a story*.

- 4. A drawing or diagram made to scale showing the structure or arrangement of something.
- 5. In perspective rendering, one of several imaginary planes perpendicular to the line of vision between the viewer and the object being depicted.
- 6. A program or policy stipulating a service or benefit: *a pension plan*.
- Synonyms: blueprint, design, project, scheme, strategy

Nau: Plans, 2006

#### plan n.

1. A scheme, program, or method worked out beforehand for the accomplishment of an objective: *a plan of attack.* 

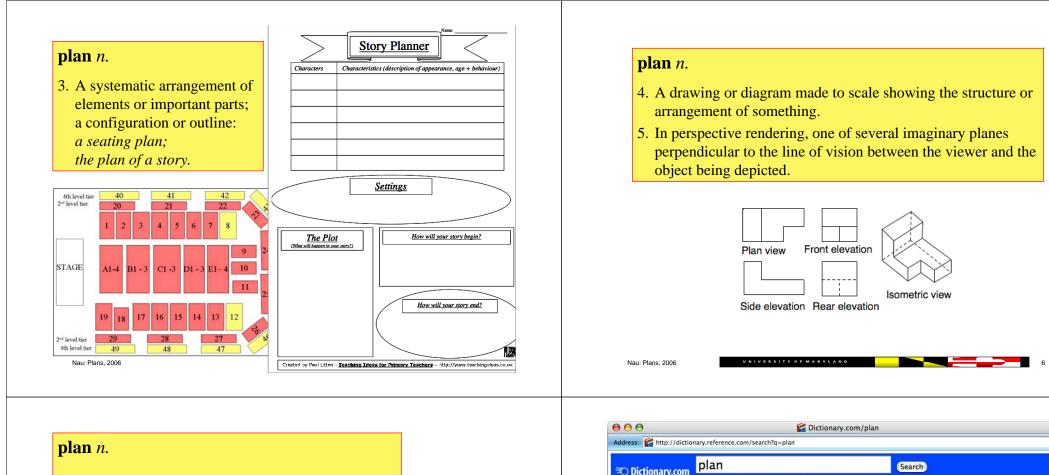


#### plan n.

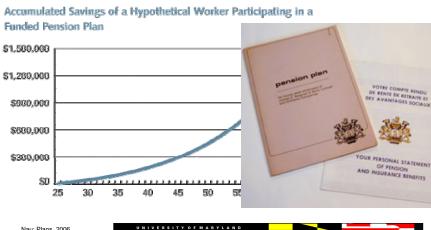
2. A proposed or tentative project or course of action: *had no plans for the evening.* 



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6. A program or policy stipulating a service or benefit: a pension plan.



#### plan n.

Home

1. A scheme, program, or method worked out beforehand for the accomplishment of an objective: a plan of attack.

Dictionary
 Thesaurus
 Web

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- 2. A proposed or tentative project or course of action: had no plans for the evening.
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Synonyms: blueprint, design, project, scheme, strategy

03 Establish datum point at bullseye (0.25, 1.00)		
[a representation] of future anter side-milling too t (-0.25, 1.25) pth 0.50	1	Generating Plans of Action
behavior usually a set of actions, with temporal and other constraints on them, for execution by some agent or agents Austin Tate [MIT Encyclopedia of the Cognitive Sciences, 1999] 02 Dry board in oven at 85 deg. F 005 B ECI 30.00 0.48 01 Setup 02 Spread photoresist from 18000 RPM spinner 005 C ECI 30.00 2.00 01 Setup 02 Photolithography of photoresist	1	<ul> <li>Computer programs to aid human planners</li> <li>Project management (consumer software)</li> <li>Plan storage and retrieval         <ul> <li>e.g., <i>variant process planning</i> in manufacturing</li> <li>Automatic schedule generation             <ul> <li>various OR and AI techniques</li> </ul> </li> <li>For some problems, we would like generate plans (or pieces of plans) automatically</li> </ul> </li> </ul>
using phototool in "real.iges" 005 D EC1 30.00 20.00 01 Setup 02 Etching of copper 005 T EC1 90.00 54.77 01 Total time on EC1 006 A MC1 30.00 4.57 01 Setup 02 Prepare board for soldering 006 B MC1 30.00 0.29 01 Setup 006 C MCI 20.00 2.50 01 Setup	A portion of a manufacturing process plan	<ul> <li>» Much more difficult</li> <li>» Automated-planning research is starting to pay off</li> <li>• Here are some examples</li> </ul>

#### **Space Exploration**

- Autonomous planning, scheduling, control
  - » NASA: JPL and Ames
- Remote Agent Experiment on Deep Space 1
- Mars rovers

QuickTime<sup>™</sup> and a TIFF (Uncompressed) decompressor are needed to see this picture.



#### Games

• Bridge Baron - Great Game Products

LeadLow(P1; S)

StandardFinesseTwo(P2; S)

ð

PlayCard(P<sub>2</sub>; S, R<sub>2</sub>)

North— ♠3

Nau: Plans, 2006

» 2004: 2nd place

PlayCard(P<sub>1</sub>; S, R<sub>1</sub>)

West— ♠2

» 1997 world champion of computer bridge [Smith, Nau, and Throop, AI Magazine, 1998]
Us:East declarer, West dummy

Finesse(P<sub>1</sub>; S)

EasyFinesse(P<sub>2</sub>; S)

 $PlayCard(P_3; S, R_3)$ 

East— **▲**J

(North—  $\blacklozenge$  Q)

#### Outline

- Conceptual model for planning
- Example planning algorithms
- What's bad
- What's good

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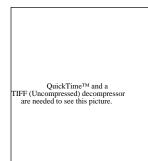
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• Directions and trends

#### **Related Reading**

- My talk today is deliberately non-technical
- For technical details:
  - » Ghallab, Nau, and Traverso Automated Planning: Theory and Practice Morgan Kaufmann, May 2004
  - » First comprehensive textbook and reference work on automated planning
  - » For further information
    - http://www.laas.fr/planning

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Opponents: defenders, South & North

East: KJ74

West: A2

BustedFinesse(P<sub>2</sub>; S)

(North—  $\bigstar$ 3)

FinesseFour(P<sub>4</sub>; S)

PlayCard(P<sub>4</sub>; S, R<sub>4</sub>')

South— ♠O

Contract:East - 3NT

FinesseTwo(P<sub>2</sub>; S)

StandardFinesseThree(P<sub>3</sub>; S)

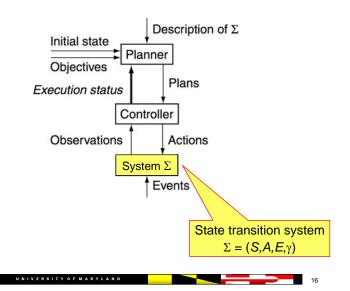
StandardFinesse(P<sub>2</sub>; S)

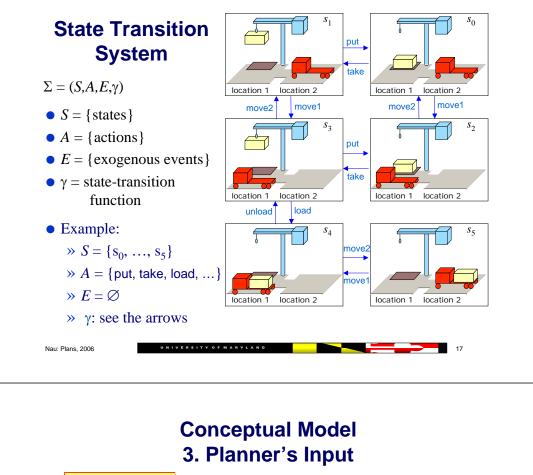
 $PlayCard(P_4; S, R_4)$ 

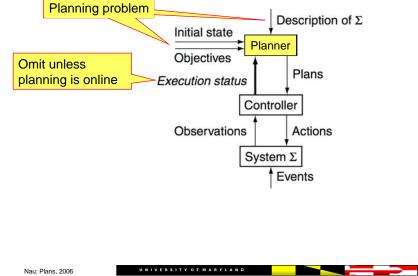
South— ♠5

On lead:West at trick 3

#### Conceptual Model 1. Environment





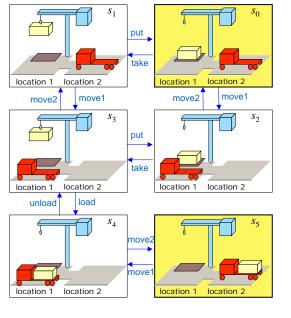


## Planning Problem

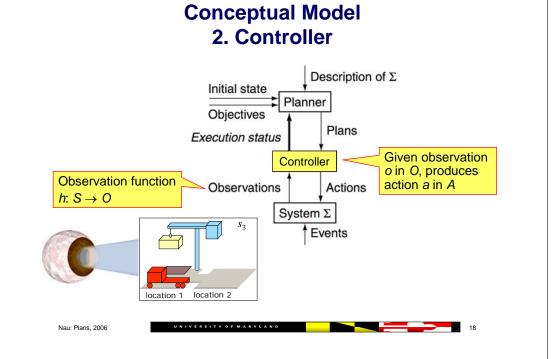
- Description of Σ
- Initial state or set of states

#### $\gg$ Initial state = $s_0$

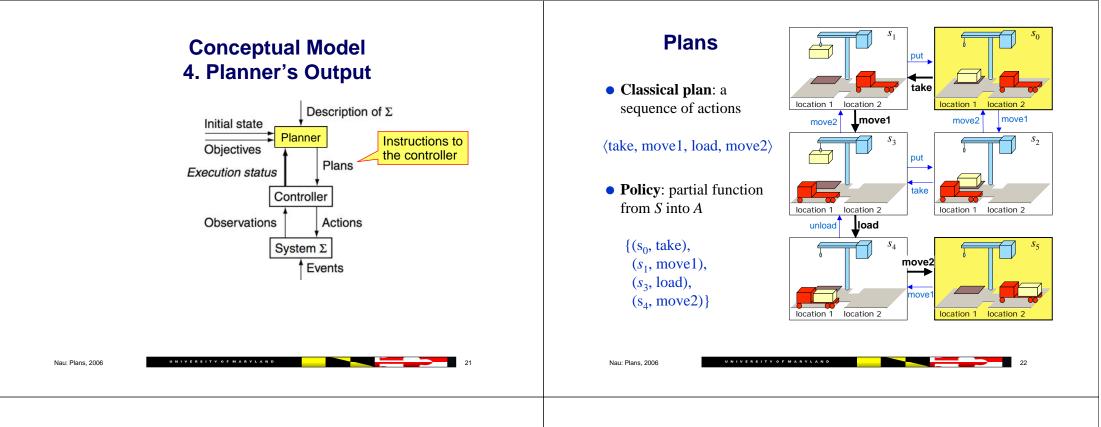
- Objective
  - » Goal state, set of goal states, set of tasks,
     "trajectory" of states, objective function, ...
  - $\gg$  Goal state =  $s_5$



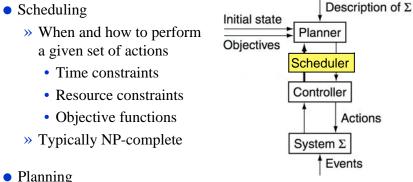
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19



#### **Planning Versus Scheduling**



- Planning
  - » Decide what actions to use to achieve some set of objectives
  - » Can be much worse than NP-complete; worst case is undecidable

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#### **Three Main Types of Planners**

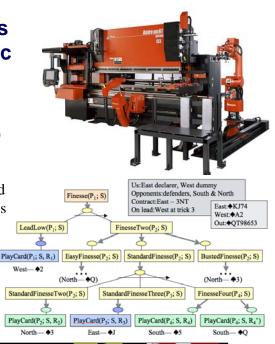
- 1. Domain-specific
- 2. Domain-independent
- 3. Configurable
- I'll briefly discuss each

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## Types of Planners 1. Domain-Specific

- Made or tuned for a specific domain
- Won't work well (if at all) in any other domain
- Most successful real-world planning systems work this way



#### Nau: Plans, 2006

Planner

Controller

System  $\Sigma$ 

**Events** 

Initial state

Objectives

Execution status

Observations

Description of  $\Sigma$ 

Plans

Actions

#### • A0: Finite system:

- » finitely many states, actions, events
- A1: Fully observable:
  - » the controller always  $\Sigma$ 's current state
- A2: Deterministic:
  - » each action has only one outcome
- A3: Static (no exogenous events):
  - » no changes but the controller's actions
- A4: Attainment goals:
  - $\gg$  a set of goal states  $S_g$
- A5: Sequential plans:
  - » a plan is a linearly ordered sequence of actions  $(a_1, a_2, \dots a_n)$
- A6: Implicit time:
  - » no time durations; linear sequence of instantaneous states

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**Restrictive Assumptions** 

- A7: Off-line planning:
  - » planner doesn't know the execution status

#### Nau: Plans, 2006

#### Types of Planners 2. Domain-Independent

- In principle:
  - » Works in any planning domain
  - » No domain-specific knowledge except the definitions of the basic actions
- In practice:

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- » Not feasible to develop domainindependent planners that work in *every* possible domain
- » Restrictive assumptions to simplify the set of domains
  - Classical planning
  - Historical focus of most research on automated planning

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## **Classical Planning**

- Classical planning requires all eight restrictive assumptions
  - » Offline generation of action sequences for a deterministic, static, finite system, with complete knowledge, attainment goals, and implicit time
- Reduces to the following problem:
  - » Given  $(\Sigma, s_0, S_g)$
  - » Find a sequence of actions (a<sub>1</sub>, a<sub>2</sub>, ..., a<sub>n</sub>) that produces a sequence of state transitions (s<sub>1</sub>, s<sub>2</sub>, ..., s<sub>n</sub>) such that s<sub>n</sub> is in S<sub>g</sub>.
- This is just path-searching in a graph
  - » Nodes = states
  - » Edges = actions
- Is this trivial?

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#### **Classical Planning**

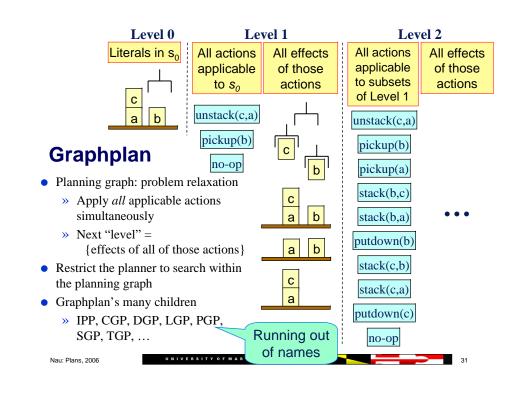
- Generalize the earlier example:
  - » Five locations, three robot carts, 100 containers, three piles
    - Then there are 10<sup>277</sup> states
- Number of particles in the universe is only about 10<sup>87</sup>

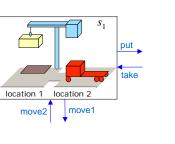
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- » The example is more than  $10^{190}$  times as large!
- Automated-planning research has been heavily dominated by classical planning
  - » Dozens (hundreds?) of different algorithms

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» I'll briefly mention a few of the best-known ones





#### С **Partial-Order Planning** а b • Decompose sets of goals into the Start individual goals dear(x), with x = a• Plan for them separately unstack(x,a)» Bookkeeping info to detect clear(a) and resolve interactions clear(b). putdown(x)handempty handempty pickup(b) pickup(a)

holding(a)

stack(a,b)

on(a,b)

а

b

с

Goal:

on(a,b) & on(b,c)

• For classical planning, stack(b,c) clear(b)

holding(a)

• Remote agent experiment and Mars rovers used temporal-planning extensions of it

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#### **Heuristic Search**

- Do an A\*-style heuristic search guided by a *heuristic function* that estimates the distance to a goal
  - » Can use planning graphs to compute the heuristic function
- Problem: A\* quickly runs out of memory
  - » So do a greedy search
- Greedy search can get trapped in local minima
   » Greedy search plus local search at local minima
- HSP [Bonet & Geffner]
- FastForward [Hoffmann]

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Nau: Plans, 2006

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#### **Translation to Other Domains**

- Translate the planning problem or the planning graph into another kind of problem for which there are efficient solvers
  - » Find a solution to that problem
  - » Translate the solution back into a plan
- Satisfiability solvers, especially those that use local search
  - » Satplan and Blackbox [Kautz & Selman]
- Integer programming solvers such as Cplex

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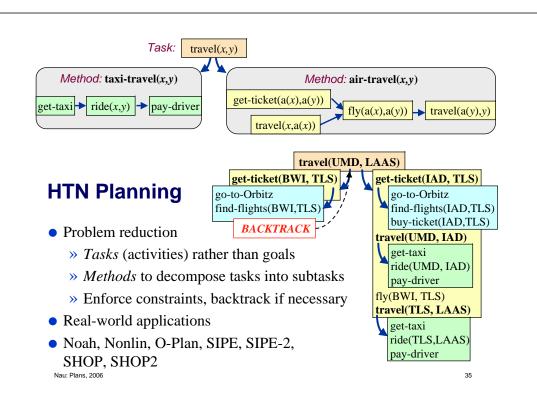
» [Vossen et al.]

Nau: Plans, 2006

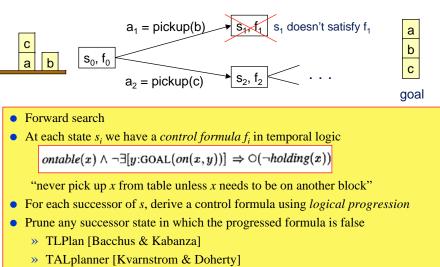
#### Types of Planners: 3. Configurable

- Domain-independent planners are quite slow compared with domain-specific planners
  - » Blocks world in linear time [Slaney and Thiébaux, A.I., 2001]
  - » Can get analogous results in many other domains
- But we don't want to write a whole new planner for every domain!
- Configurable planners
  - » Domain-independent planning engine
  - » Input includes info about how to solve problems in the domain
    - Hierarchical Task Network (HTN) planning
    - Planning with control formulas

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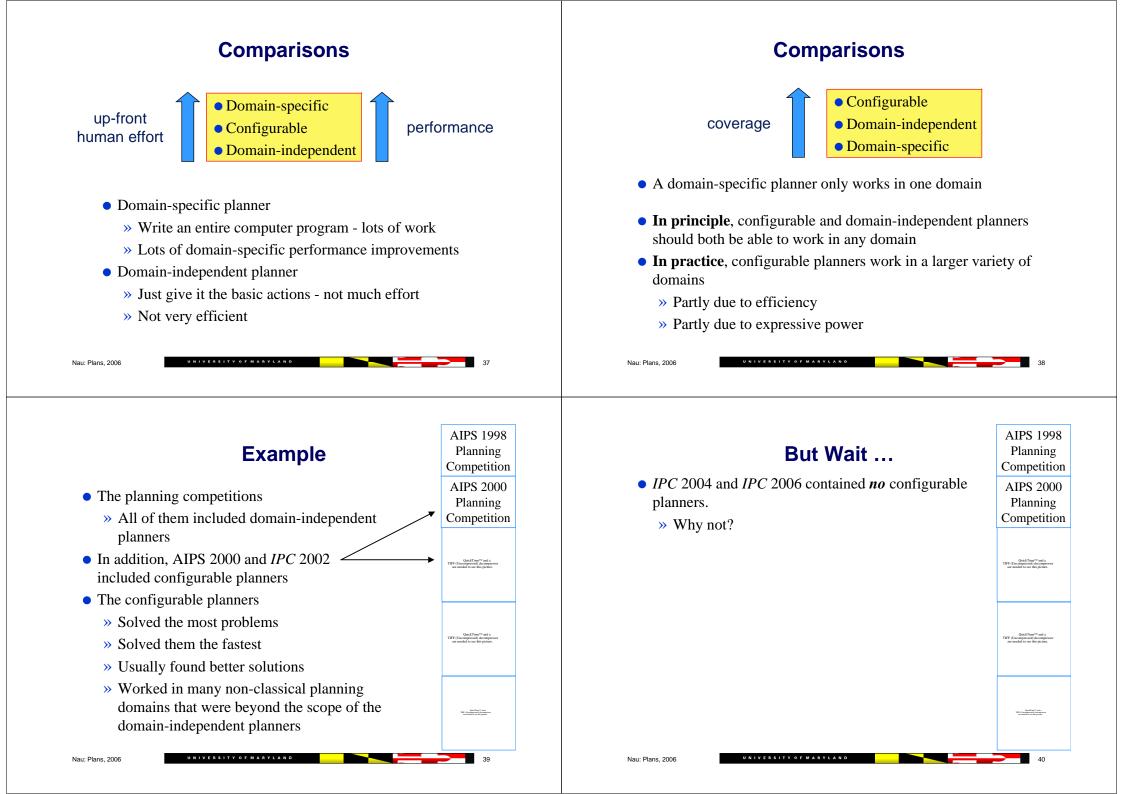


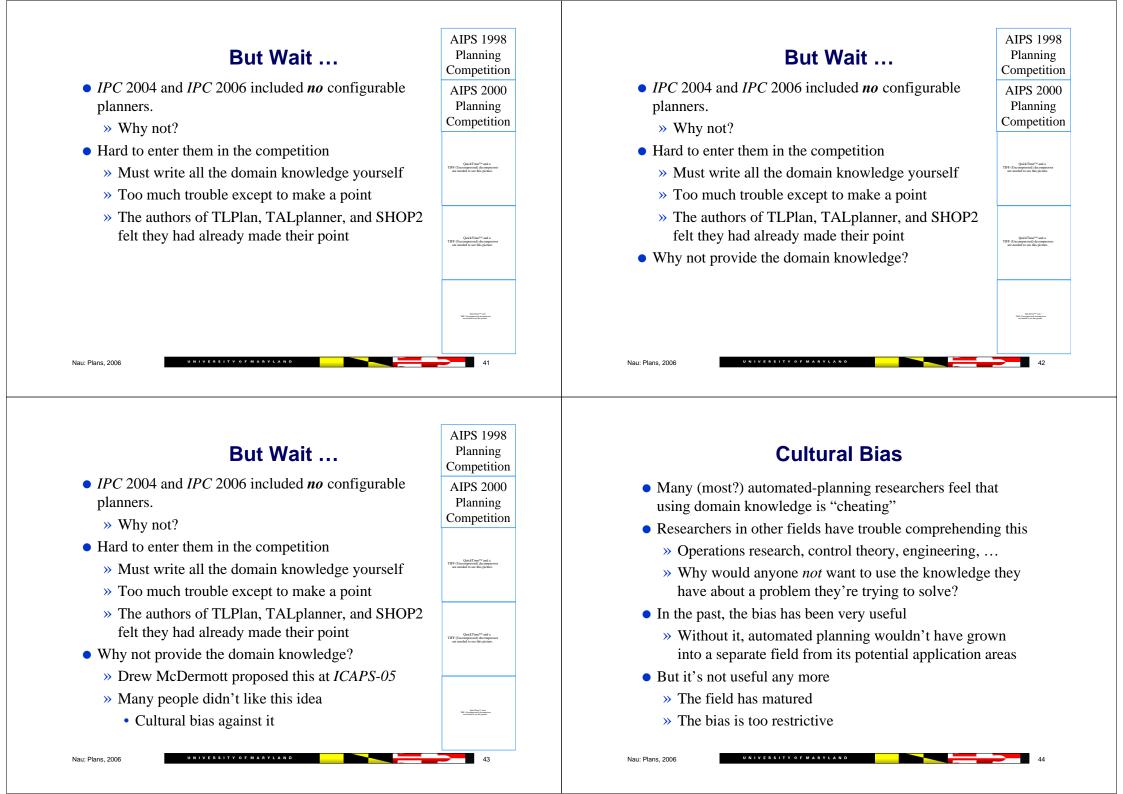
## **Planning with Control Formulas**



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Nau: Plans, 2006

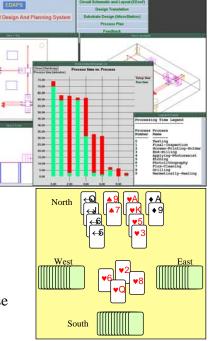




## Example

- Typical characteristics of application domains
  - » Dynamic world
  - » Multiple agents
  - » Imperfect/uncertain info
  - » External info sources
  - users, sensors, databases
  - » Durations, time constraints, asynchronous actions
  - » Numeric computations
    - geometry, probability, etc.
- Classical planning excludes all of these

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#### In Other Words ...

- We **like** to think classical planning is domain-independent planning
- But it isn't!
  - » Classical planning only includes domains that satisfy some very specific restrictions
  - » Classical planners depend heavily on those restrictions
- This is fine for the **blocks world**
- Not so fine for the real world

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QuickTime<sup>™</sup> and a TIFF (LZW) decompressor are needed to see this picture.

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## Good News, Part 1

- We're already moving away from classical planning
- Example: the planning competitions
  - » AIPS 1998, AIPS 2000, IPC 2002, IPC 2004
- Increasing divergence from classical planning
  - » 1998, 2000: classical planning
  - » 2002: added elementary notions of time durations, resources
  - » 2004: added inference rules, derived effects, and a separate track for planning under uncertainty
  - » 2006: added soft goals, trajectory constraints, preferences, plan metrics

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AIPS 1998 Planning Competition AIPS 2000 Planning Competition

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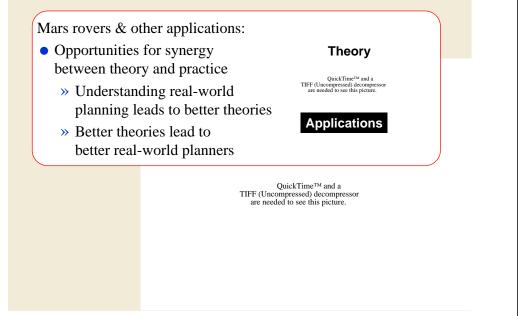
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#### **Good News, Part 2**

- Success in high-profile applications
  - » A success like the Mars rovers is a big deal
  - » Creates excitement about building planners that work in the real world

QuickTime™ and a TIFF (Uncompressed) decompressor are needed to see this picture.

## Good News, Part 3



#### Good News, Part 4

- Classical planning research has produced some very powerful techniques for reducing the size of the search space
- We can generalize these techniques to work in non-classical domains
- Examples:

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- 1. Partial order planning
  - Extended to do temporal planning

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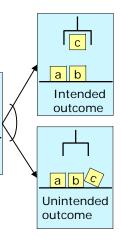
- > RAX (Deep Space 1)
- > Mars rovers
- 2. HTN planning
  - Lots of applications
- 3. Planning under uncertainty ...

**Digression:** What planning under uncertainty is

- Actions with several possible outcomes
  - » Action failures gripper drops its load
  - » Exogenous events road closed
- Two primary models
  - » Markov Decision Processes (MDPs)
    - Probabilities, costs, rewards, optimize expected utility
    - Dynamic programming
  - » Nondeterministic planning domains
    - No numbers
    - Solutions: weak, strong, strong-cyclic, others

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• Symbolic model checking



a b

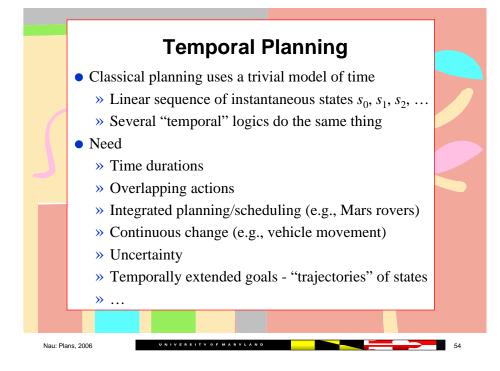
Grasp box c

## Good News, Part 4 (continued)

- 3. General way to nondeterminize forward-chaining planners
  - » Rewrite them to work in nondeterministic domains
    - TLPlan  $\rightarrow$  ND-TLPlan
    - TALplanner  $\rightarrow$  ND-TALplanner
    - SHOP2  $\rightarrow$  ND-SHOP2
  - » Big (exponential) speedups compared to previous planners for nondeterministic domains [Kuter and Nau, *AAAI*-04]
  - » Even bigger speedups if we use the BDD representation used in the previous planners for nondeterministic domains
    - [Kuter, Nau, Pistore, and Traverso, ICAPS-05]
- Analogous results for MDPs [Kuter and Nau, AAAI-05]

Nau: Plans, 2006





#### **Planning in Dynamic Environments**

- Automated planning research
  - » Information is static; planner starts with all of it
- Real-world planning

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- » Acquire information during planning and execution
  - Applications: web services, many others
- » What info to look for? Where to get it?

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- » How to deal with lag time and information volatility?
  - During execution
  - and during planning [Au et al., ECAI-04, 06]
- Candidate for a new IPC track?

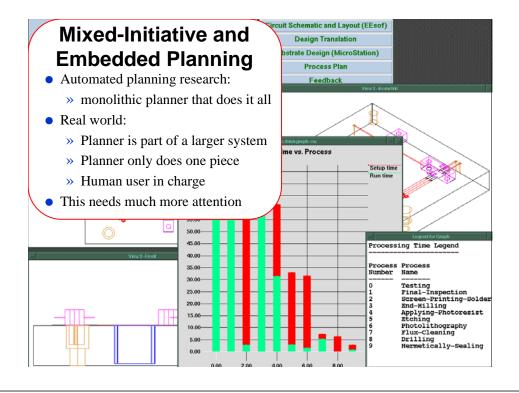
## Acquiring Domain Knowledge

- How to get the domain knowledge needed to plan efficiently?
  - » One of the most neglected topics for planning research, but one of the most important
  - » If we could do this well on real-world problems, planners would be hundreds of times more useful
- Researchers are starting to realize this
  - » At ICAPS-05 there was an informal "Knowledge Engineering Competition"

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- GUIs for creating knowledge bases for planning
- Ways for planners to learn domain knowledge
- Overlap with HCI, ML, and CBR

006





#### **Overlap with Other Fields**

- Various kinds of planning are studied in many different fields
  - » AI planning, computer games, game theory, OR, economics, psychology, sociology, political science, industrial engineering, systems science, control theory
- The research groups are often nearly disjoint
  - » Different terminology, assumptions, ideas of what's important
  - » Hard to tell what the similarities and differences are
- Potential for cross-pollination

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» Combine ideas and approaches from different fields

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#### **Example: Planning Under Uncertainty**

- AI planning, OR, control theory all use MDP models
  - » OR & control theory
    - Infinitely many states, continuous sets
    - · Actions, costs, rewards: differentiable functions
    - Linear and nonlinear optimization
  - » Automated planning
    - Finitely many states
    - No good continuous approximations
    - Discrete optimization
- Many important problems are hybrids of both
  - » Combine and extend the techniques

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