

规划问题求解 (AI Planning)

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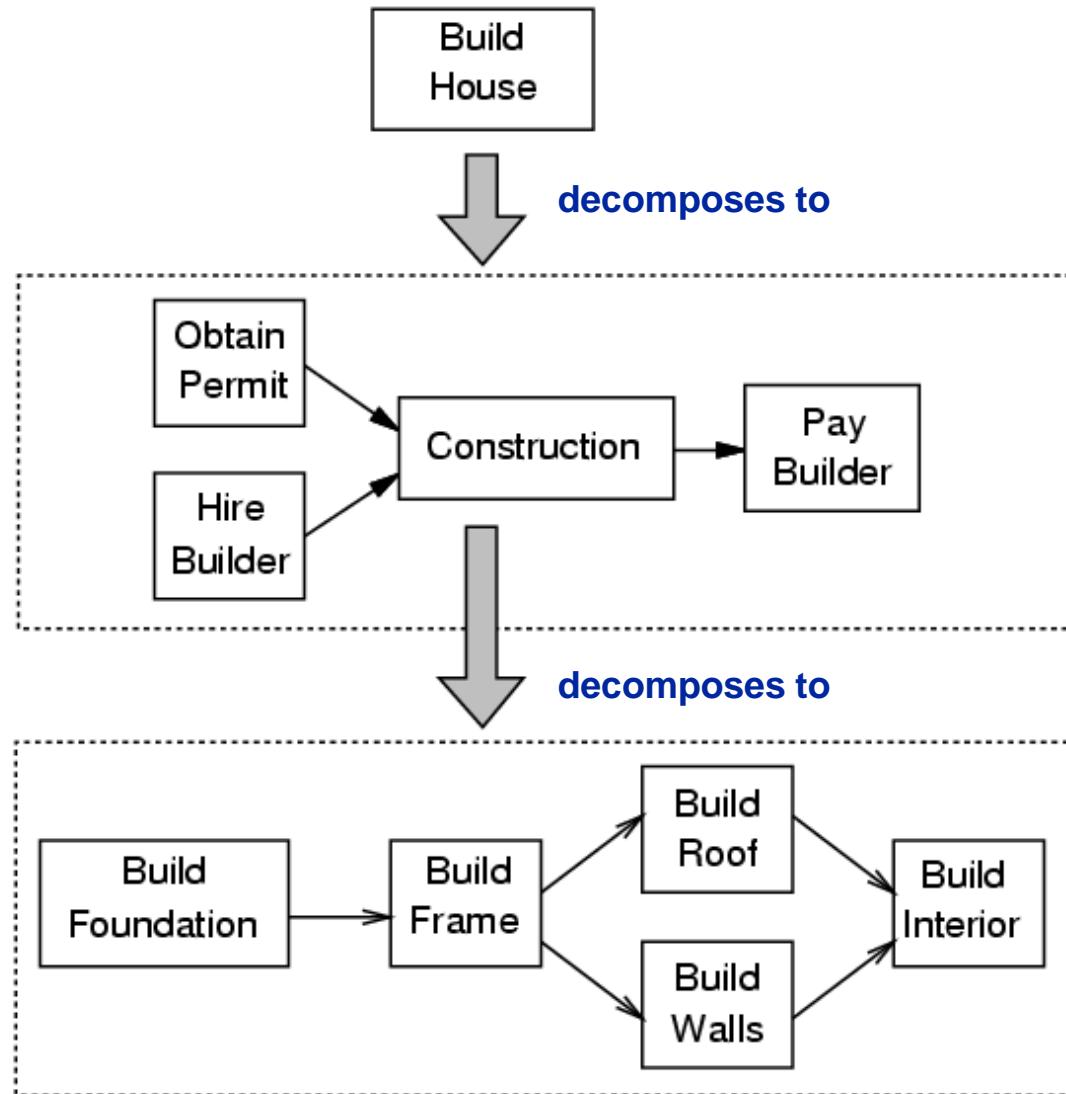
Outline

- STRIPS-like plan representation
- Planning with state-space search
- Partial-order planning (POP)
- Planning graphs (GraphPlan)
- Hierarchical task network (HTN) planning
- Probabilistic planning

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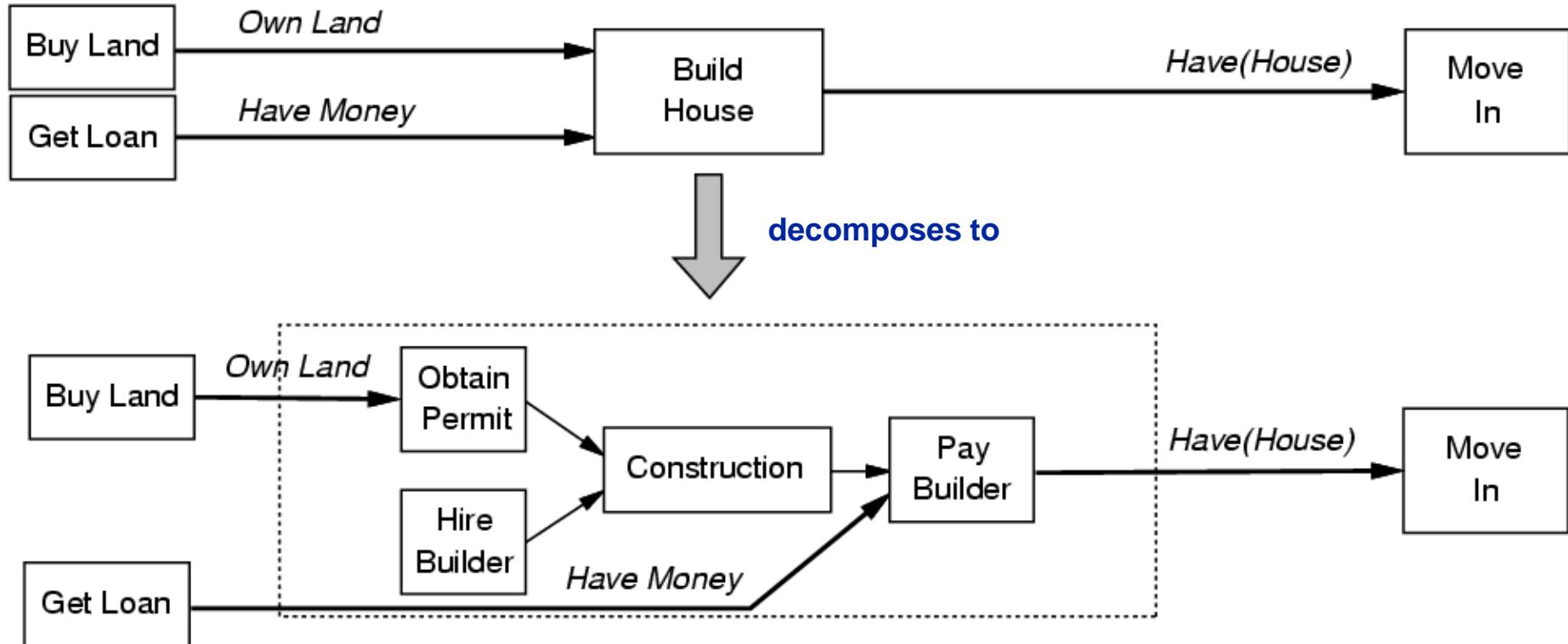
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Hierarchical decomposition of actions



Task reduction

从抽象到具体.



- Naturally encode hierarchical decomposition of actions
- Action hierarchy as domain-specific task knowledge

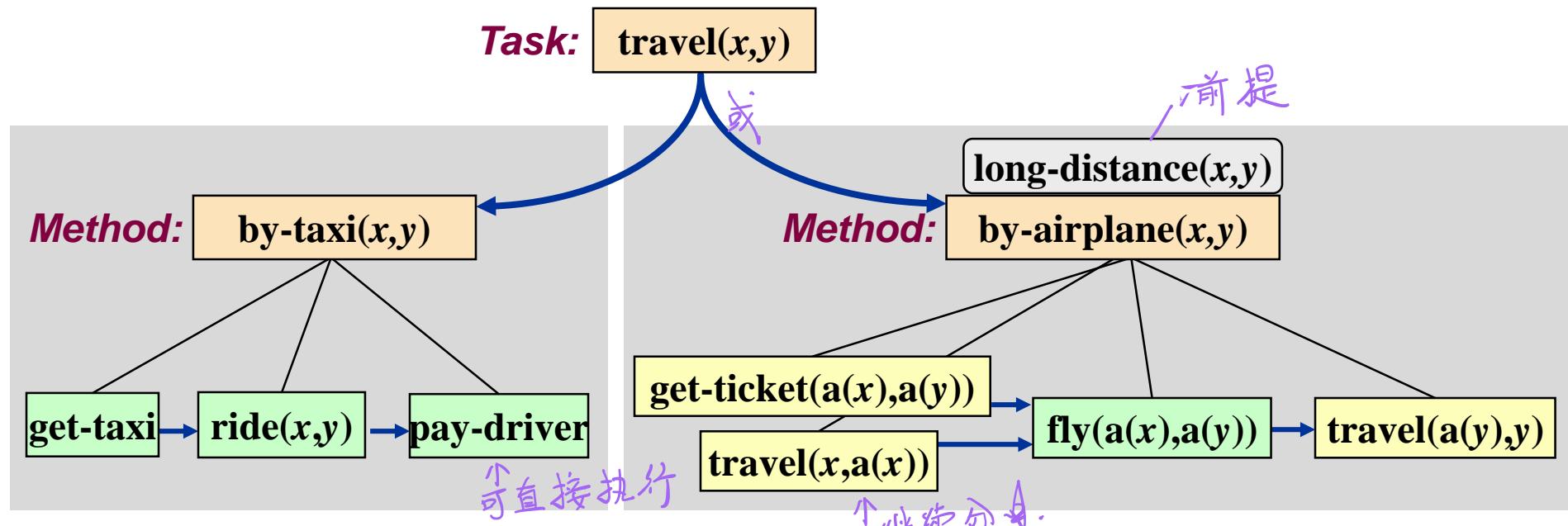
HTN (层次任务网络) planning

Hierarchical task network (HTN) planning [Erol et al 94]

构建 层次结构.

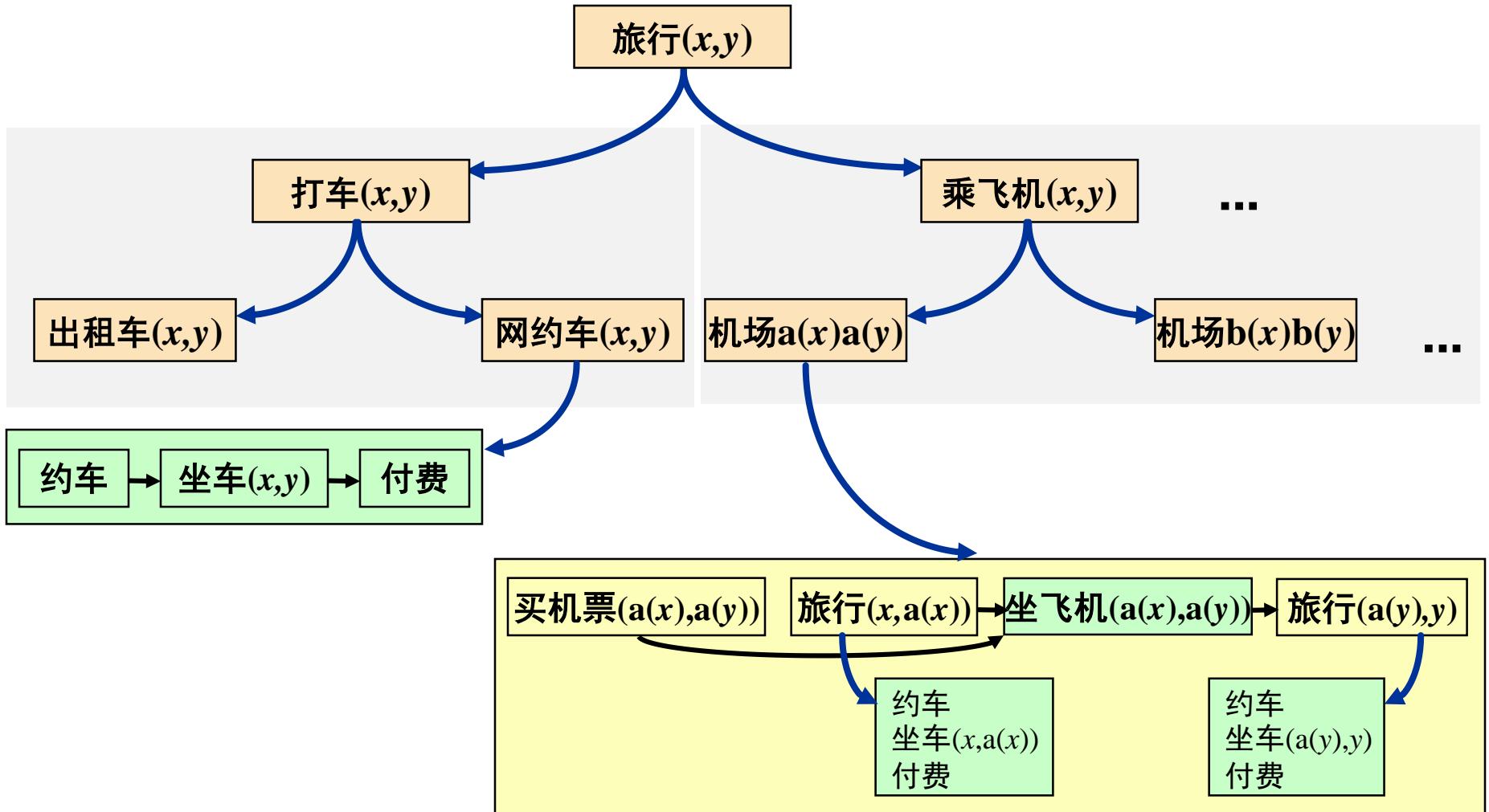
- Capture hierarchical structure of the planning domain
- Planning domain contains non-primitive actions and schemas for reducing them
抽象动作.
- Reduction schemas: 立观性比较大
 - Given by the designer
 - Describe preferred ways to accomplish a task

Task network example

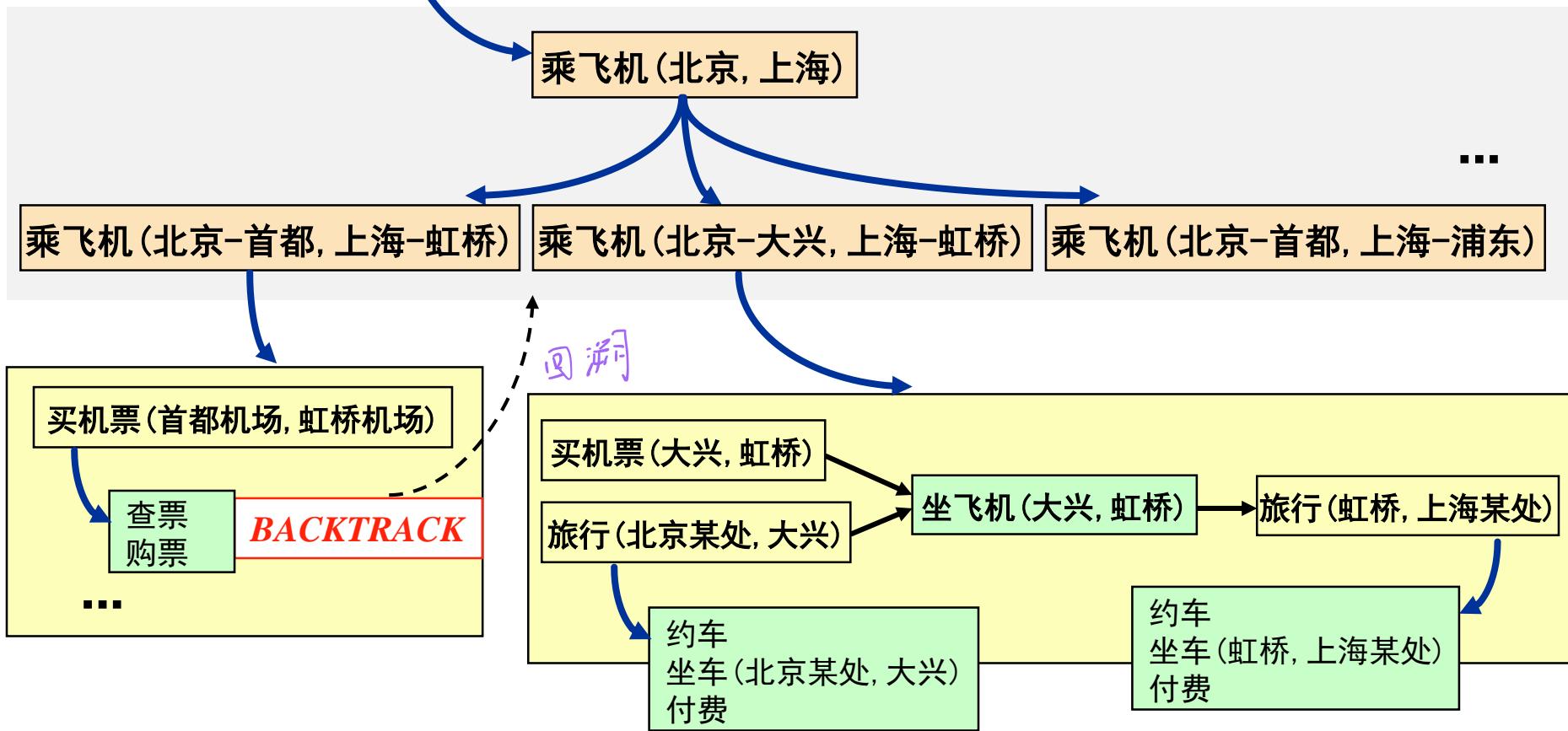


- Tasks rather than goals
 - Primitive (原始的) vs nonprimitive (抽象的) actions
- Methods describe task reduction
 - Ways to decompose tasks into subtasks

Task network example



Task network example

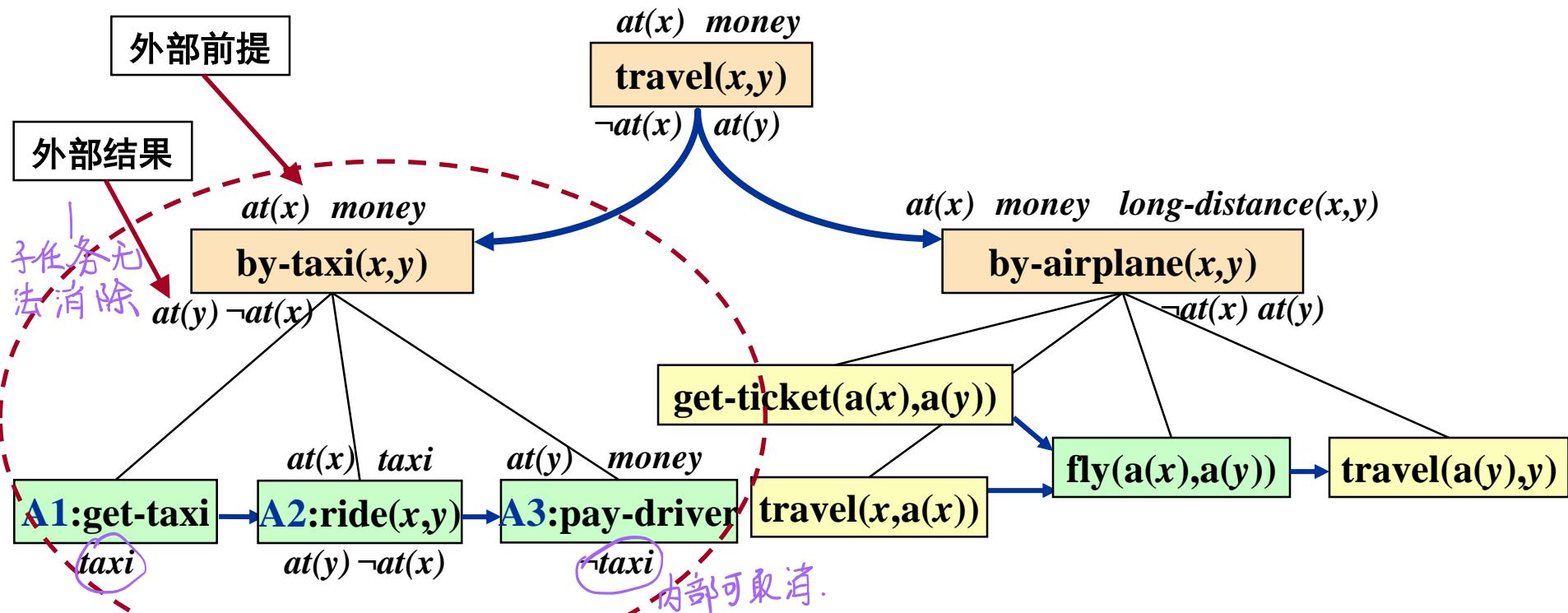


HTN planning intuition

Problem reduction:

- Decompose tasks into subtasks 分为任务为了任务.
- Handle constraints 处理 冲突
- Resolve interactions 解决交互问题
- If necessary, backtrack and try other decompositions
如有必要，需要回溯 和 尝试 其他 分解.

Task network example

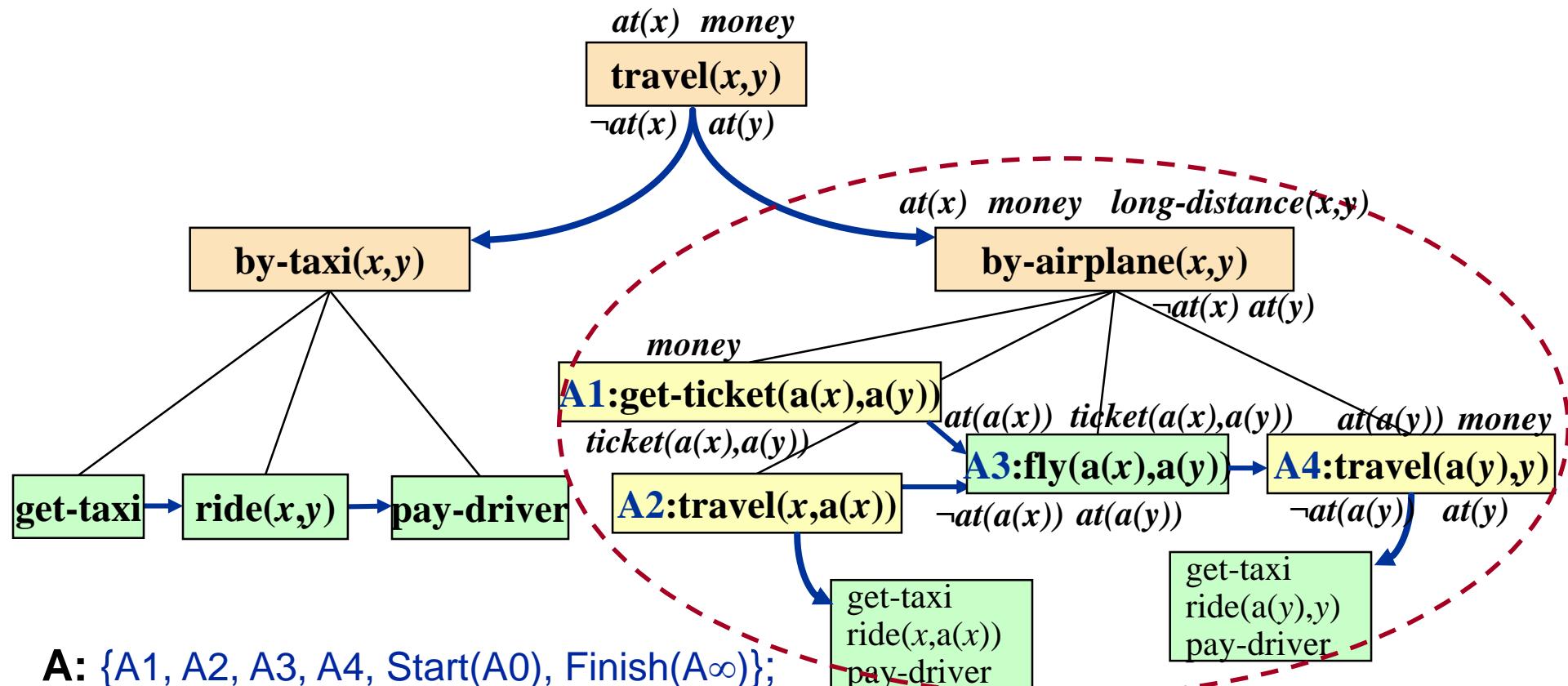


Primitive actions: $\{A1, A2, A3, Start(A0), Finish(A\infty)\}$;

Orderings: $\{A1 < A2, A2 < A3\}$

Causal links: $\{(A0 \xrightarrow{at(x)} A2), (A0 \xrightarrow{money} A3), (A1 \xrightarrow{taxi} A2), (A2 \xrightarrow{at(y)} A3), (A2 \xrightarrow{at(y)} A\infty)\}$

Task network example



A: {A1, A2, A3, A4, Start(A0), Finish(A ∞)};

O: {A1 < A3, A2 < A3, A3 < A4}

L: {(A0 $\xrightarrow{\text{money}}$ A1), (A0 $\xrightarrow{\neg at(x)}$ A2), (A0 $\xrightarrow{\text{money}}$ A2), (A0 $\xrightarrow{\text{money}}$ A4), (A1 $\xrightarrow{\text{tkt}(a(x),a(y))}$ A3),
 (A2 $\xrightarrow{\at(a(x))}$ A3), (A3 $\xrightarrow{\at(a(y))}$ A4), (A4 $\xrightarrow{\at(y)}$ A ∞)}

Basic HTN Procedure

1. Input a planning problem P
2. If P contains only *primitive* tasks, then resolve the conflicts and return *solution*: If the conflicts cannot be resolved, return *failure*
3. Choose a *non-primitive* task t in P
4. Choose an expansion for t
5. Replace t with the expansion
6. Find interactions among the tasks in P and suggest ways to handle them
7. Choose one way and apply it
7. Go to step 2

total ordering primitives
satisfying constraints

Find conflicts and suggest
ways to resolve them

Properties of HTN planning

Sound, complete and has formal semantics [Erol et al 94]

形式化语义，表达能力更强

- Strictly more expressive than classical planning
- Encode expert knowledge essential to task reasoning 使用了领域知识
- Handle actions and propositions with variables |

The idea of *task decomposition* is critical

- Efficient ways to represent complex task knowledge
- Practical planner applied to real-world problems 和真实世界类似
- Critical applications in large-scale intelligent systems 在大规模智能系统有重大应用.

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Classical planning

- Most advances seen *first* in classical planning
很多实际环境不满足
- Many practical environments don't satisfy classical planning assumptions
 - Possible to handle *minor* assumption violations through re-planning and execution monitoring 重规划 监测
 - Advantageous to rely on *widely-used* (and efficient) classical planning techniques
- Classical planning techniques often shed light on the effective ways of handling *non-classical* planning problems
 - Most of the efficient techniques for handling non-classical scenarios are based on classical ideas/advances 有指导意义

Classical planning assumptions

- Discrete time 时间 离散
 - All effects are immediate 结果实时.
 - Fully observable environment 完全可观测 环境
 - Deterministic effects 确定性结果 .
 - No dynamic changes in plan execution 在规划|执行过程中没有 动态变化
 - Planning agent is the sole agent of changes 单智能体变化
-

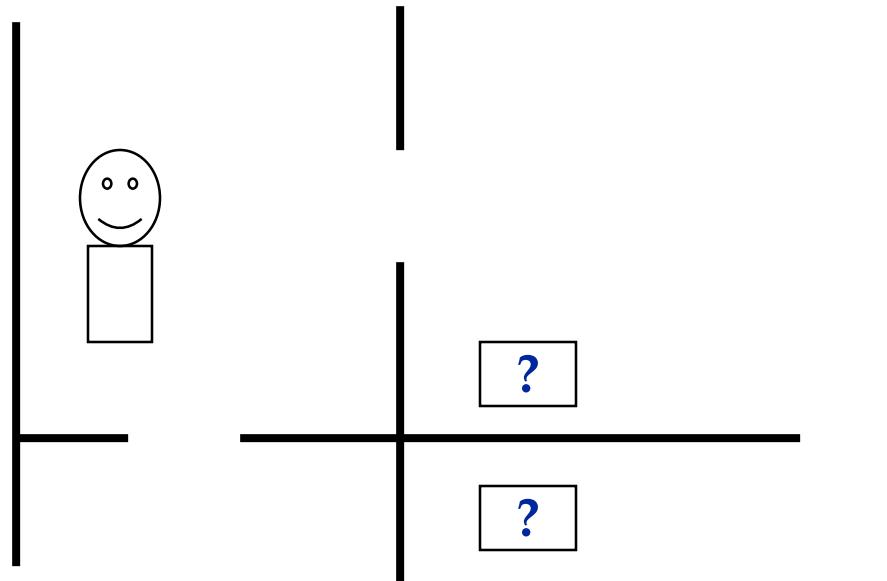
Classical planning assumptions

- Discrete time
 - All effects are immediate
 - **Fully observable environment**
 - **Deterministic effects**
 - **No dynamic changes in plan execution**
 - Planning agent is the sole agent of changes
-
- 
- Uncertainty**

Sources of uncertainty

不能观测所有状态

When we try to execute plans, uncertainty from several different sources can affect plan success.

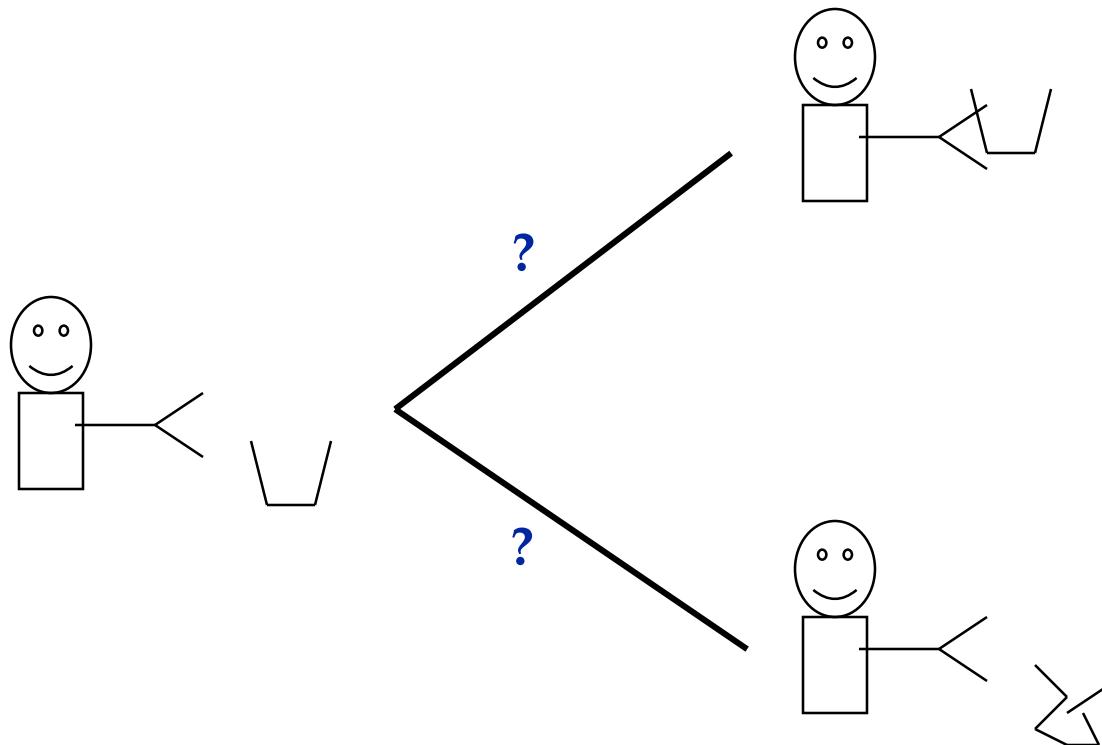


Firstly, we might have uncertainty about the *state of the world*.

Sources of uncertainty

结果不确定

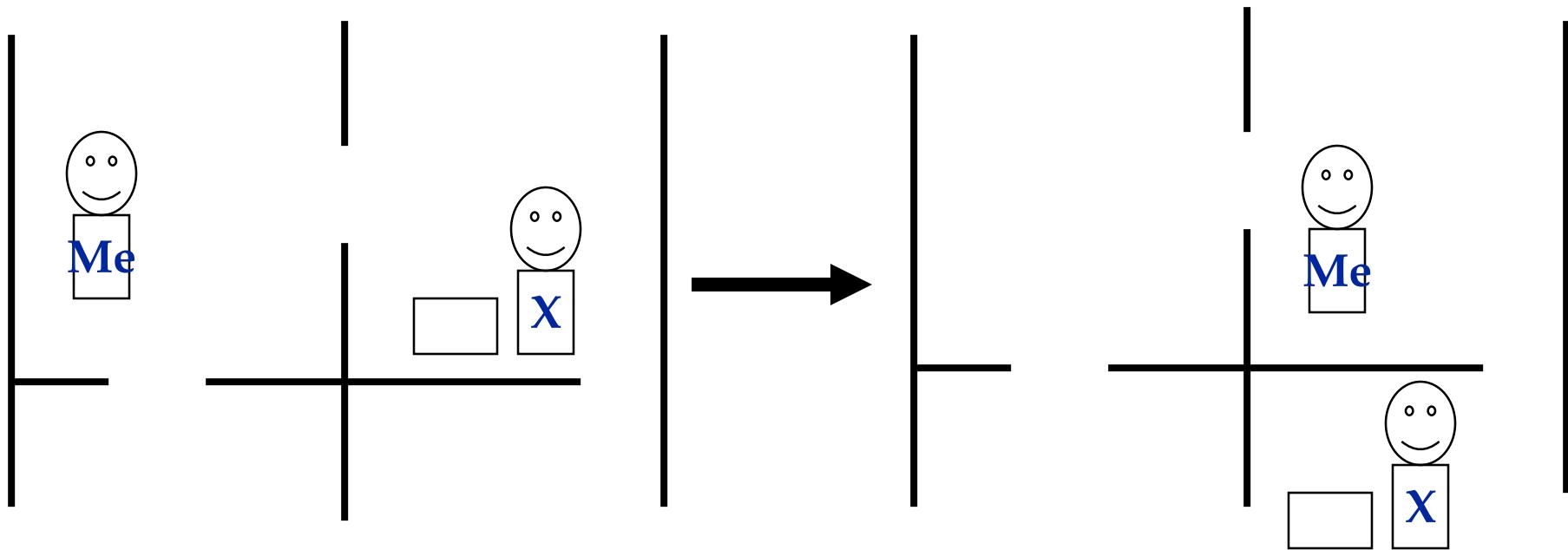
Actions we take might have *uncertain effects* even when we know the world state.



Sources of uncertainty

其他 agent 会改变状态

External agents might be changing the world while we execute our plan.



重规划

Dealing with uncertainty: Re-planning

- Make a plan assuming nothing bad will happen
- Monitor for problems during execution
- Build a *new plan* if a problem is found
 - Either re-plan to the goal state
 - Or try to amend the existing plan

① 无法实现实时监测。
② 结果可能无法即时挽回。

Dealing with uncertainty: Conditional planning

- Deal with *contingencies* (偶发事情、意外) at planning time before they occur
- Utilize *conditional planning* (条件规划) to address every possible contingency in the policy

E.g. if *<test1>* then *plan_A*
 else if *<test2>* then *plan_B*
 else if *<test3>* then *plan_C*
 else ...

无法考虑所有情况

Trade-offs in strategies for uncertainty

- Re-planning: Can't predict the plan steps are needed before the contingency is discovered
 ■ Re-planning agent: "Why are you taking an umbrella? It's not raining!"
可能来不及
- Conditional planning: Impossible to plan for every contingency
 ■ Conditional planning agent: "Why are you leaving the house? Class may be cancelled. It might rain. You might have won the lottery. Was that an earthquake?..."
考慮不了所有情況

只考慮比較重要的情況

Probabilistic planning: The middle ground

- Partial knowledge about *uncertainty*: different contingencies have different probabilities of occurring
- Plan ahead for *likely* contingencies that may need steps taken before they occur
- Use probability theory to judge plans that address some contingencies:
seek a plan that is above some minimum probability of success

Some issues to think about

- How to represent the occurrence of *likely* contingencies in action representation?
- How do we figure out the probability of *plan success*? What are the representations to support this?
- Can we distinguish bad *outcomes* (e.g. not holding the glass) from *really* bad outcomes (e.g. broken the glass, spilled the sulfuric acid..)?
后果有的严重 (效用价值)

Some issues to think about

- How to represent the occurrence of *likely* contingencies in action representation?
- How do we figure out the probability of *plan success*? What are the representations to support this?
- Can we distinguish bad *outcomes* (e.g. not holding the glass) from *really* bad outcomes (e.g. broken the glass, spilled the sulfuric acid..)?
 - Probabilistic action representation
 - Probability of plan success
 - Plan utility

Representing actions with uncertain outcomes

(at car ?x)
(in china car)

Drive-china ?x ?y

↓
~(at car ?x)
(at car ?y)
~(at china ?x)
(at china ?y)

(at car ?x)
(in china car)

Drive-china ?x ?y

0.7 0.3
↓ ↓
~(at car ?x)
(at car ?y)
~(at china ?x)
(at china ?y)
 ~ (at car ?x)
 (at car ?y)
 ~ (at china ?x)
 (at china ?y)
 (broken china)

Deterministic effects

Non-deterministic effects

“~” 表示否定，同 “ \neg ”

(at car ?x)
(in china car)

Drive-china ?x ?y

(packed china)

0.95 0.05

↓ ↓
~(at car ?x)
(at car ?y)
~(at china ?x)
(at china ?y)
 ~ (at car ?x)
 (at car ?y)
 ~ (at china ?x)
 (at china ?y)
 (broken china)

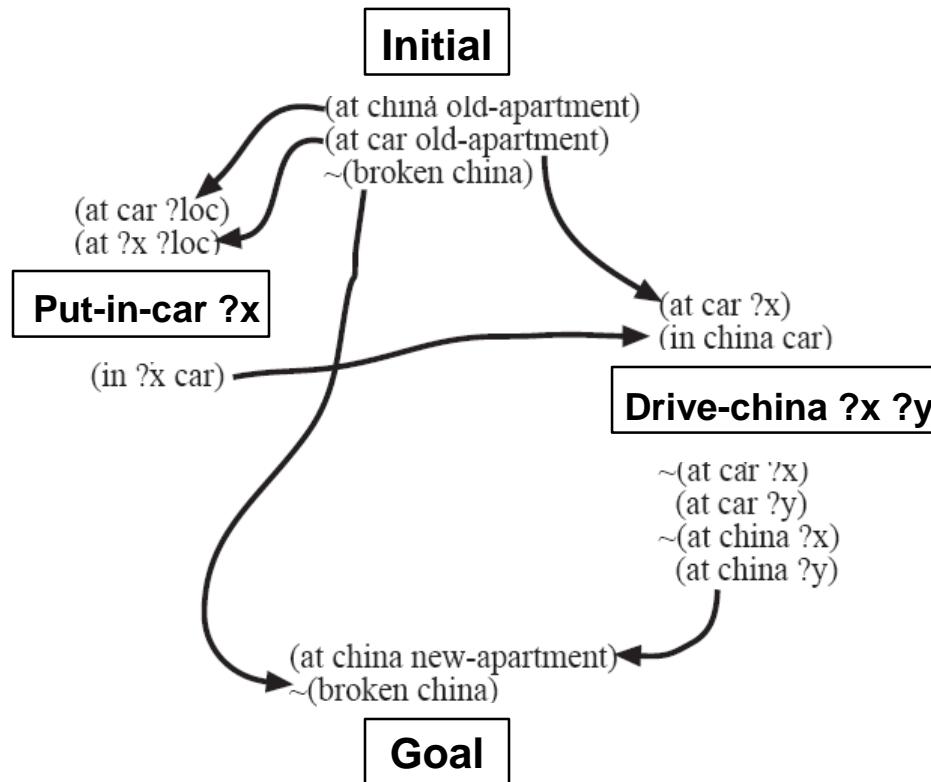
~(packed china)

0.7 0.3

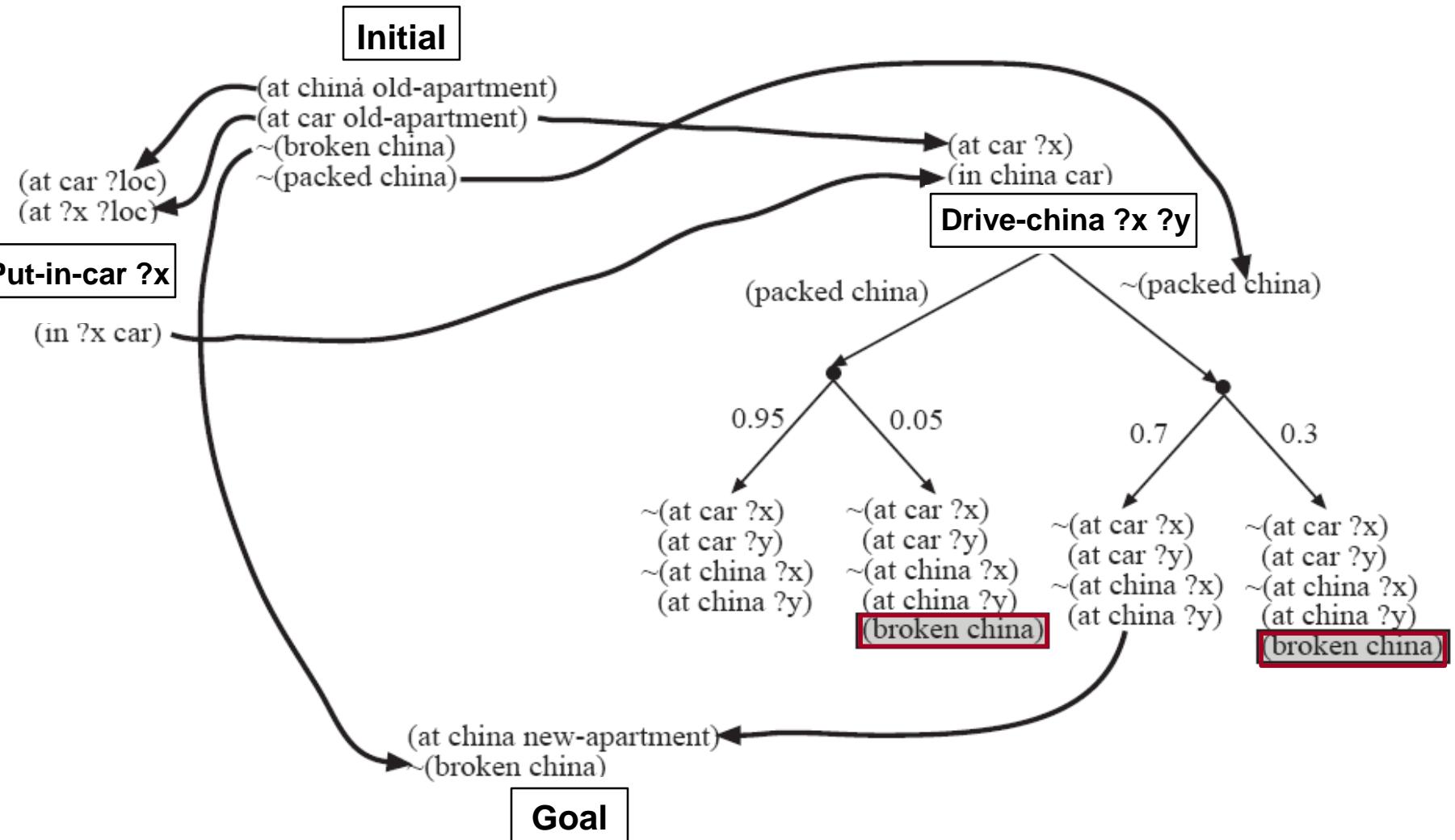
Non-deterministic effects with conditional probabilities

A partial-order plan to move china

- An POP-based planner might come up with this plan for deterministic action representation:



A plan that works 70% of the time...



Recall: POP algorithm

POP((A, O, L), agenda, actions)

Initial plan: $\{Start, Finish\}$ and preconditions in *Finish* as open conditions

1. If **agenda** is empty, then **return** (A, O, L) 结束条件
2. Pick (Q, A_{need}) from **agenda** (子) 目标
3. **Choose** an action A_{add} that adds effect Q 动作选择
 - If no such action exists, **fail**
 - Add the link $A_{add} \xrightarrow{Q} A_{need}$ to **L** and the ordering $A_{add} < A_{need}$ to **O**
 - If A_{add} is new, add it to **A** 规划扩充
4. Remove (Q, A_{need}) from **agenda**. If A_{add} is new, for each of its preconditions P add (P, A_{add}) to **agenda** 更新(子) 目标
5. For every action A_t in A that threatens any causal link A_p → A_c in **L**
 - **Choose** to add $A_t < A_p$ or $A_c < A_t$ to **O**
 - If neither choice is consistent, **fail**
6. POP((A, O, L), **agenda**, actions)

保护因果连接：
- 降级(Demotion): $A_t < A_p$
- 升级(Promotion): $A_c < A_t$

Modifications to the UCPOP algorithm

- Allow more than one causal link for each condition in the plan
- Confront a threat by decreasing the probability of an outcome in which the threatened condition is negated
- Terminate when sufficient probability reached (may still have threats)

Finding plans for moving china

■ Goals:

- (at china new-apartment)
- ~(broken china)

■ Actions:

- (at car ?loc)
- (at ?x ?loc)

Put-in-car ?x

- (in ?x car)

■ Plans:

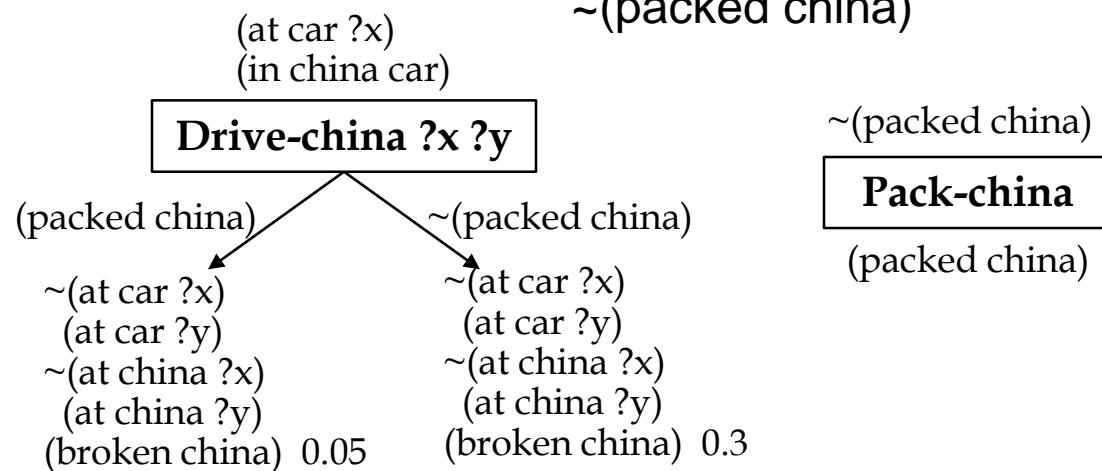
Start → {Pack-china, Put-in-car china} → Drive-china → *Finish* (Prob: 0.95)

Start → Put-in-car china → Drive-china → *Finish* (Prob: 0.70)

....

■ Initial states:

- (at china old-apartment)
- (at car old-apartment)
- ~(broken china)
- ~(packed china)



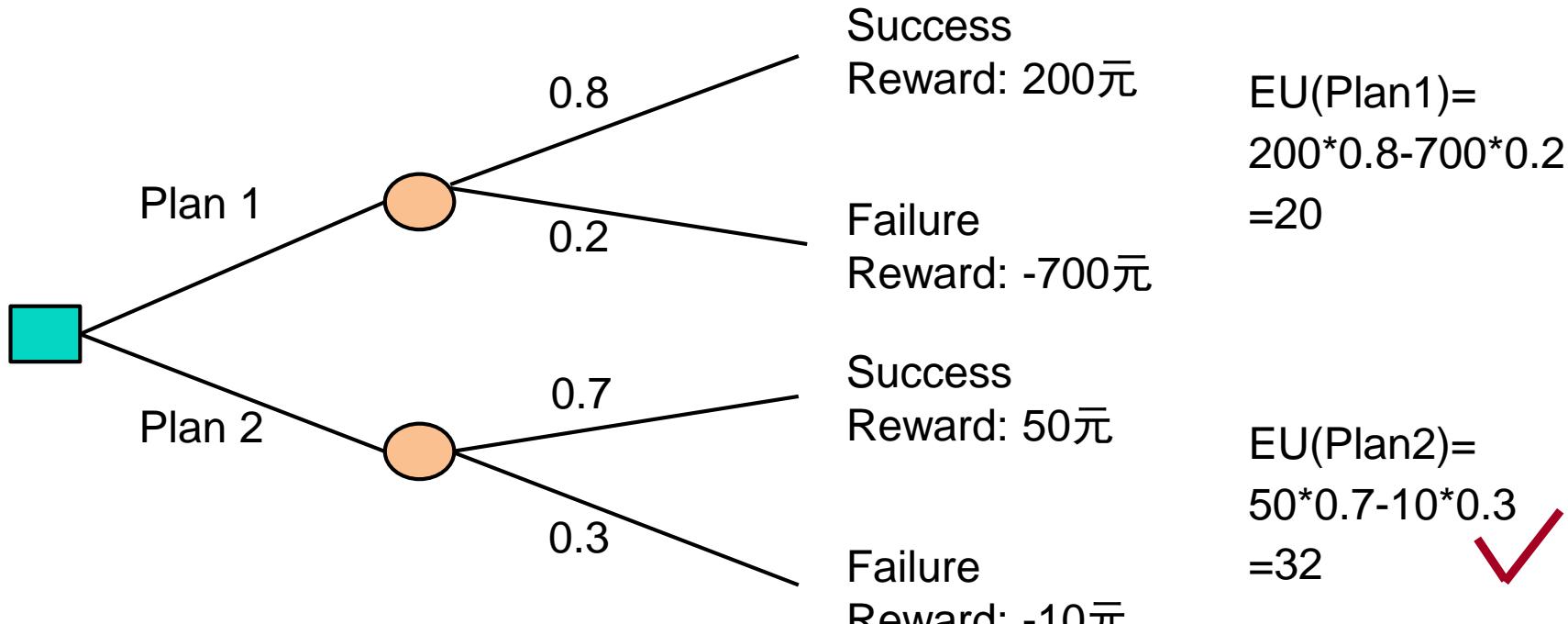
概率无法衡量后果的严重性

DRIPS: Decision-theoretic Refinement Planner

- Consider plan *utility*, taking into account action costs and benefits of different states
- Search for a plan with *Maximum Expected Utility* (*MEU*), not just above a threshold
- A skeletal planner, makes use of utility ranges of alternative plans in order to search efficiently
- Prune alternatives whose utility ranges are completely below the range of some *dominated plan*
剪枝

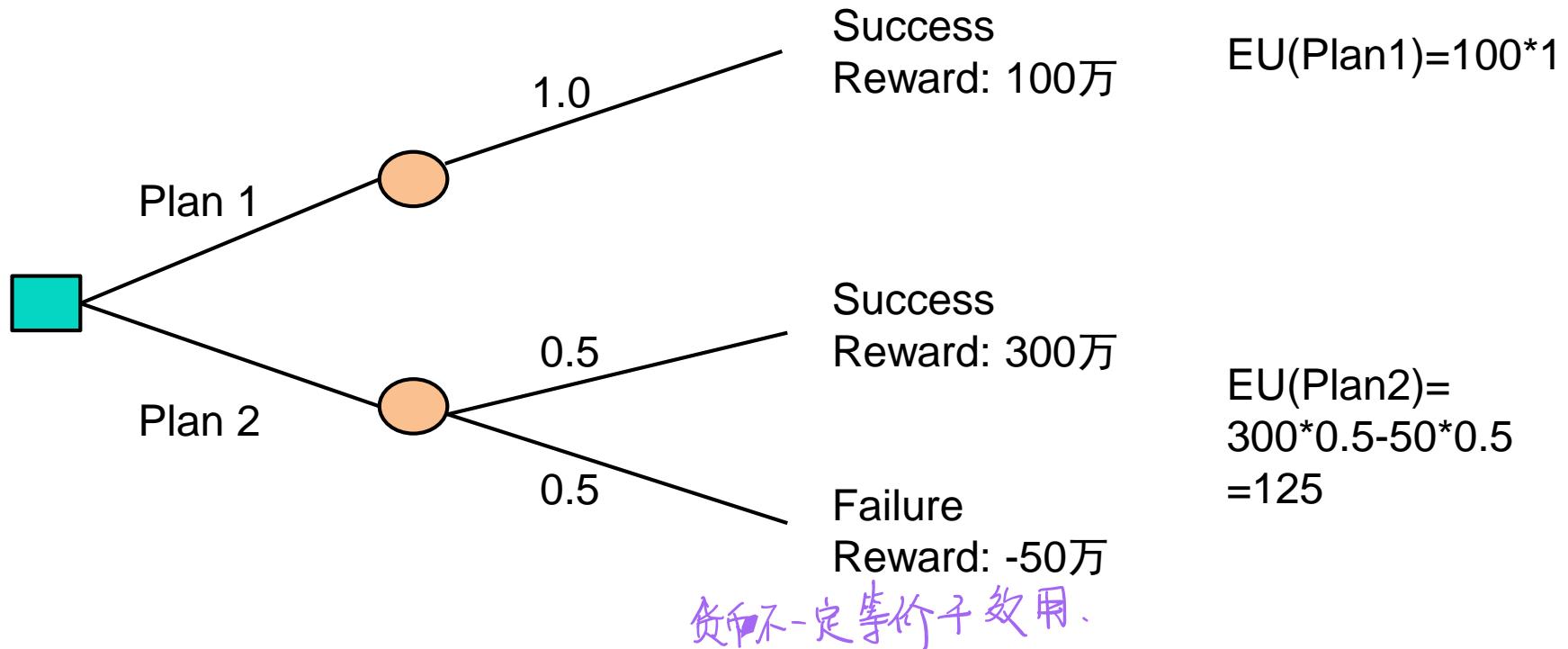
Example of decision theory

- People are utility maximizers



- Decision node: Agent plays
- Chance node: Nature plays

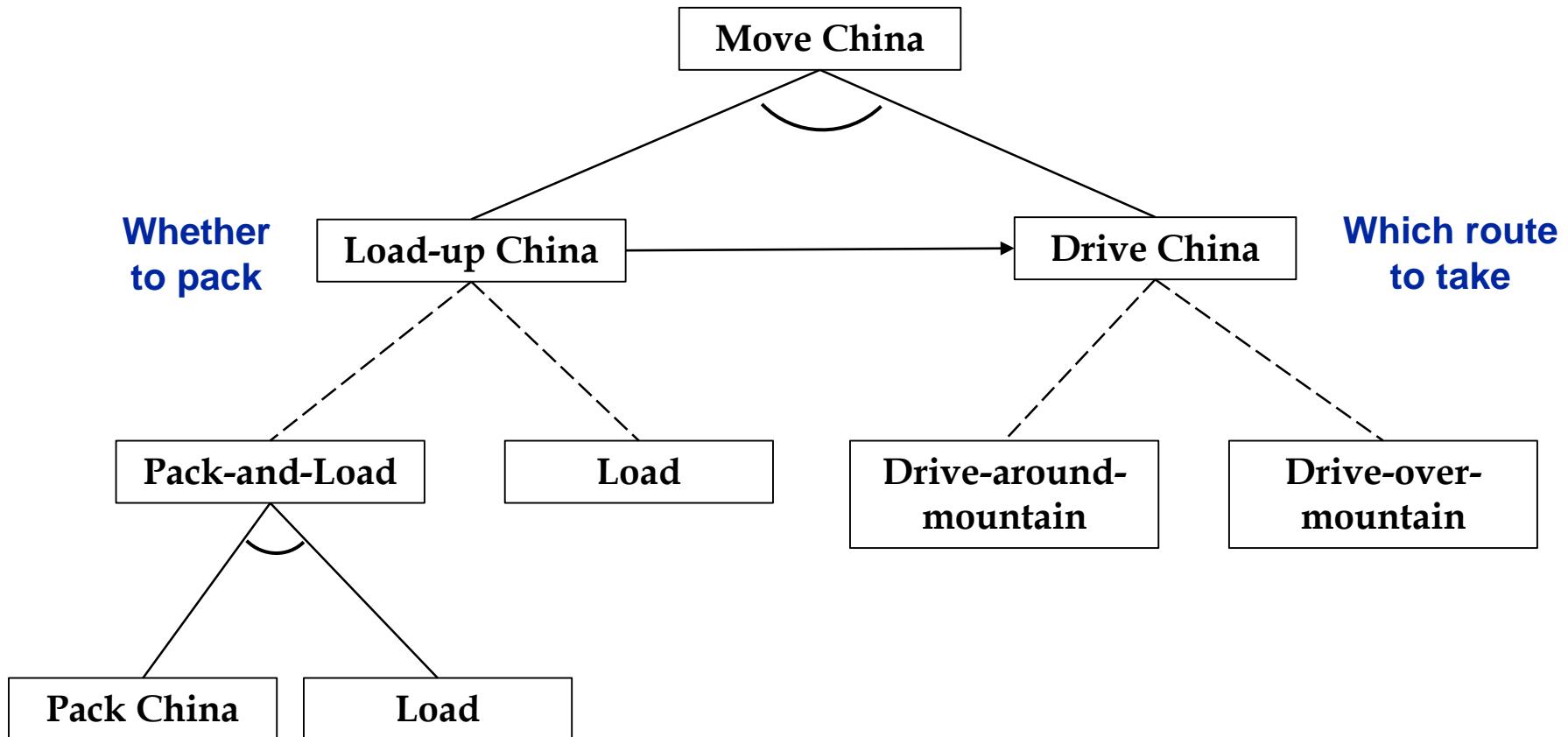
Example of decision theory



- Money is not necessarily equal to preferences ...
 - Utility theory does not assume money is people's utility function
 - People assign utility to money. Different people have different utility functions

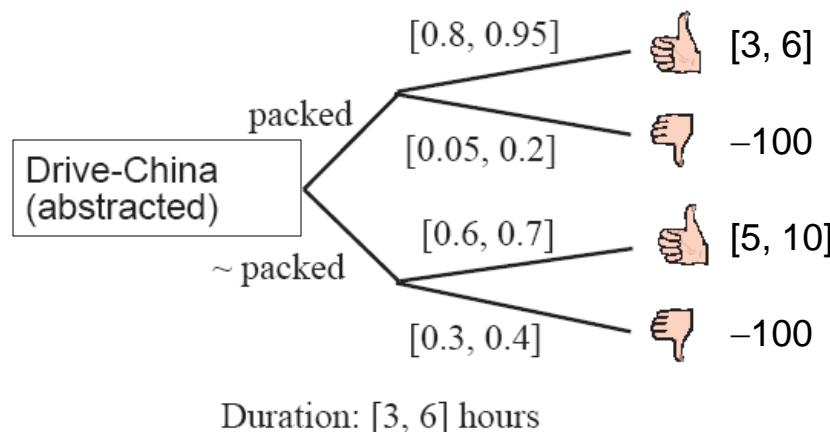
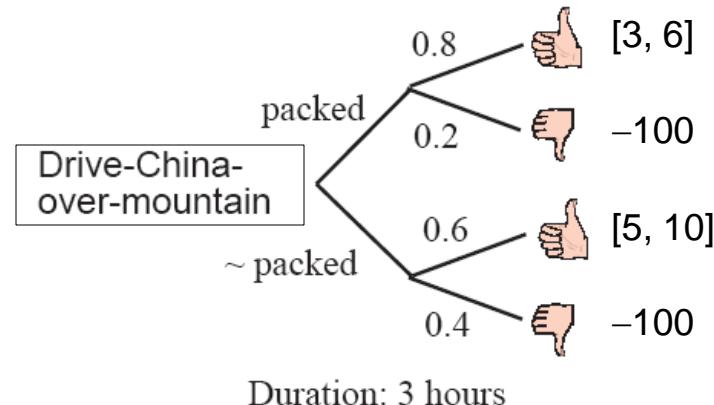
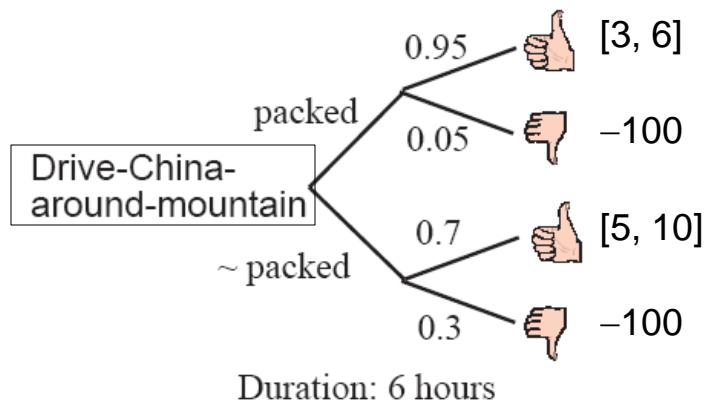
Abstract plan for moving china

- Use an *abstraction-decomposition network* to describe all possible specializations of an abstract plan to move china



Utilities of action outcomes

- Use utility to represent the costs / benefits of action outcomes



Refining abstract action Load-up-china

- Use utility ranges to compute the *dominance* between *alternative plans*

Plan A: Pack and load china

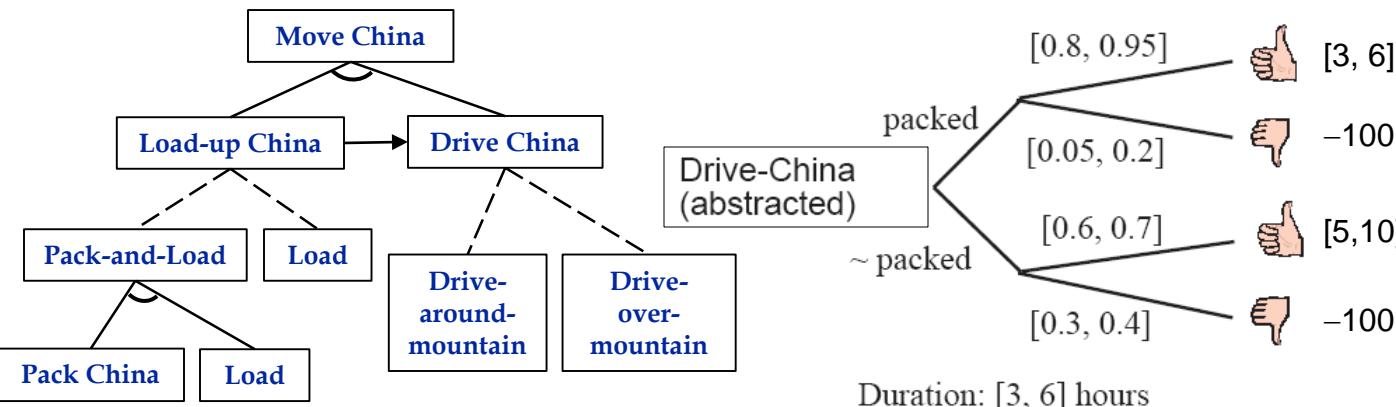
Outcome	Duration	Utility	Probability
Not broken	[3 6]	[3 6]	[0.8 0.95]
Broken	[3 6]	-100	[0.05 0.2]

Plan utility:
[-20+2.4, -5+5.7]

Plan B: Load china, don't pack

Not broken	[3 6]	[5 10]	[0.6 0.7]
Broken	[3 6]	-100	[0.3 0.4]

Plan utility:
[-40+3, -30+7]



Other approaches

MaxPlan (Inspired by SATPLAN):

- Given a boolean formula **with variables** (either *choice* variables or *chance* variables), find an assignment to the choice variables **that maximizes the probability that the formula is true**
- We have control to *choice* variables:
 - e.g. which actions to use
- We cannot control *chance* variables:
 - e.g. the weather, the outcome of each action ...
- Then use standard algorithm **to compute and maximize probability of success**

Other approaches

MDPs and POMDPs:

大规模. 无数爆炸

- MDP-based approaches can find optimal solutions and deal elegantly with costs
- However,
 - They use rather compact action representation
 - Need to control potential state space explosion
 - Complexity issue and tractability

Recent trends in planning research

Trends:

- Combining *planning* with *learning* technique
 - Bridging the Gap between AI Planning and Reinforcement Learning (PRL) Workshop @ ICAPS 2020-2023
 - Bridging the Gap between AI Planning and Reinforcement Learning (PRL) Workshop @ IJCAI 2022
- Explainable automated *planning* and decision making
 - Explainable AI Planning (XAIP) Workshop @ ICAPS 2018-2022
 - Explainable AI (XAI) Workshop @ IJCAI 2017-2020, 2022
 - Explainable AI Workshop 2021 Organized by NIST
 - Explainable and Transparent AI and MAS (EXTRAAMAS) Workshop @ AAMAS 2019-2023

小结：关于规划

偏序规划 (POP)
图规划 (GraphPlan)
约束满足规划 (SATPlan)
时序规划 (TLPlan)

Classical planning

Hierarchical
extension

层次任务网络规划
(HTN)

Probabilistic
extension

概率规划

Probabilistic planning

Multi-agent
extension

多智能体规划

Multi-Agent planning

Multi-agent planning

结合自身特长 处理其他 Agents

- Collective effort of multiple agents to combine their knowledge, information and capabilities so as to develop solution(s) to the planning problem that each agent could not have solved as well alone
- Assume that agents are cooperative: each agent must formulate plans for what it will do that take into account (sufficiently well) the plans of other agents

合作情况下

除了自己目标，考虑其他 agent
或集体目标

What aspects are multi-agent?

- Multi-agent planning could refer to just the *results* of the planning process:

A centralized planning process formulates plans that specify how each of the multiple agents should do

- Centralized planning for distributed plans 分发任务.

- Multi-agent planning could refer to the *process* of formulating plans:

Multiple agents participate in the construction of a single joint plan

- Distributed planning for centralized plans 部分规划 合并为
单-规划)

What aspects are multi-agent?

- Both the planning process and results are multi-agent:
 - Each agent applies its local expertise and awareness to develop its local plan
 - Agents use communication and/or shared knowledge to shape their local plans and conform better to others' plans, in order to work together effectively
- Distributed planning for distributed plans

Levels of working together

	As a process	As a result
最基本 Coordination 协调	没有集体利益，只是考虑 Decision-making takes others into account	Individuals are not thwarted by others
Cooperation 合作	Individual decisions further the collective welfare	Individuals appear to be “working together”
Collaboration /Teamwork 团队任务	Individuals work toward shared goals 共享目标	Common goals are achieved well and/or efficiently

Coordinating multiple agents 契约

- Social conventions (协定)
- Flocking behavior in nature
 - Cohesion (凝聚): get closer to the average position of neighbors
 - Separation (分离): avoid getting too close to any one neighbor
 - Alignment (对齐): get closer to the average heading of neighbors
- Agent communication
- Agent modeling
 - Build mind model of other agents 建立其他 agent 心智模型
 - Intention/goal recognition 行为意图识别



Agent modeling

- Representing and reasoning about *other* agents
 - Other agents' capabilities, skills, responsibilities
 - Other agents' mental states, plans, goals, actions
 - Other teams' goals, plans, states
- ...
- *Explicitly* model other agents
 - *Implicit* modeling if all coordination with other agents compiled in without actual representation and reasoning about other agents

Modeling other agents: A critical technique

- In dynamic, complex, multi-agent environments:
 - Agents respond unpredictably to the environment
 - Interactions with other agents not always feasible
 - E.g. communications failures, opponent behavior, etc
 - State of other agents not known at design time
- Agents must model other agents at *execution time*
 - To detect failures in their execution (leading to diagnosis, recovery)
 - To coordinate with others, render assistance
 - To visualize, ascertain progress
 - To compete against other agents

....

Plan recognition (规划/意图识别)

- Key agent modeling technique
 - Infer other agent's current plans and goals based on the available *observations*
 - Other agents may be unable to *communicate* directly for a variety of reasons
- Major challenge: *Ambiguity* 歧义.
 - There may be multiple plans to explain the observed actions

Resources

- UCPOP Planner:
<https://www.swmath.org/software/20687>
- GraphPlan Planner:
<https://en.wikipedia.org/wiki/Graphplan> (External links)
- UMCP Planner:
<http://www.cs.umd.edu/projects/plus/umcp>
- SHOP/SOP2 Planner:
<http://www.cs.umd.edu/projects/shop>
- Planning Domain Definition Language (PDDL):
<https://planning.wiki/guide/whatis/pddl>

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