

Planning and Optimization

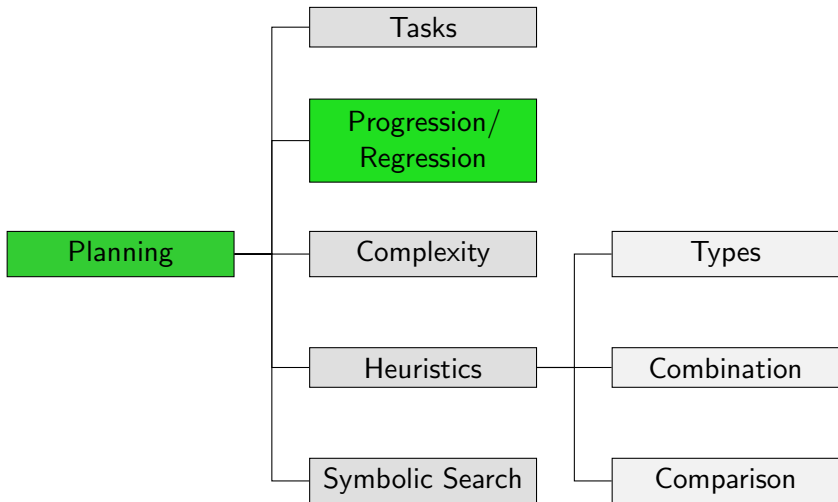
B2. Regression: Introduction & STRIPS Case

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Content of this Course



Regression

Forward Search vs. Backward Search

Searching planning tasks in forward vs. backward direction is **not symmetric**:

- forward search starts from a **single** initial state;
backward search starts from a **set** of goal states
 - when applying an operator o in a state s in forward direction, there is a **unique successor state** s' ;
if we just applied operator o and ended up in state s' , there can be **several possible predecessor states** s
- ↪ in most natural representation for backward search in planning, each search state corresponds to a **set of world states**

Planning by Backward Search: Regression

Regression: Computing the possible predecessor states $\text{regr}(S', o)$ of a set of states S' (“subgoal”) given the last operator o that was applied.

↪ formal definition in next chapter

Regression planners find solutions by backward search:

- start from set of goal states
- iteratively pick a previously generated subgoal (state set) and **regress it** through an operator, generating a new subgoal
- solution found when a generated subgoal includes initial state

pro: can handle many states simultaneously

con: basic operations complicated and expensive

Search Space Representation in Regression Planners

identify state sets with **logical formulas** (again):

- each **search state** corresponds to a **set of world states** (“subgoal”)
- each search state is represented by a **logical formula**:
 φ represents $\{s \in S \mid s \models \varphi\}$
- many basic search operations like detecting duplicates are NP-complete or coNP-complete

Search Space for Regression

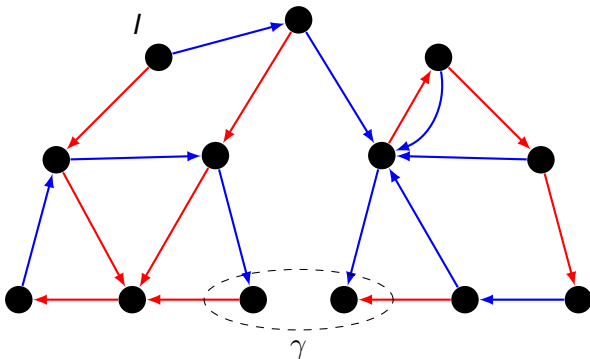
Search Space for Regression

search space for regression in a planning task $\Pi = \langle V, I, O, \gamma \rangle$
(search states are formulas φ describing sets of world states;
actions of search space are operators $o \in O$)

- **init()** \rightsquigarrow returns γ
- **is_goal(φ)** \rightsquigarrow tests if $I \models \varphi$
- **succ(φ)** \rightsquigarrow returns all pairs $\langle o, \text{regr}(\varphi, o) \rangle$
where $o \in O$ and $\text{regr}(\varphi, o)$ is defined
- **cost(o)** \rightsquigarrow returns $\text{cost}(o)$ as defined in Π
- **h(φ)** \rightsquigarrow estimates cost from I to φ (\rightsquigarrow Parts C–F)

Regression Example

Regression Planning Example (Depth-first Search)



Regression Planning Example (Depth-first Search)

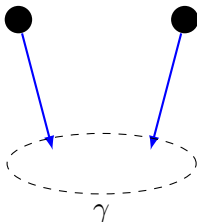
γ



Regression Planning Example (Depth-first Search)

$$\varphi_1 = \text{regr}(\gamma, \rightarrow)$$

$$\varphi_1 \rightarrow \gamma$$

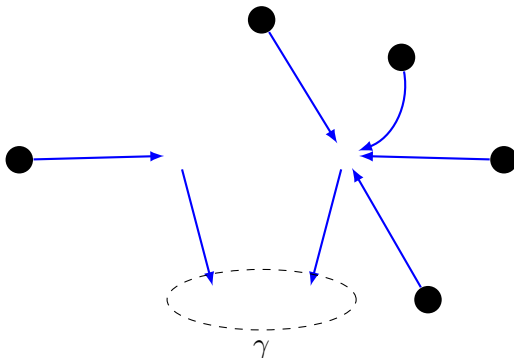


Regression Planning Example (Depth-first Search)

$$\varphi_1 = \text{regr}(\gamma, \rightarrow)$$

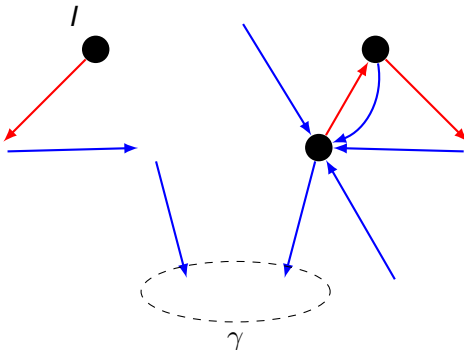
$$\varphi_2 = \text{regr}(\varphi_1, \rightarrow)$$

$$\varphi_2 \rightarrow \varphi_1 \rightarrow \gamma$$



Regression Planning Example (Depth-first Search)

$$\begin{aligned}\varphi_1 &= \text{regr}(\gamma, \rightarrow) & \varphi_3 &\xrightarrow{\text{red}} \varphi_2 \xrightarrow{\text{blue}} \varphi_1 \xrightarrow{\text{blue}} \gamma \\ \varphi_2 &= \text{regr}(\varphi_1, \rightarrow) \\ \varphi_3 &= \text{regr}(\varphi_2, \rightarrow), I \models \varphi_3\end{aligned}$$



Regression for STRIPS Tasks

Regression for STRIPS Planning Tasks

Regression for conflict-free STRIPS planning tasks

is much simpler than the general case:

- Consider subgoal φ that is conjunction of atoms $a_1 \wedge \dots \wedge a_n$ (e.g., the original goal γ of the planning task).
- **First step**: Choose an operator o that deletes no a_i .
- **Second step**: Remove any atoms added by o from φ .
- **Third step**: Conjoin $pre(o)$ to φ .

↪ Outcome of this is **regression** of φ w.r.t. o .
It is again a **conjunction of atoms**.

optimization: only consider operators adding at least one a_i

Note: “conflict-free” is not a serious restriction for STRIPS tasks

STRIPS Regression

Definition (STRIPS Regression)

Let $\varphi = \varphi_1 \wedge \dots \wedge \varphi_n$ be a conjunction of atoms, and let o be a conflict-free STRIPS operator which adds the atoms a_1, \dots, a_k and deletes the atoms d_1, \dots, d_l . (W.l.o.g., $a_i \neq d_j$ for all i, j .)

The **STRIPS regression** of φ with respect to o is

$$\text{sregr}(\varphi, o) := \begin{cases} \perp & \text{if } \varphi_i = d_j \text{ for some } i, j \\ \text{pre}(o) \wedge \bigwedge (\{\varphi_1, \dots, \varphi_n\} \setminus \{a_1, \dots, a_k\}) & \text{else} \end{cases}$$

Note: $\text{sregr}(\varphi, o)$ is again a conjunction of atoms, or \perp .

Does this Capture the Idea of Regression?

For our definition to capture the concept of **regression**, it should satisfy the following property:

Regression Property

For all sets of states described by a conjunction of atoms φ , all states s and all STRIPS operators o ,

$$s \models \text{sregr}(\varphi, o) \quad \text{iff} \quad s[[o]] \models \varphi.$$

This is indeed true. We do not prove it now because we prove this property for general regression (not just STRIPS) later.

Summary

Summary

- **Regression search** proceeds backwards from the goal.
- Each search state corresponds to a **set of world states**, for example represented by a **formula**.
- Regression is simple for **STRIPS** operators.
- The theory for **general regression** is more complex. This is the topic of the following chapters.