Planning and Optimization B3. General Regression, Part I

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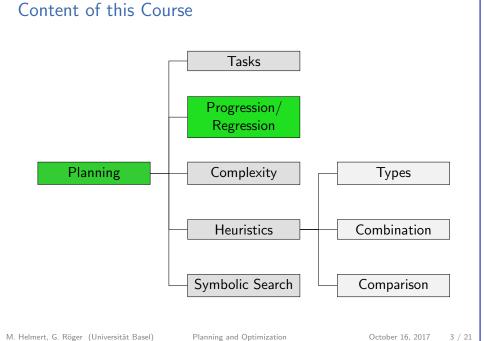
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- **B3.1** Regressing State Variables
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Regression for General Planning Tasks

- ▶ With disjunctions and conditional effects, things become more tricky. How to regress $a \lor (b \land c)$ with respect to $\langle q, d \rhd b \rangle$?
- ▶ In this chapter, we show how to regress general sets of states through general operators.
- ▶ We extensively use the idea of representing sets of states as formulas.

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B3. General Regression, Part I Regressing State Variables

B3.1 Regressing State Variables

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B3. General Regression, Part I

Regressing State Variables

Regressing State Variables: Motivation

Key question for general regression:

- ▶ Assume we are applying an operator with effect *e*.
- ► What must be true in the predecessor state for state variable *v* to be true in the successor state?

If we can answer this question, a general definition of regression is only a small additional step.

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B3. General Regression, Part I

Regressing State Variables

Regressing State Variables: Key Idea

Assume we are in state s and apply effect e to obtain successor state s'.

State variable v is true in s' if

- ▶ effect *e* makes it true, or
- ▶ it remains true, i.e., it is true in s and not made false by e.

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Regressing State Variables

Regressing a State Variable Through an Effect

Definition (Regressing a State Variable Through an Effect)

Let e be an effect of a propositional planning task, and let v be a state variable.

The regression of v through e, written regr(v, e), is defined as the following logical formula:

$$regr(v, e) = effcond(v, e) \lor (v \land \neg effcond(\neg v, e)).$$

Question: Does this capture add-after-delete semantics correctly?

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Regressing State Variables

Regressing State Variables: Example

Example

Let
$$e = (b \triangleright a) \land (c \triangleright \neg a) \land b \land \neg d$$
.

$$\begin{array}{c|c} v & regr(v,e) \\ \hline a & b \lor (a \land \neg c) \\ b & \top \lor (b \land \neg \bot) \equiv \top \\ c & \bot \lor (c \land \neg \bot) \equiv c \\ d & \bot \lor (d \land \neg \top) \equiv \bot \\ \hline \end{array}$$

Reminder: $regr(v, e) = effcond(v, e) \lor (v \land \neg effcond(\neg v, e))$

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Regressing State Variables

Regressing State Variables: Correctness (1)

Lemma (Correctness of regr(v, e))

In a propositional planning task, let v be a state variable, o an operator and s a state in which o is applicable.

Then $s \models regr(v, eff(o))$ iff $s[o] \models v$.

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Proof (continued).

(⇐): Proof by contraposition.

and hence $s \not\models effcond(v, e)$.

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Regressing State Variables: Correctness (3)

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Regressing State Variables

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Regressing State Variables

Regressing State Variables: Correctness (2)

Proof.

Let e = eff(o).

 (\Rightarrow) : We know $s \models regr(v, e)$, and hence $s \models effcond(v, e) \lor (v \land \neg effcond(\neg v, e)).$

Do a case analysis on the two disjuncts.

Case 1: $s \models effcond(v, e)$.

Then $s[o] \models v$ by the first case in the def. of s[o] (Chapter A4).

Case 2: $s \models (v \land \neg effcond(\neg v, e))$.

Then $s \models v$ and $s \not\models effcond(\neg v, e)$.

We may additionally assume $s \not\models effcond(v, e)$

because otherwise we can apply Case 1 of this proof.

Then $s[o] \models v$ by the third case in the def. of s[o] (Chapter A4).

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▶ Case 1: $s \models \neg v$. Then v is false before applying o and remains false, so $s[o] \not\models v$. ▶ Case 2: $s \models effcond(\neg v, e)$. Then v is deleted by o and not simultaneously added, so $s[o] \not\models v$.

We show that if regr(v, e) is false in s, then v is false in s[o].

▶ Hence $s \models \neg effcond(v, e) \land (\neg v \lor effcond(\neg v, e))$.

▶ From the first conjunct, we get $s \models \neg effcond(v, e)$

▶ By prerequisite, $s \not\models effcond(v, e) \lor (v \land \neg effcond(\neg v, e))$.

▶ From the second conjunct, we get $s \models \neg v \lor effcond(\neg v, e)$.

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Effects

Regressing Formulas Through Effects

Regressing Formulas Through Effects: Idea

- ► We can now generalize regression from state variables to general formulas over state variables.
- The basic idea is to replace every occurrence of every state variable v by regr(v, e) as defined in the previous section.
- ▶ The following definition makes this more formal.

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B3.2 Regressing Formulas Through

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Regressing Formulas Through Effects

Regressing Formulas Through Effects: Definition

Definition (Regressing a Formula Through an Effect)

In a propositional planning task, let e be an effect, and let φ be a formula over state variables.

The regression of φ through e, written $regr(\varphi, e)$, is defined as the following logical formula:

$$\begin{aligned} \mathit{regr}(\top, e) &= \top \\ \mathit{regr}(\bot, e) &= \bot \\ \mathit{regr}(v, e) &= \mathit{effcond}(v, e) \lor (v \land \neg \mathit{effcond}(\neg v, e)) \\ \mathit{regr}(\neg \psi, e) &= \neg \mathit{regr}(\psi, e) \\ \mathit{regr}(\psi \lor \chi, e) &= \mathit{regr}(\psi, e) \lor \mathit{regr}(\chi, e) \\ \mathit{regr}(\psi \land \chi, e) &= \mathit{regr}(\psi, e) \land \mathit{regr}(\chi, e). \end{aligned}$$

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Regressing Formulas Through Effects

Regressing Formulas Through Effects: Example

Example

Let
$$e = (b \triangleright a) \land (c \triangleright \neg a) \land b \land \neg d$$
.

Recall:

- $ightharpoonup regr(a, e) \equiv b \lor (a \land \neg c)$
- $regr(b, e) \equiv \top$
- $regr(c, e) \equiv c$
- $regr(d, e) \equiv \bot$

We get:

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$$regr((a \lor d) \land (c \lor d), e) \equiv ((b \lor (a \land \neg c)) \lor \bot) \land (c \lor \bot)$$
$$\equiv (b \lor (a \land \neg c)) \land c$$
$$\equiv b \land c$$

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Regressing Formulas Through Effects: Correctness (1)

Lemma (Correctness of $regr(\varphi, e)$)

Let φ be a logical formula, o an operator and s a state in which o is applicable.

Then $s \models regr(\varphi, eff(o))$ iff $s[o] \models \varphi$.

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Proof.

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The proof is by structural induction on φ .

Induction hypothesis: $s \models regr(\psi, eff(o))$ iff $s[o] \models \psi$ for all proper subformulas ψ of φ .

Base case $\varphi = T$:

We have $regr(\top, eff(o)) = \top$, and $s \models \top$ iff $s[o] \models \top$ is correct.

Regressing Formulas Through Effects: Correctness (2)

Base case $\varphi = \bot$:

We have $regr(\bot, eff(o)) = \bot$, and $s \models \bot$ iff $s[o] \models \bot$ is correct.

Base case $\varphi = v$:

We have $s \models regr(v, eff(o))$ iff $s[o] \models v$ from the previous lemma.

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B3. General Regression, Part I

Regressing Formulas Through Effects

Regressing Formulas Through Effects: Correctness (3)

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Proof (continued).
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Inductive case $\varphi = \neg \psi$:

$$s \models regr(\neg \psi, eff(o)) \text{ iff } s \models \neg regr(\psi, eff(o))$$

 $\text{iff } s \not\models regr(\psi, eff(o))$
 $\text{iff } s \llbracket o \rrbracket \not\models \psi$
 $\text{iff } s \llbracket o \rrbracket \models \neg \psi$

Inductive case $\varphi = \psi \vee \chi$:

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s \models regr(\psi \lor \chi, eff(o)) \text{ iff } s \models regr(\psi, eff(o)) \lor regr(\chi, eff(o))
                                   iff s \models regr(\psi, eff(o)) or s \models regr(\chi, eff(o))
                                   iff s[o] \models \psi or s[o] \models \chi
                                   iff s[o] \models \psi \lor \chi
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Inductive case $\varphi = \psi \wedge \chi$:

Like previous case, replacing "∨" by "∧" and replacing "or" by "and".

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B3.3 Summary

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B3. General Regression, Part I

Summary

- ▶ Regressing a state variable through an (arbitrary) operator must consider two cases:

 - state variables made true (by add effects)
 state variables remaining true (by absence of delete effects)
- ▶ Regression of state variables can be generalized to arbitrary formulas φ by replacing each occurrence of a state variable in φ by its regression.

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