

Computing Programs, Controllers and Grammars with classical planning

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2017

Introduction



Figure 1: Sorting a specific list of numbers.

Introduction

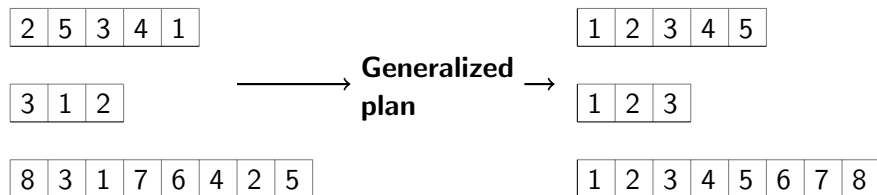


Figure 2: Sorting different lists of numbers with variable size.

Introduction

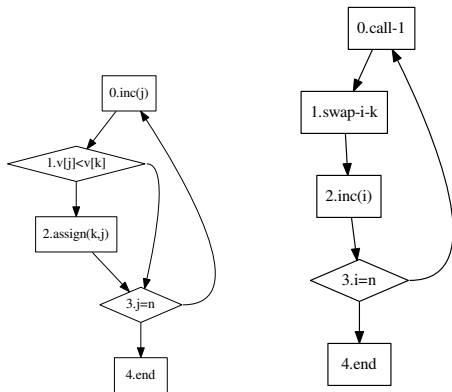


Figure 3: Generalized plan that corresponds to the *select-sort* algorithm (auxiliary procedure left, and main procedure, right).

Introduction

Background

- Classical planning with conditional effects
- Generalized planning

Method

- Representation
- Computation

Classical planning

- ▶ A **classical planning task** is a tuple $P = \langle F, A, I, G \rangle$
 - F set of fluents (propositional state variables)
 - A set of actions
 - I initial state
 - G goal condition
- ▶ A **classical plan** $\pi = \langle a_1, \dots, a_n \rangle$ *solves* P iff its application starting from I satisfies G

Classical planning with conditional effects

► **Actions** $a \in A$

- (1) **Preconditions**, $\text{pre}(a)$. *Applicable* in state s iff $\text{pre}(a) \subseteq s$
- (2) **Conditional effects**, $C \triangleright E \in \text{cond}(a)$. Two sets of literals: the *condition* C and the *effect* E
 - *Triggered effects*, depend on the state s where a is applied

$$\text{eff}(s, a) = \bigcup_{C \triangleright E \in \text{cond}(a), C \subseteq s} E,$$

- *Resulting state*, computed applying the triggered effects

$$\theta(s, a) = (s \setminus \neg\text{eff}(s, a)) \cup \text{eff}(s, a)$$

Generalized planning

A **generalized planning task** is a set of individual classical planning tasks with conditional effects $\mathcal{P} = \{P_1, \dots, P_\tau\}$

- ▶ Each $P_t = \langle F, A, I_t, G_t \rangle$, $1 \leq t \leq \tau$
 - ▶ Share the same planning frame $\Phi = \langle F, A \rangle$
 - ▶ **Differ** only in the **initial state** and **goals**
- ▶ A generalized plan Π solves \mathcal{P} iff **solves every** $P_t \in \mathcal{P}$

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Representing generalized plan as programs

Given planning frame $\Phi = \langle F, A \rangle$, a **program** Π is a **numbered list of instructions**. Each *instruction* is either:

- ▶ An action $a \in A$
- ▶ **Conditional goto**, $goto(i', !f)$
 - i' is the target program line
 - $f \in F$ is the jump condition
- ▶ **Termination**, marking the end of the program

Representing generalized plan as programs

0. dec(x)
1. goto(0,!(x=0))
2. dec(y)
3. goto(2,!(y=0))
4. end

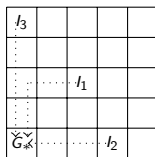


Figure 4: Generalized plan for navigating to cell (0,0) in a grid.

Representing generalized plan as programs

Execution of program Π on planning problem $P = \langle F, A, I, G \rangle$:

1. Set $s = I$ and $pc = 0$ (*current state* and *program counter*)
2. Update s and pc with instruction w on line pc :
 - 2.1 If $w \in A$ then: $s = \theta(s, w)$ and $pc = pc + 1$
 - 2.2 If $w = goto(i', !f)$ then: $pc = i'$ if $\neg f \in s$ else $pc = pc + 1$
 - 2.3 If w is a termination instruction execution **ends**

Representing generalized plan as programs

1. Π **solves** P , iff Π ends and G holds in s .
2. Π **solves** $\mathcal{P} = \{P_1, \dots, P_\tau\}$, iff Π solves every $P_t \in \mathcal{P}$
3. Π **fails** on P if:
 - 3.1 Execution ends but goals do not hold
 - 3.2 The precondition of an instruction $a \in A$ does not hold
 - 3.3 Execution enters an infinite loop

A classical planner can detect these success/failure conditions

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Computing programs with classical planning

Compiling classical planning problem $P = \langle F, A, I, G \rangle$ into another classical planning problem $P_n = \langle F_n, A_n, I_n, G_n \rangle$:

n bounds the **max** number of **program lines**

F_n extends F with fluents for coding

- **program counter**, $\{pc = 0, \dots, pc = n\}$
- **program lines**, $\{ins_{i,w}\}$ where $0 \leq i \leq n$,
 $w \in A \cup goto(i', !f) \cup \{end, nil\}$

A_n replaces A with actions for **Programming** and **Executing** instruction w at program line i

- $Program_{w_i}$ and $Execute_{w_i}$

I_n extends I with $pc = 0$ and $\{ins_{i,nil}\}$ (**empty program lines**)

G_n extends G with fluent $done$ added by $Execute_{end_i}$

Computing programs with classical planning

```
*0. nil          0. dec(x)
  1. nil          1. goto(0,!(x=0))
  2. nil          2. dec(y)
  3. nil          3. goto(2,!(y=0))
  4. end          4. end
```

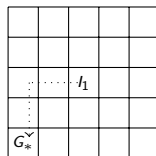


Figure 5: Empty program and generalized plan for navigating to cell (0,0) in a grid.

Computing programs with classical planning

Compiling generalized planning problem $\mathcal{P} = \{P_1, \dots, P_\tau\}$ into classical planning problem $P_n = \langle F_n, A_n, I_n, G_n \rangle$:

F_n includes $\{\text{test}_t\}_{1 \leq t \leq \tau}$ to indicate the current problem

A_n includes actions $\text{Execute}_{\text{end}_i, t}$ applicable when $G_t \subseteq s$

1. Reset the program counter, $pc = 0$
2. Reset the state $s = I_{t+1}$ (if $t < \tau$) or adds *done* (if $t = \tau$)

I_n extends I_1 with $pc = 0$ and $\{\text{ins}_{i, \text{nil}}\}$

Computing programs with classical planning

```
*0. nil          0. dec(x)
 1. nil          1. goto(0,!(x=0))
 2. nil          2. dec(y)
 3. nil          3. goto(2,!(y=0))
 4. end          4. end
```

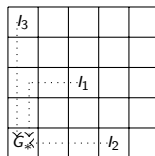


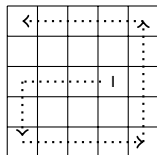
Figure 6: Empty program and generalized plan for navigating to cell (0,0) in a grid.

Computing programs with classical planning

Limitations of the approach

- ▶ May require *informative* examples to generalize
- ▶ May require high-level state features [Lotinac et al., 2016]
- ▶ May require too long programs [Segovia-Aguas et al., 2016a]

Generalized plans with a call stack



```
main:  0. call( $\Pi_1$ )
       1. call( $\Pi_2$ )
       2. call( $\Pi_3$ )
       3. call( $\Pi_4$ )
       4. end
```

```
 $\Pi_2$ :  0. inc(x)
       1. goto(0,!(x=n))
       2. end
```

```
 $\Pi_3$ :  0. inc(y)
       1. goto(0,!(y=n))
       2. end
```

```
 $\Pi_4$ :  0. dec(x)
       1. goto(0,!(x=0))
       2. end
```

(a)

(b)

Figure 7: Program for visiting the corners of a $n \times n$ grid using four auxiliary procedures (Π_1 as defined by the program in Figure 4).

Generalized plans with a call stack

We recursively define a **call stack** $\Omega \oplus (j, i)$:

- ▶ Ω is a call stack with finite size $|\Omega| = \ell$
- ▶ (j, i) is the element at the top of the stack
 - ▶ j is the **procedure** Π_j , $0 \leq j \leq m$
 - ▶ i is the **line**, $0 \leq i \leq |\Pi_j|$

Generalized plans with a call stack

Execution of program Π on planning problem $P = \langle F, A, I, G \rangle$:

1. Set $s = I$ and $\Omega = (0, 0)$ (*current state and stack*)
2. Update s and $\Omega \oplus (j, i)$ with instruction w :
 - 2.1 If $w \in A$, then $s = \theta(s, w)$ and $\Omega = \Omega \oplus (j, i + 1)$
 - 2.2 If $w = goto(i', !f)$ then: $\Omega = \Omega \oplus (j, i')$ if $\neg f \in s$ else $\Omega = \Omega \oplus (j, i + 1)$
 - 2.3 If $w = call(j')$, $\Omega = \Omega \oplus (j, i + 1) \oplus (j', 0)$ (recursive call when $j = j'$).
 - 2.4 If w is an end instruction pops the top element from the stack. Execution terminates if the stack becomes empty

Some experimental results

Domain	Procs	Solution	Lines	Instances	Time(s)	Total t.(s)	Plan length
Blocks	2	NP	4,3	5,5	2,2	4	46,61
Blocks	1	OP	6	5	85	85	73
Fibonacci	2	NP	3,3	2,5	2,177	179	12,129
Fibonacci	1	OP	5	2	3570	3570	56
Grid	3	NP	5,5,2	2,2,4	611,631,2	1244	43,43,213
Grid	1	ME	10	5	-	-	-
Gripper	3	NP	3,3,3	2,2,2	1,1,2	4	12,12,54
Gripper	1	OP	4	2	1	1	77
Hall-A	5	NP	5,5,5,3,5	2,2,2,2,2	3,7,3,1,4	18	44,40,40,73,155
Hall-A	1	ME	14	2	-	-	-
List	1	R	5	6	5	5	120
Reverse	1	OP	4	2	22	22	38
Sorting	2	NP	4,4	4,3	14,16	30	73,188
Sorting	1	ME	7	4	-	-	-
Summatory	1	OP	3	2	1	1	26
Tree/DFS	1	RP	6	1	165	165	51
Visitall	3	NP	4,2,4	2,2,2	1,1,5	7	47,18,238
Visitall	1	ME	7	2	-	-	-
Visual-Marker	3	NP	4,2,4	4,2,5	2,1,10	13	82,18,205
Visual-Marker	1	ME	8	2	-	-	-

Table 1: Procedures (1: only main; >1: DCK as auxiliary procedures were used), solution kind (NP=Nested Procedures, OP=One Procedure, R=Recursivity, RP=Recursivity with Parameters, ME=Memory error) and per each procedure: program lines, instances used to generate each procedure, planning time and plan length.

Hierarchical Finite State Machines

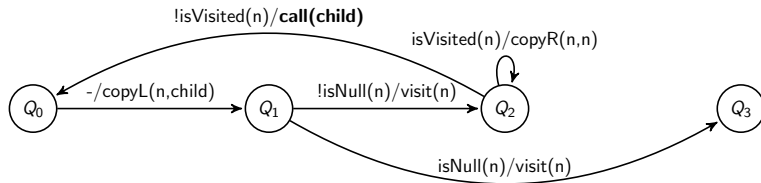


Figure 8: Hierarchical FSC $C[n]$ that traverses a binary tree. The lone parameter of the controller $[n]$ represents the current tree node.

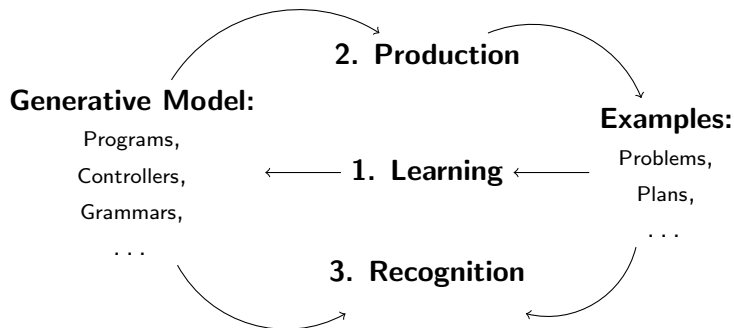
Context Free Grammars

	Π_0 : 0. choose(1 5 8)
S \rightarrow aSa	1. parse_a
S \rightarrow bSb	2. call(Π_0)
S \rightarrow ϵ	3. parse_a
	4. end
	5. parse_b
	6. call(Π_0)
	7. parse_b
	8. end

Figure 9: CFG and the corresponding planning program.

Conclusions

- ▶ **Learning** different **generative models** from **small** sets of examples with **classical planning**



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