

PROGRAM SYNTHESIS

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OUTLINE

- The multiple benefits of Domain Specific Languages

"a program has structure, it is not a raw piece of text"

- The need for proper interfaces

"LLMs + fact-checking = future"

- The TRUE Programmatic Reinforcement Learning

"Program you are, and to program you will return"

environment

policy

BACK IN 2011: FLASHFILL (Microsoft)

	A	B	C	D	E	F
1	Attendees	Office Address	Hi _____	E-mail	Zip Code	
2	Gerald Parker	320 N Morgan St #600, Chicago, IL 60607	Hi Gerald			
3	Phillip Davis	111 8th Ave. New York, NY 10011				
4	Ralph Gonzales	2930 Pearl Street, Boulder, CO. 80301				
5	Shawn Jenkins	1160 Bordeaux Drive, Sunnyvale, CA 94089				
6	Thomas Murphy	1160 Bordeaux Drive, Sunnyvale, CA 94089				
7	Norma Long	2930 Pearl Street, Boulder, CO. 80301				
8	Beverly Harris	320 N Morgan St #600, Chicago, IL 60607				
9	Melissa Torres	1160 Bordeaux Drive, Sunnyvale, CA 94089				
10	Joshua Watson	111 8th Ave. New York, NY 10011				
11	Larry Wright	500 W 2nd St. Austin, TX 78701				
12	Lois Brooks	2930 Pearl Street, Boulder, CO. 80301				
13	Steve James	1160 Bordeaux Drive, Sunnyvale, CA 94089				
14	Helen Rivera	2930 Pearl Street, Boulder, CO. 80301				
15	Samuel Bailey	111 8th Ave. New York, NY 10011				
16	Kenneth Gonzalez	2930 Pearl Street, Boulder, CO. 80301				
17	Sharon Washington	1160 Bordeaux Drive, Sunnyvale, CA 94089				
18	Mildred Evans	1160 Bordeaux Drive, Sunnyvale, CA 94089				
19						
20						
21						
22						

(this picture is actually Smartfill, Google's version)

Limitations:

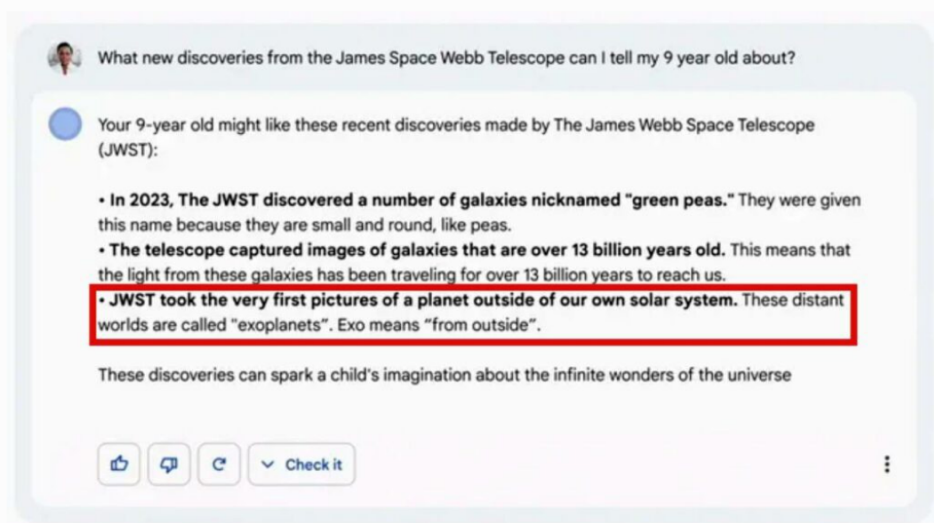
- it doesn't know that Paris is in France!
- it doesn't output the program so we cannot trust the results

BACK IN 2021: CoDEX (Open AI / Github)

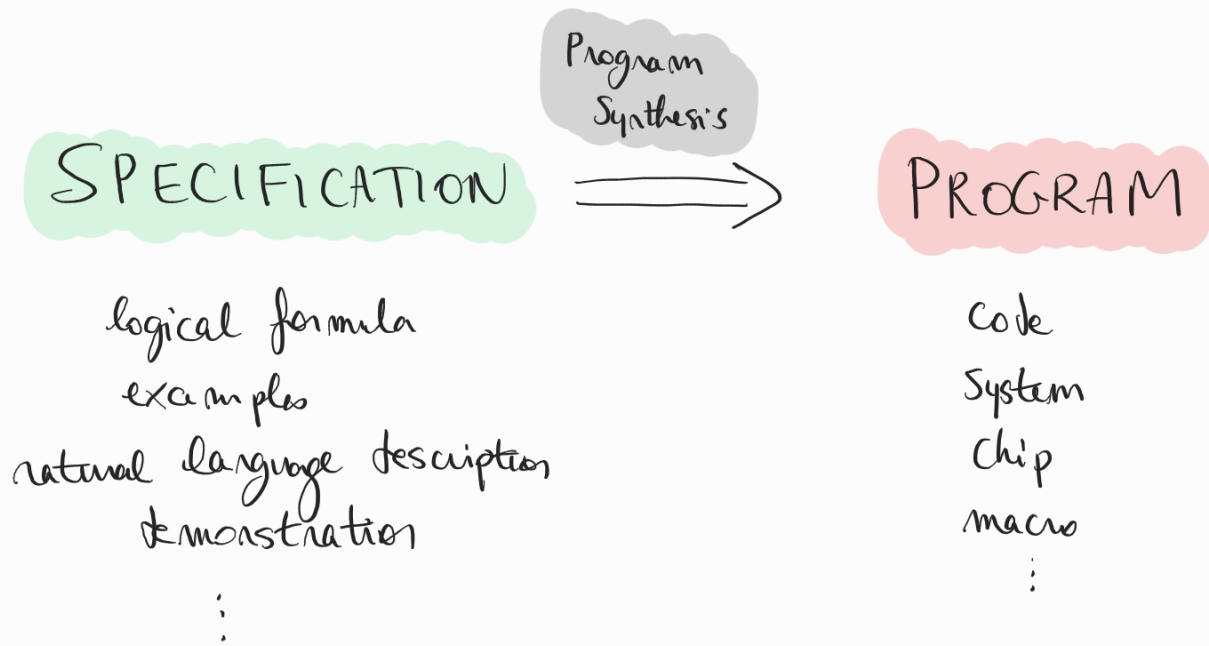
```
1 import datetime
2
3 def parse_expenses(expenses_string):
4     """Parse the list of expenses and return the list of triples (date, value, currency).
5     Ignore lines starting with #.
6     Parse the date using datetime.
7     Example expenses_string:
8         2016-01-02 -34.01 USD
9         2016-01-03 2.59 DKK
10        2016-01-03 -2.72 EUR
11    """
12    expenses = []
13    for line in expenses_string.splitlines():
14        if line.startswith("#"):
15            continue
16        date, value, currency = line.split(" ")
17        expenses.append((datetime.datetime.strptime(date, "%Y-%m-%d"),
18                        float(value),
19                        currency))
20    return expenses
```

Limitations: treats code as raw text: a lot of syntactic errors!

A FEW MONTHS AGO: BARD (Google)



Limitations: it doesn't know, it learns (from biased data)



- One of the oldest dream of AI
- For decades only for delusional researchers
- The upcoming revolution for developers
| end users ?

RESEARCH PERSPECTIVE ON SYNTHESIS

- = **HUGE** practical success and impact
- theory is lagging behind but **highly needed**
- at the intersection of
 - * Programming languages
 - * Software Engineering
 - * Machine learning
 - * Algorithms

PART I

SYNTHESIS IN A DSL

All Large Language Models (LLM) tools I know
write code in Python / C / Java / ...

I believe this is **WRONG**: complicated syntax / semantics

program aliasing

no partial executions

hard to reason with

basis errors

no lifting to other languages

DSL: Domain Specific Language

- Designed for a specific class of tasks
 - Defined by a family of primitives
- ⇒ FUNCTIONAL PROGRAMMING

EXAMPLE DSL

list of primitives

sort	:	$\text{list}(\text{int}) \rightarrow \text{list}(\text{int})$
map	:	$(t_0 \rightarrow t_1) \rightarrow \text{list}(t_0) \rightarrow \text{list}(t_1)$
succ	:	$\text{int} \rightarrow \text{int}$
	:	$\vdots$

their types

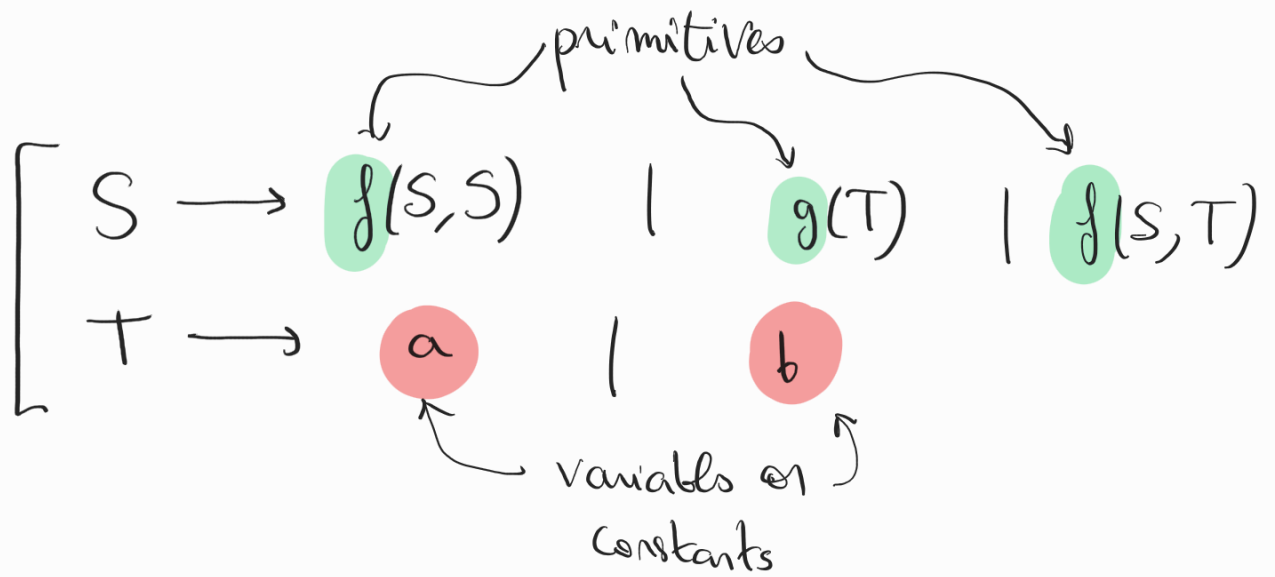
TYPES

atomic types: int , bool , String , $\text{list}(\text{int})$, ...

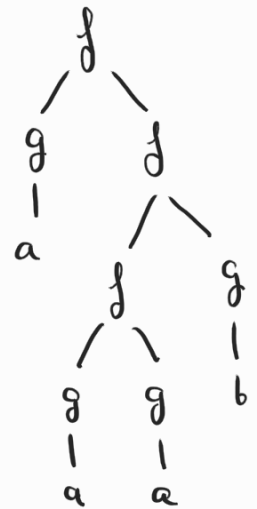
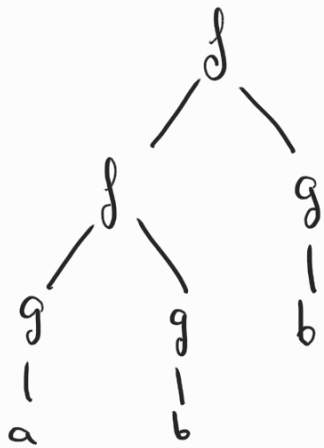
generic types: t_0 , t_1 , ...

constructor: $z_1 \rightarrow z_2$: functions

CONTEXT FREE GRAMMARS

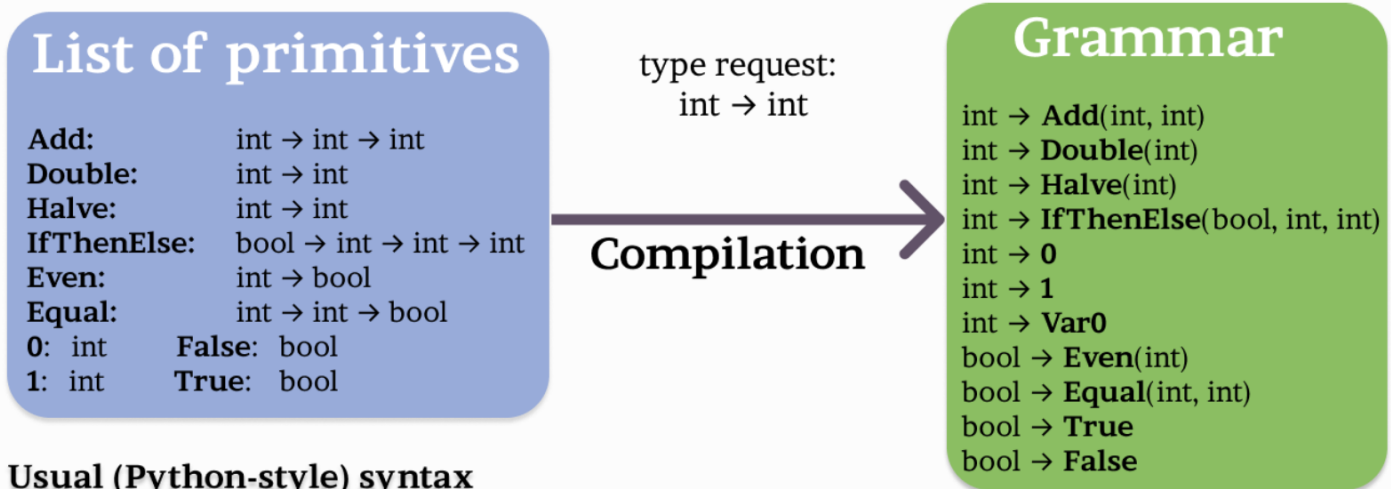


Example trees generated by CFG:



$\mathcal{L}(G)$: set of trees generated by the CFG G

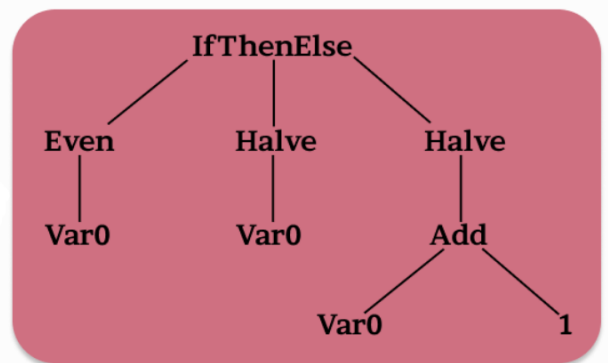
SOME EXAMPLES OF REASONING



Usual (Python-style) syntax

```
if Even(var0):
    Halve(var0)
else:
    Halve(Add var0 1)
```

Equivalent AST representation



TYPE-BASED COMPILATION

list of primitive

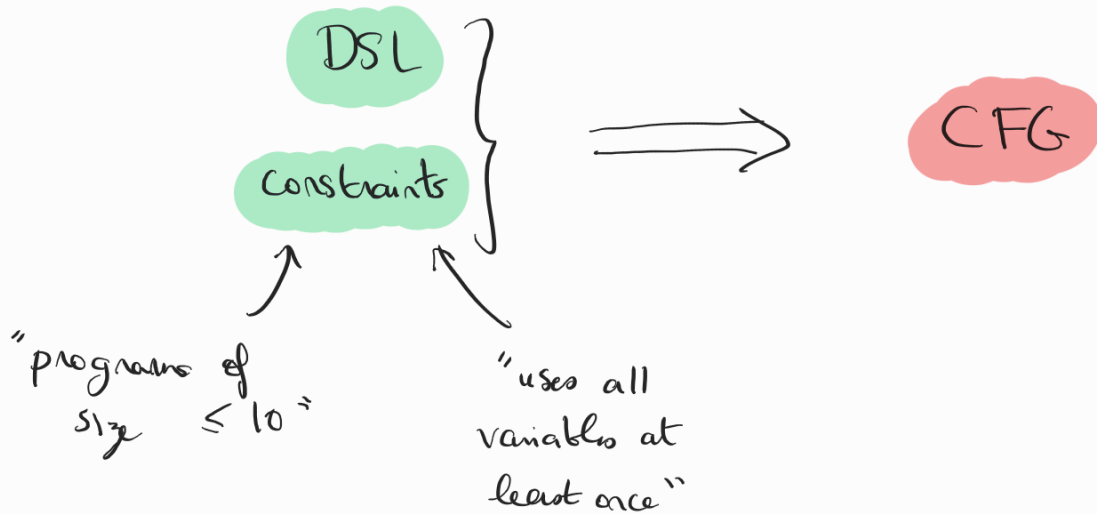
$$\left\{ \begin{array}{l} \text{sort} : \text{list}(\text{int}) \rightarrow \text{list}(\text{int}) \\ \text{map} : (t_0 \rightarrow t_1) \rightarrow \text{list}(t_0) \rightarrow \text{list}(t_1) \\ \text{succ} : \text{int} \rightarrow \text{int} \\ \vdots \end{array} \right.$$


CFG G

$$\left\{ \begin{array}{l} \text{list}(\text{int}) \xrightarrow{\text{red}} \text{sort}(\text{list}(\text{int})) \\ \text{list}(\text{int}) \xrightarrow{\text{red}} \text{map}(\text{int} \rightarrow \text{int}, \text{list}(\text{int})) \\ \text{list}(\text{int}) \xrightarrow{\text{red}} \text{input} \\ \text{int} \rightarrow \text{int} \xrightarrow{\text{red}} \text{succ}() \\ \vdots \end{array} \right.$$

$\mathcal{L}(G) = \text{all correctly typed programs}$

ADVANCED COMPILATION



RECENT WORK: SEMANTICAL PRUNING

(Théo Matricon)

automatically generating equations
to reduce the DSL

Example equations:

- $\text{Halve}(\text{Double}(P)) \equiv P$
- $\text{Add}(P, \text{Add}(Q, R)) \equiv \text{Add}(\text{Add}(P, Q), R)$
- $\text{Add}(P, Q) \equiv \text{Add}(Q, P)$

Objective: construct a CFG representing programs
up to equivalence

Unfortunately that's **impossible**: equivalence of programs
is very quickly undecidable

Remark: $\text{Add}(P, Q) \equiv \text{Add}(Q, P)$ is already too hard

But: we can quotient by "regular equivalence"

Key ingredient: well order on programs

PART II

KNOWLEDGE MEETS SYNTHESIS

Smartfill doesn't know that Paris is in France

Band doesn't know what James Webb Telescope does

ChatGPT cannot add numbers

BUT

Wikipedia and/or Calculators do !

So far, synthesis was either:

- purely syntactic (FlashFill)
- or looks for semantics in training data

RESEARCH QUESTION

How to use external knowledge in synthesis?

RECENT WORK: WIKICODER (Théo Matignon)

a program synthesis tool which queries Wikipedia when it's lost

BUT: Wikipedia's structure is hard to use

A LOT to be done

PART III

PROGRAMMATIC RL

Experiment on MountainCar

In ~20s we found:

input variables (position, velocity)

```
If ( sign(var1) ≥ var0 ) :  
    left  
Else:  
    Right
```

actions

a program

Baseline Deep learning (DQN): 5 mn for a neural policy ...

QUESTIONS

Q1 what is a good DSL for policies?

Q2 how do we synthesise policies as programs?

So far;

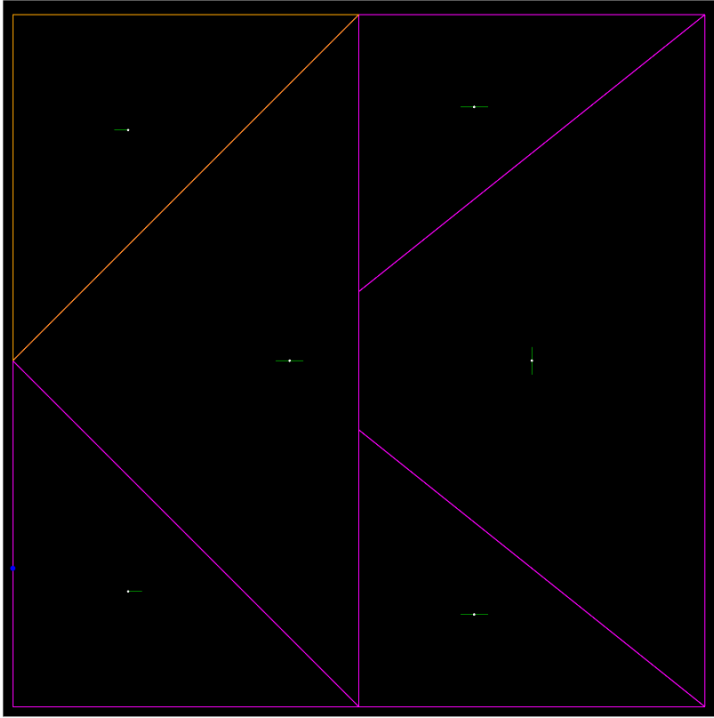
most answers to Q1: decision trees or finite state controllers

most answers to Q2: extract from a neural policy

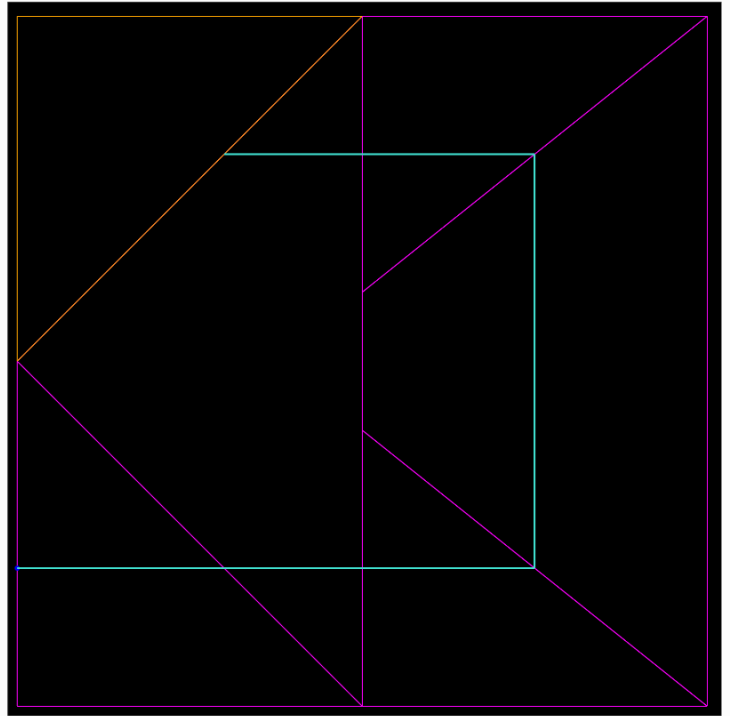
GRIDWORLDS

(Guruprerana Shabadi)

problem

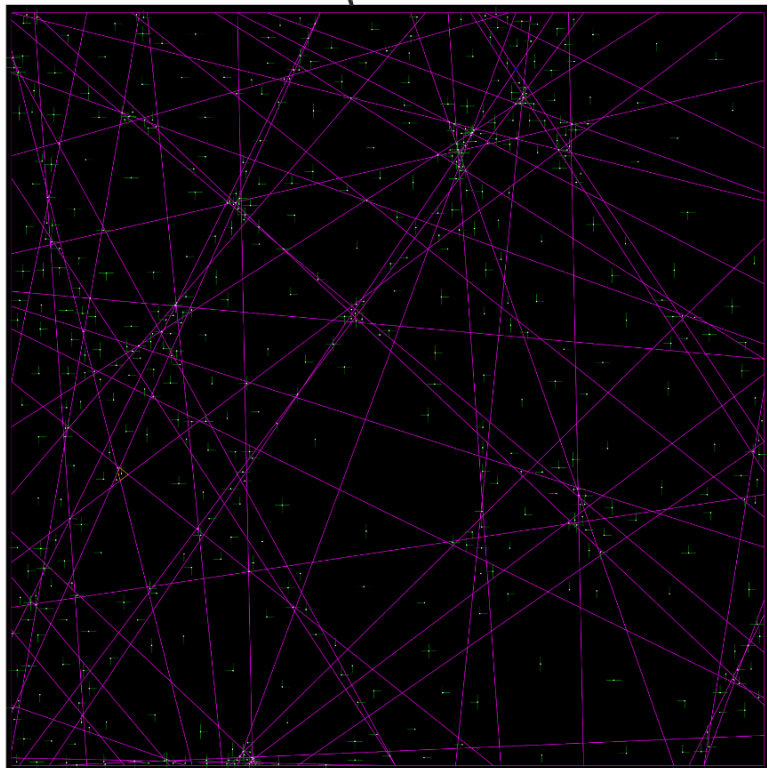


Solution

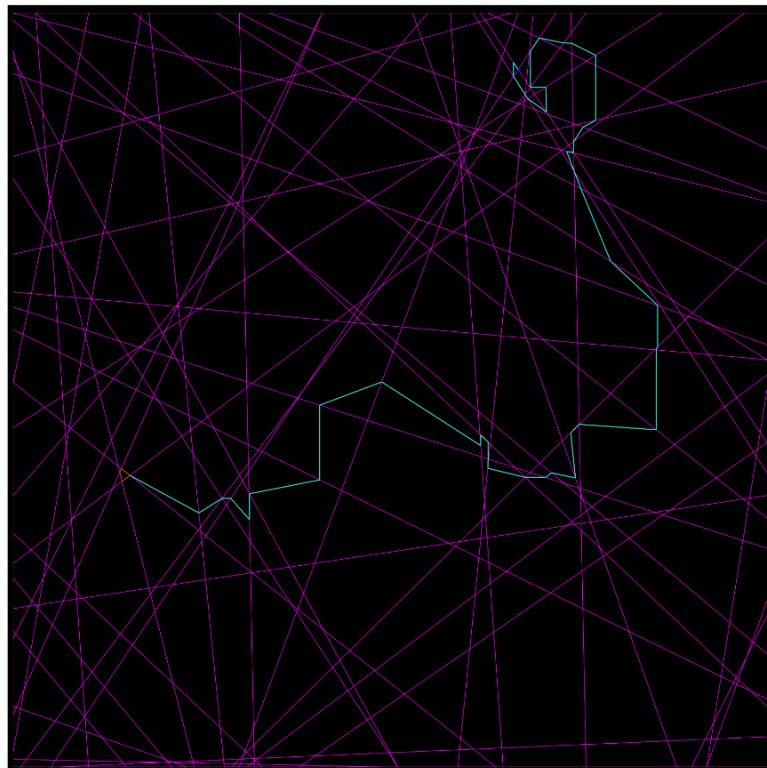


Key feature: the environment is generated as a PRBS program

problem



Solution

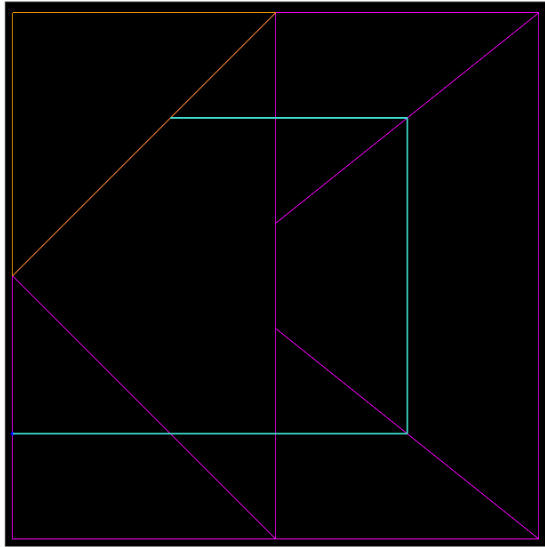


COMPARING TWO APPROACHES

(1) Solve as explicit MDP and return optimal policy as explicit function **states $\rightarrow$ actions**

(2) find a policy as program using backward analysis

Example



generated policy

```
Until([(0, 0), (0, 25)]):  
  [(0, 25), (0, 0)] ->  
  [(5, 20), (25, 0)], Preference: Neutral, Edge: [(0, 25), (25, 0)]  
  
Until([(0, 25), (25, 0)]):  
  [(0, 25), (25, 0)] ->  
  [(25, 0), (25, 20)], Preference: Neutral, Edge: [(25, 20), (25, 0)]  
  
Until([(25, 0), (25, 20)]):  
  [(25, 20), (25, 0)] ->  
  [(25, 20), (50, 0)], Preference: Neutral, Edge: [(25, 20), (50, 0)]  
  
Until([(25, 20), (50, 0)]):  
  [(25, 20), (50, 0)] ->  
  [(25, 30), (50, 50)], Preference: Neutral, Edge: [(25, 30), (50, 50)]  
  
Until([(25, 30), (50, 50)]):  
  [(25, 30), (50, 50)] ->  
  [(25, 30), (25, 50)], Preference: Neutral, Edge: [(25, 30), (25, 50)]  
  
Until([(25, 30), (25, 50)]):  
  [(25, 30), (25, 50)] ->  
  [(0, 25), (25, 50)], Preference: Neutral, Edge: [(0, 25), (25, 50)]  
  
Until([(0, 25), (25, 50)]):  
  
Win!
```

Summary of experimental results:

the program as policies are infinitely more succinct!

and **verifiable**, **explainable**, **readable**, **executable**, ...

environment as program : SMALL

environment as explicit MDP : LARGE

policy as program : SMALL

policy as explicit $S \rightarrow A$: LARGE

Motto:

Program you are, and to program you will return

environment

policy

JOIN US

We have open positions in Bordeaux!

PhD and postdoc

both theory and practice