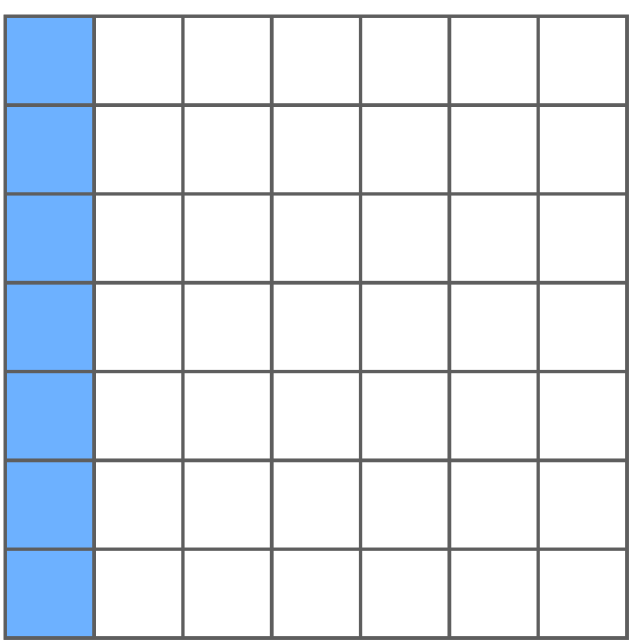


## Contribution

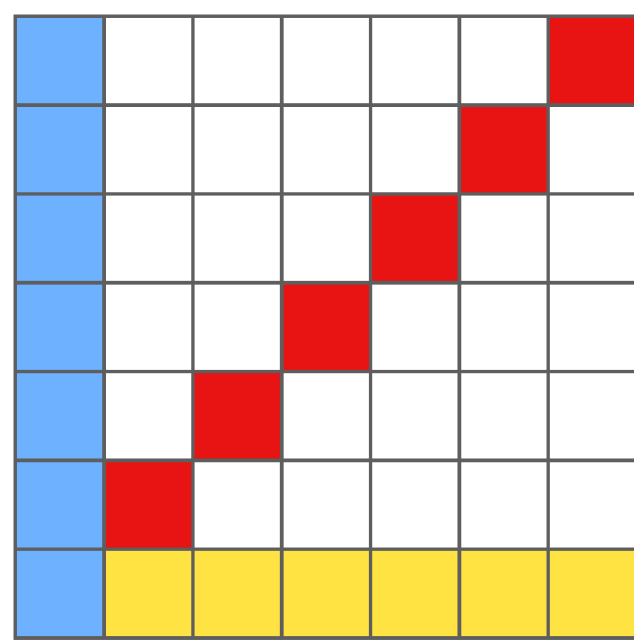
We show symbolic approaches can learn directly from raw data, such as individual pixels, by learning relations between raw elements.

## Visual reasoning

Input



Output



```
out(X,Y,C) :- in(X,Y,C).
out(X,Y,yellow) :- empty(X,Y), height(X).
out(X,Y,red) :- empty(X,Y), height(X+Y-1).
```

- ▶ An output pixel is colour  $C$  if it is colour  $C$  in the input.
- ▶ An output pixel is yellow if it is in the bottom row and empty in the input.
- ▶ An output pixel is red if it is empty in the input and the sum of its coordinates  $X$  and  $Y$  equals  $H + 1$ , i.e. it is on the diagonal.

## List functions

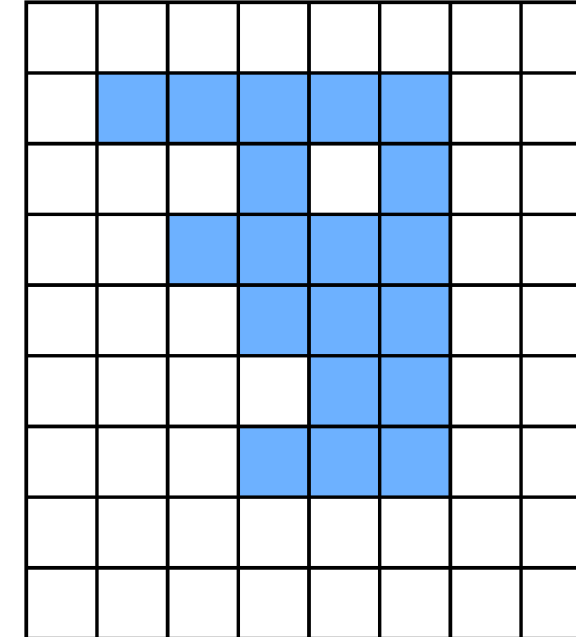
Input	Output
[81, 43]	[2, 43, 81, 2]
[1, 63, 21, 16]	[4, 16, 21, 63, 1, 4]

```
out(1,E-1):- end(E).
out(I,V):- end(E), in(I1,V), add(I,I1,E+1).
out(E+1,E-1):- end(E).
```

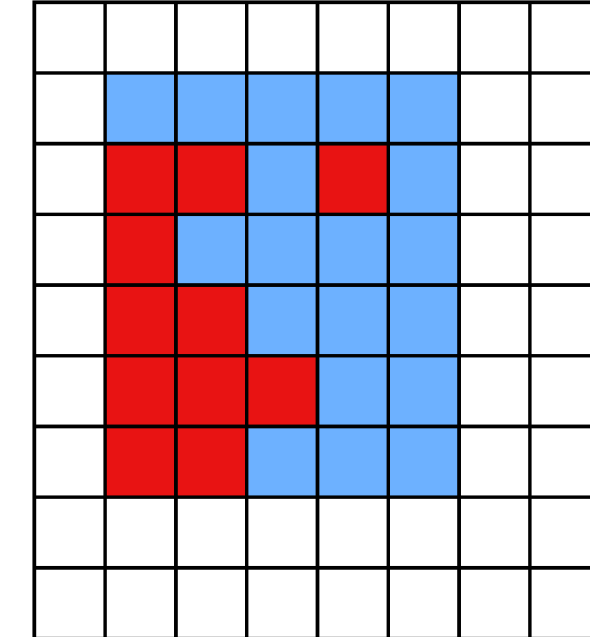
- ▶ The output element at index 1 is  $E - 1$ , i.e. the size of the input list.
- ▶ The output element at index  $I$  is the input element at index  $I1$ , where  $I + I1 = E + 1$ .
- ▶ The output element at index  $E + 1$  is  $E - 1$ .

## Visual reasoning

Input



Output

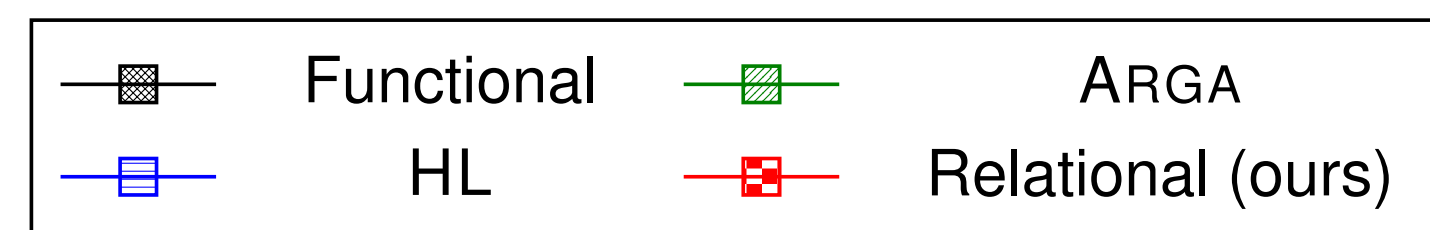


```
out(X,Y,C):- in(X,Y,C).
out(X,Y,red):- empty(X,Y), in(X1,Y,C), in(X,Y1,C).
```

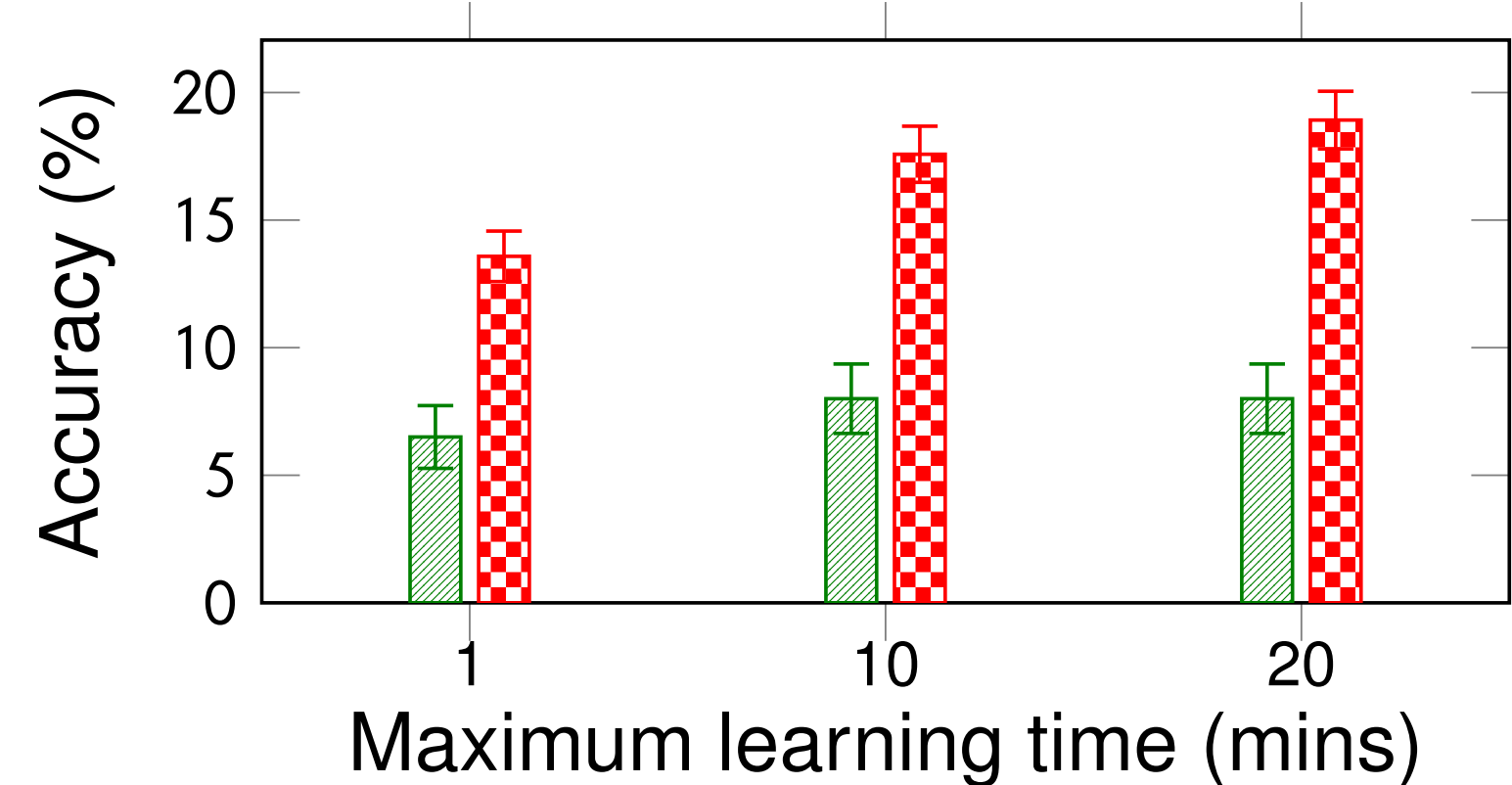
- ▶ An output pixel is colour  $C$  if it is colour  $C$  in the input.
- ▶ An output pixel is red if it is empty in the input and
  - ▶ an input pixel in the row  $X$  has colour  $C$ , and
  - ▶ an input pixel in the column  $Y$  has colour  $C$ .

## Evaluation

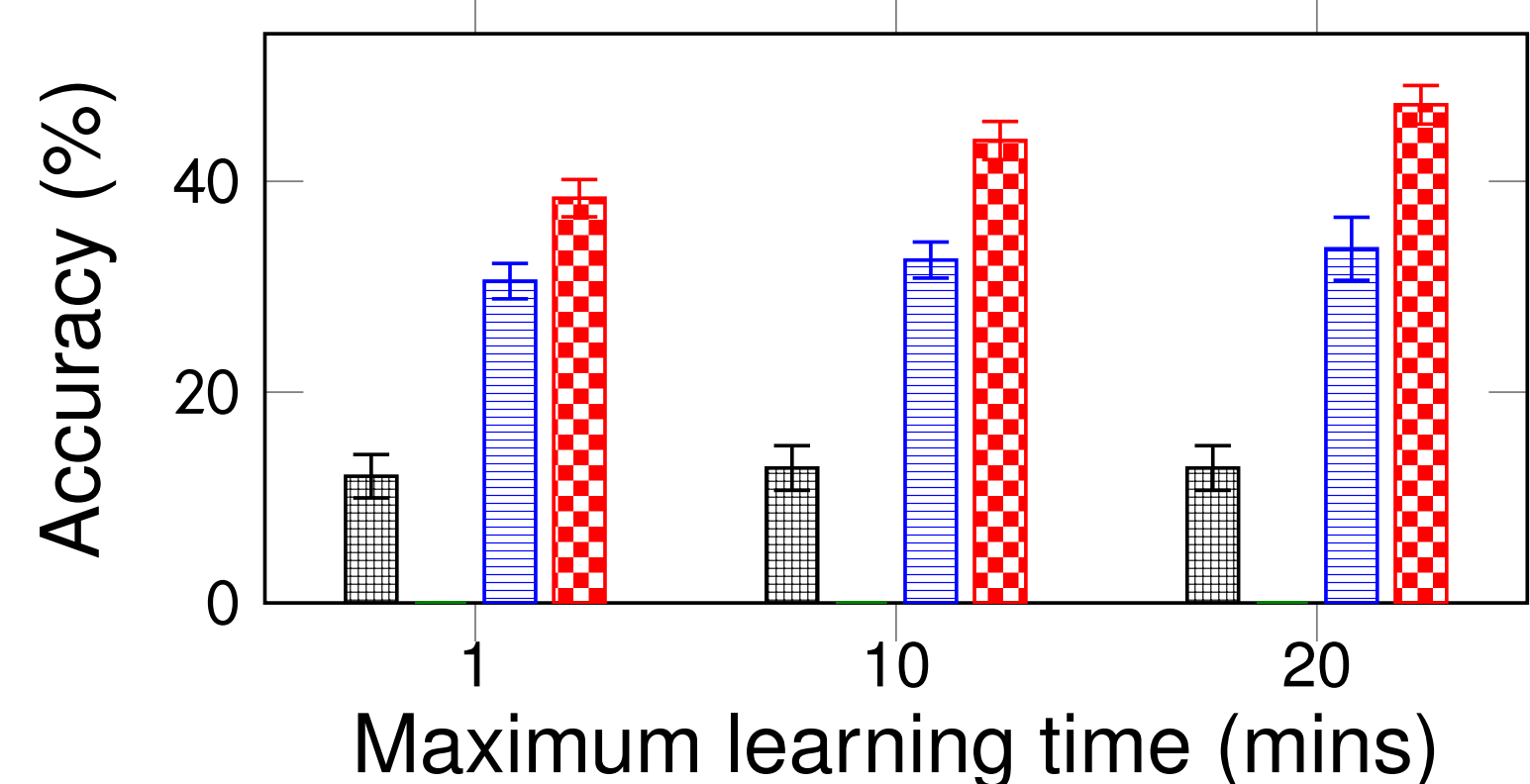
How does a general-purpose symbolic system with a relational representation compare against domain-specific approaches?



Visual reasoning



List functions



## Reference

Céline Hocquette and Andrew Cropper, *Relational decomposition for program synthesis*, 2024.