# Learning efficient logic programs

Input	Output
[s,h,e,e,p]	е
[a,1,p,a,c,a]	а
[c,h,i,c,k,e,n]	?

Input	Output
[s,h,e,e,p]	е
[a,1,p,a,c,a]	а
[c,h,i,c,k,e,n]	С

```
%% E+
f([s,h,e,e,p],e).
f([a,1,p,a,c,a],a).
f([c,h,i,c,k,e,n],c).
%% E-
f([s,h,e,e,p],p).
f([a,1,p,a,c,a],1).
%% BK
head([H|_],H).
tail([_|T],T).
length(A,B):- ...
element(A,B):- ...
mergesort(A,B):- ...
```

```
%% metagol
f(A,B):-head(A,B),tail(A,C),element(C,B).
f(A,B):-tail(A,C),f(C,B).
```

```
%% metagol
f(A,B):-head(A,B),tail(A,C),element(C,B).
f(A,B):-tail(A,C),f(C,B).
```

```
def f(xs):
    h = head(xs)
    t = tail(xs)
    if h is in t:
        return h
    return f(t)
```

# %% alternative f(A,B):-mergesort(A,C),f1(C,B). f1(A,B):-head(A,B),tail(A,C),head(C,B). f1(A,B):-tail(A,C),f1(C,B).

```
%% alternative
f(A,B):-mergesort(A,C),f1(C,B).
f1(A,B):-head(A,B),tail(A,C),head(C,B).
f1(A,B):-tail(A,C),f1(C,B).
          def f(xs):
              ys = sorted(xs)
              return f1(ys)
          def f1(xs):
              h = head(xs)
              t = tail(xs)
              hh = head(t)
              if h == hh:
                   return h
              return f1(t)
```

## Idea

# Input

- examples **E**
- background knowledge **B**
- **cost** : Program × Example → N

#### Idea

- 1. Learn any program H
- 2. Repeat while possible:
  - a. Learn program H' where max\_cost(H',E) < max\_cost(H,E)
  - b. **H=H'**
- 3. Return H

```
prove([],P,P).
prove([Atom|Atoms],P1,P2):-
    prove_aux(Atom, P1, P3),
    prove(Atoms, P3, P2).
prove_aux(Atom,P,P):-
    call(Atom).
prove_aux(Atom, P1, P2):-
    metarule(Atom, Body, Subs),
    save(Subs,P1,P3),
    prove(Body, P3, P2).
```

```
prove([],P,P).
prove([Atom|Atoms],P1,P2):-
    prove_aux(Atom, P1, P3),
    prove(Atoms, P3, P2).
prove_aux(Atom,P,P):-
    call(Atom).
prove_aux(Atom, P1, P2):-
    metarule(Atom, Body, Subs),
    save(Subs,P1,P3),
    prove(Body, P3, P2).
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prove([],P,P).
prove([Atom|Atoms],P1,P2):-
    prove_aux(Atom, P1, P3),
    prove(Atoms, P3, P2).
prove_aux(Atom,P,P):-
    call(Atom).
prove_aux(Atom, P1, P2):-
    metarule(Atom, Body, Subs),
    save(Subs,P1,P3),
    prove(Body, P3, P2).
```

```
prove([],P,P).
prove([Atom|Atoms],P1,P2):-
    prove_aux(Atom, P1, P3),
    prove(Atoms, P3, P2).
prove_aux(Atom,P,P):-
    call(Atom).
prove_aux(Atom, P1, P2):-
    metarule(Atom, Body, Subs),
    save(Subs, P1, P3),
    prove(Body, P3, P2).
```

#### Metaopt

```
prove([],P,P,C,C).
prove([Atom|Atoms],P1,P2,C1,C2):-
    prove_aux(Atom, P1, P3, C1, C3),
    prove(Atoms, P3, P2, C3, C2).
prove_aux(Atom, P, P, C1, C2):-
    pos_cost(Atom, Cost).
    C2 is C1+Cost,
    bound(MaxCost),
    C2 < MaxCost.
prove_aux(Atom, P1, P2, C1, C2):-
    metarule(Atom, Body, Subs),
    save(Subs,P1,P3),
    C3 is C1+1,
    prove(Body, P3, P2, C3, C2).
```

#### Metaopt

```
prove([],P,P,C,C).
prove([Atom|Atoms],P1,P2,C1,C2):-
    prove_aux(Atom, P1, P3, C1, C3),
    prove(Atoms, P3, P2, C3, C2).
prove_aux(Atom, P, P, C1, C2):-
    pos_cost(Atom, Cost).
    C2 is C1+Cost,
    bound(MaxCost),
    C2 < MaxCost.
prove_aux(Atom, P1, P2, C1, C2):-
    metarule(Atom, Body, Subs),
    save(Subs,P1,P3),
    C3 is C1+1,
    prove(Body, P3, P2, C3, C2).
```

#### Metaopt

```
prove([],P,P,C,C).
prove([Atom|Atoms],P1,P2,C1,C2):-
    prove_aux(Atom, P1, P3, C1, C3),
    prove(Atoms, P3, P2, C3, C2).
prove_aux(Atom, P, P, C1, C2):-
    pos_cost(Atom, Cost).
    C2 is C1+Cost,
    bound(MaxCost),
    C2 < MaxCost.
prove_aux(Atom, P1, P2, C1, C2):-
    metarule(Atom, Body, Subs),
    save(Subs, P1, P3),
    C3 is C1+1,
    prove(Body, P3, P2, C3, C2).
```

#### Iterative descent

- 1. Learn any program **H** with minimal clauses
- 2. Repeat while possible:
  - a. Learn program H' where max\_cost(H',E) < max\_cost(H,E)
  - b. **H=H'**
- 3. Return H

Metaopt prunes as it learns

Positive examples: size of the leftmost successful branch

Positive examples: size of the leftmost successful branch

```
pos_cost(Atom,Cost):-
    statistics(inferences,I1),
    call(Atom),
    statistics(inferences,I2),
    Cost is I2-I1.
```

Negative examples: size of the finitely-failed SLD-tree

Negative examples: size of the finitely-failed SLD-tree

```
neg_cost(Atom,Cost):-
    statistics(inferences,I1),
    \+ call(Atom),
    statistics(inferences,I2),
    Cost is I2-I1.
```

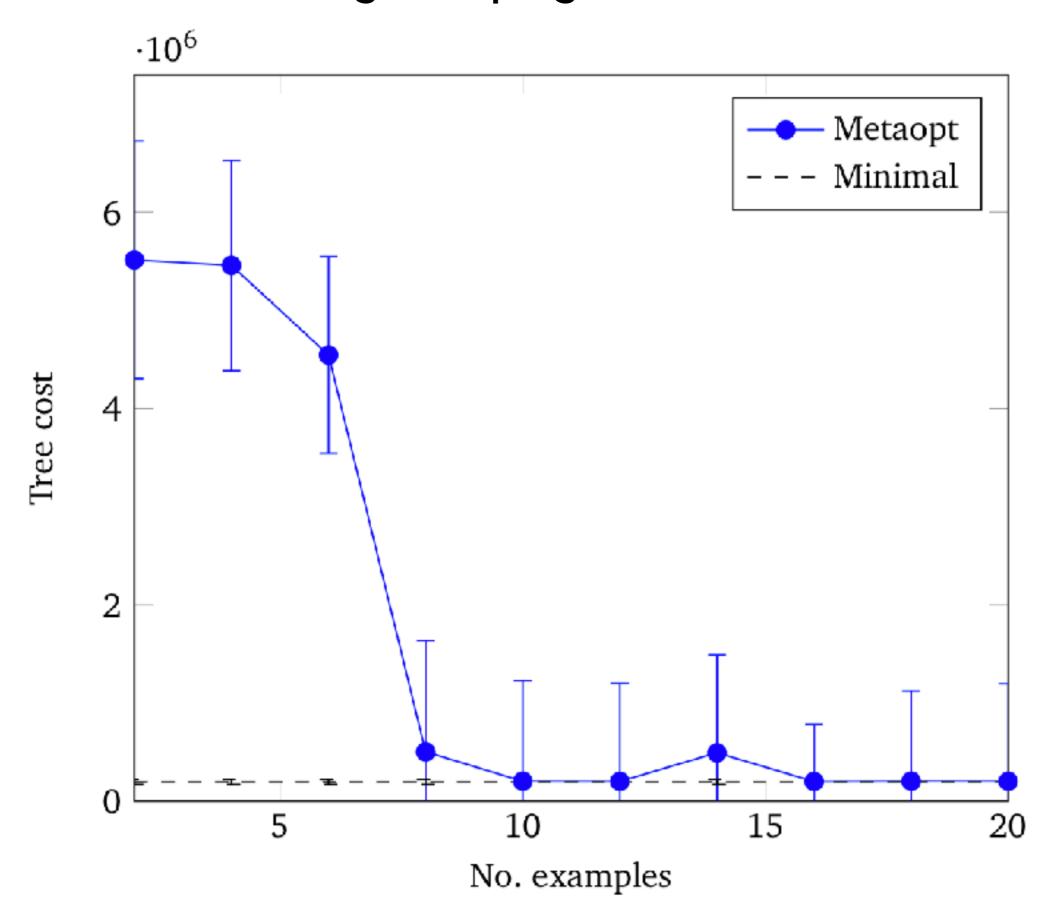
- any arity logics
- no user-supplied costs
- backtracking and non-determinism

Input	Output
[s,h,e,e,p]	е
[a,l,p,a,c,a]	a
[c,h,i,c,k,e,n]	С

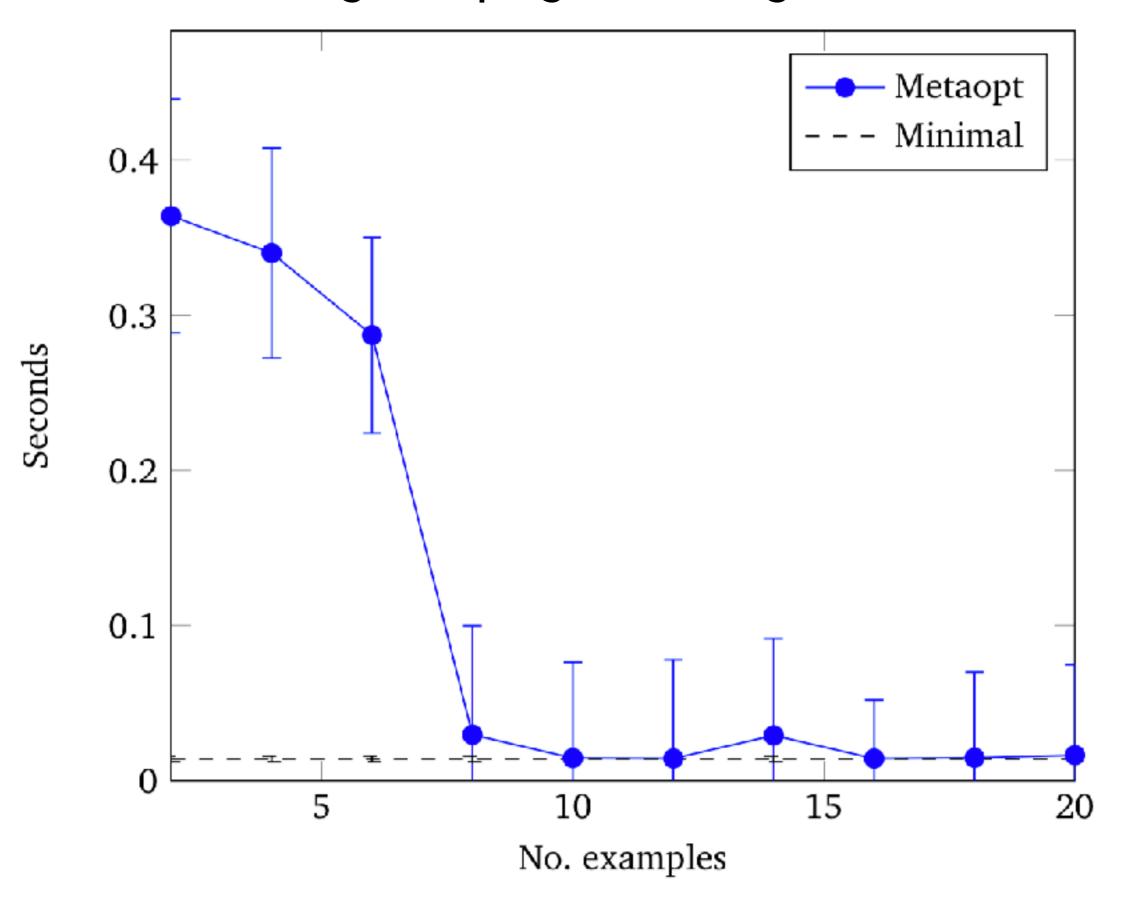
Input	Output
[s,h,e,e,p]	е
[a,l,p,a,c,a]	а
[c,h,i,c,k,e,n]	С

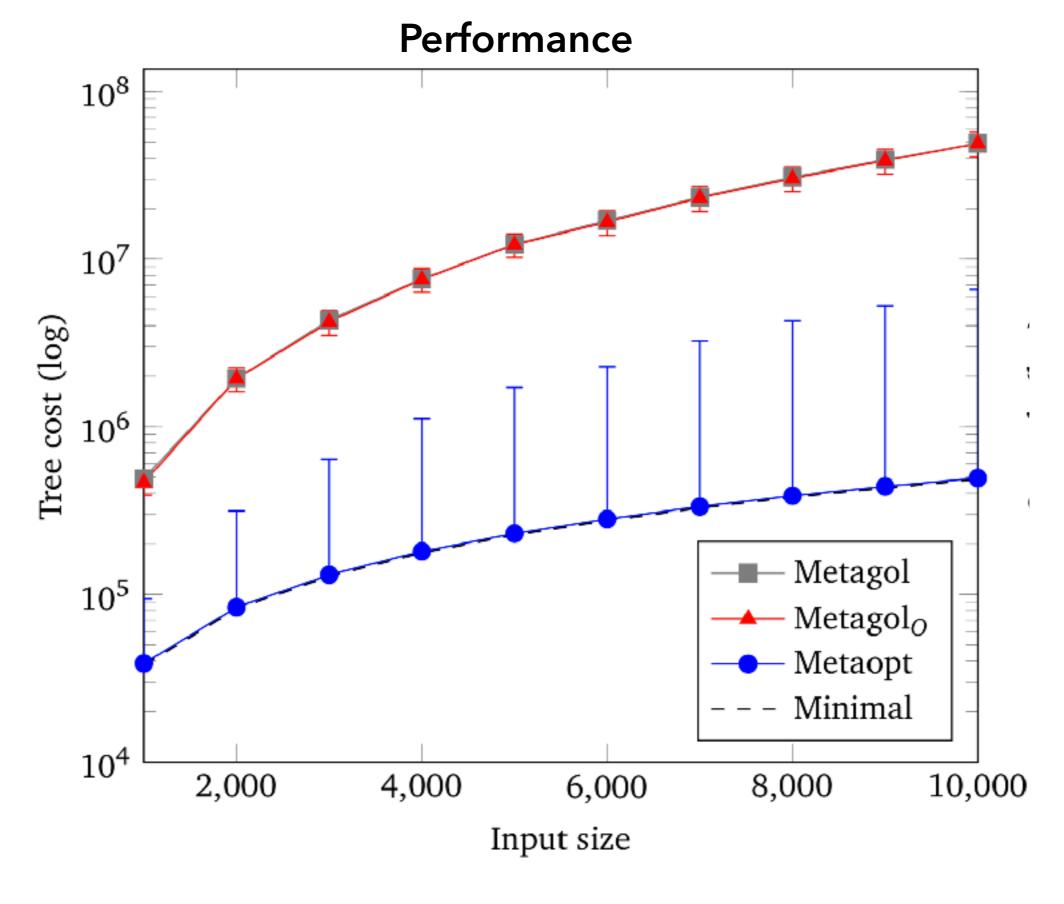
```
f(A,B):-mergesort(A,C),f1(C,B).
f1(A,B):-head(A,B),tail(A,C),head(C,B).
f1(A,B):-tail(A,C),f1(C,B).
```

# Convergence: program tree costs



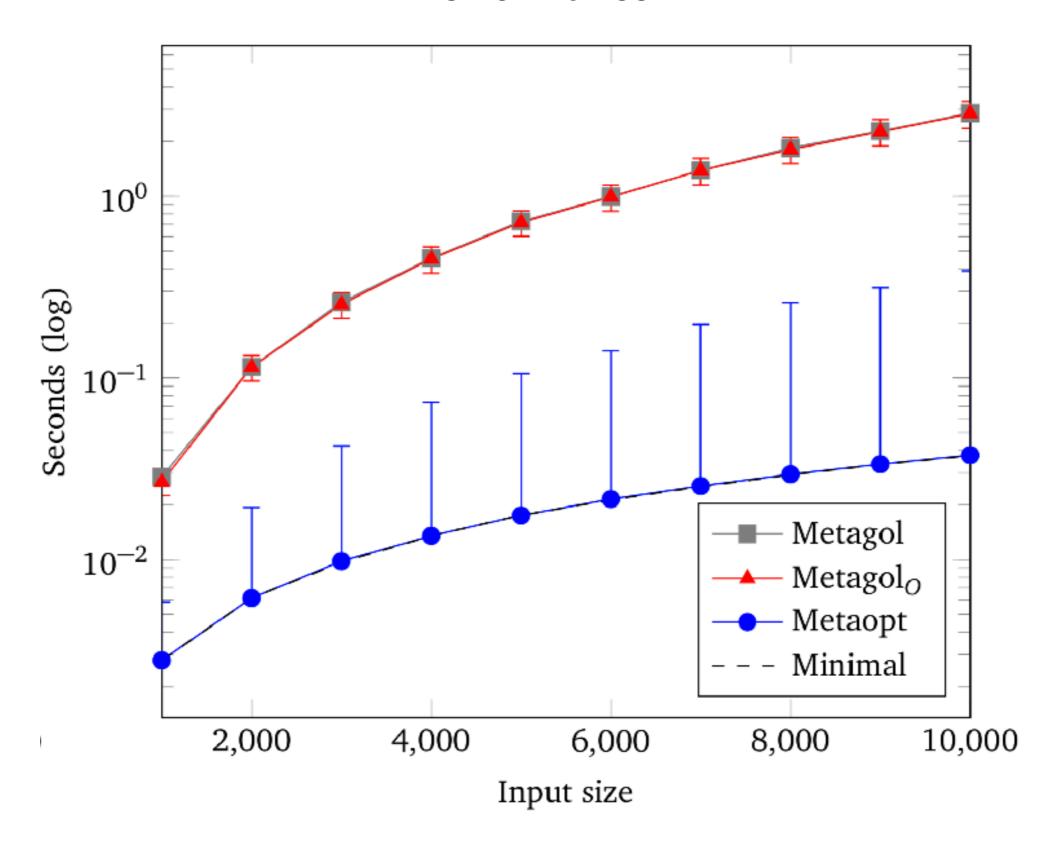
# Convergence: program running times





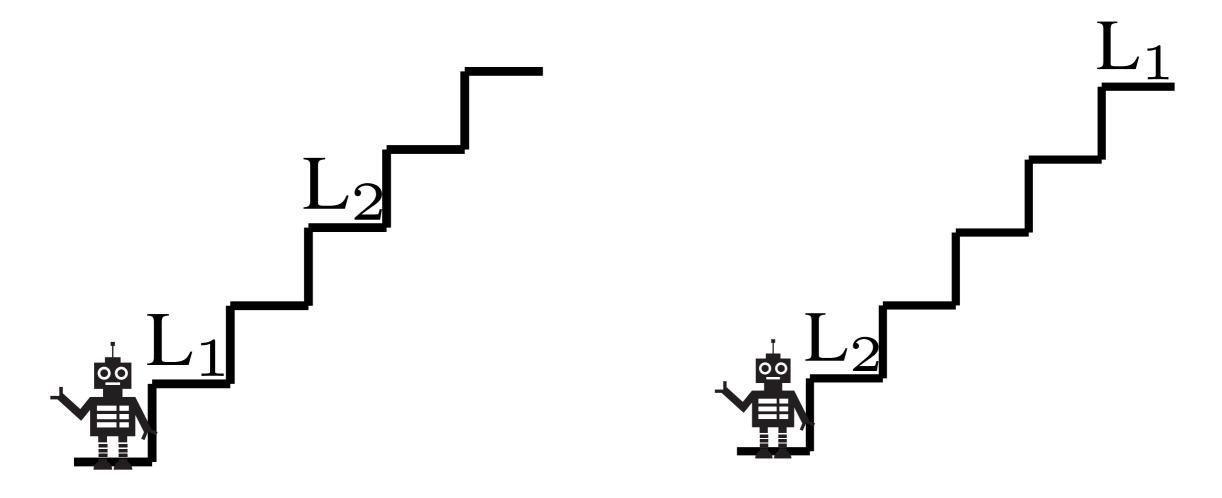
(a) Tree costs

## **Performance**



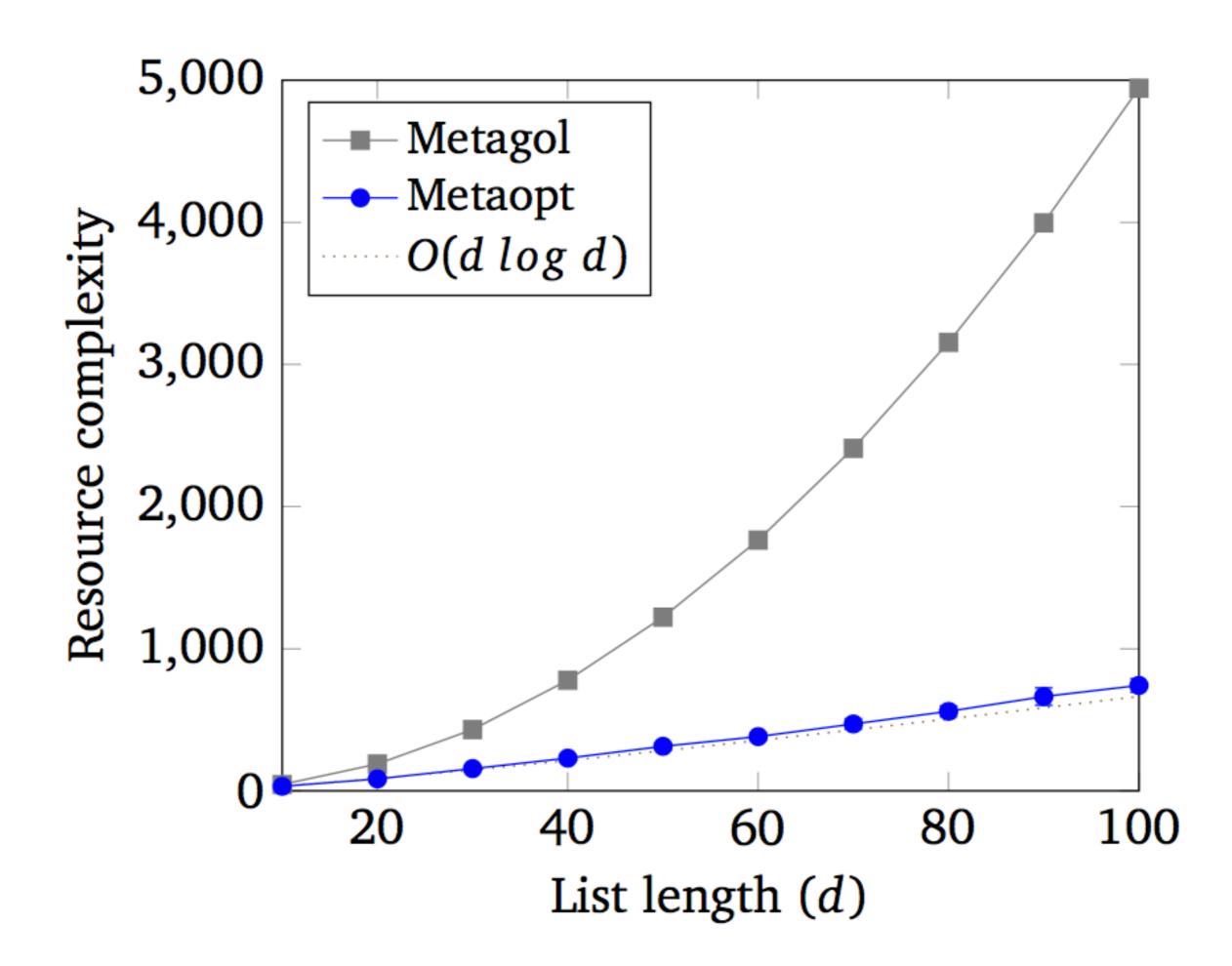
(b) Program runtimes

# Resource complexity

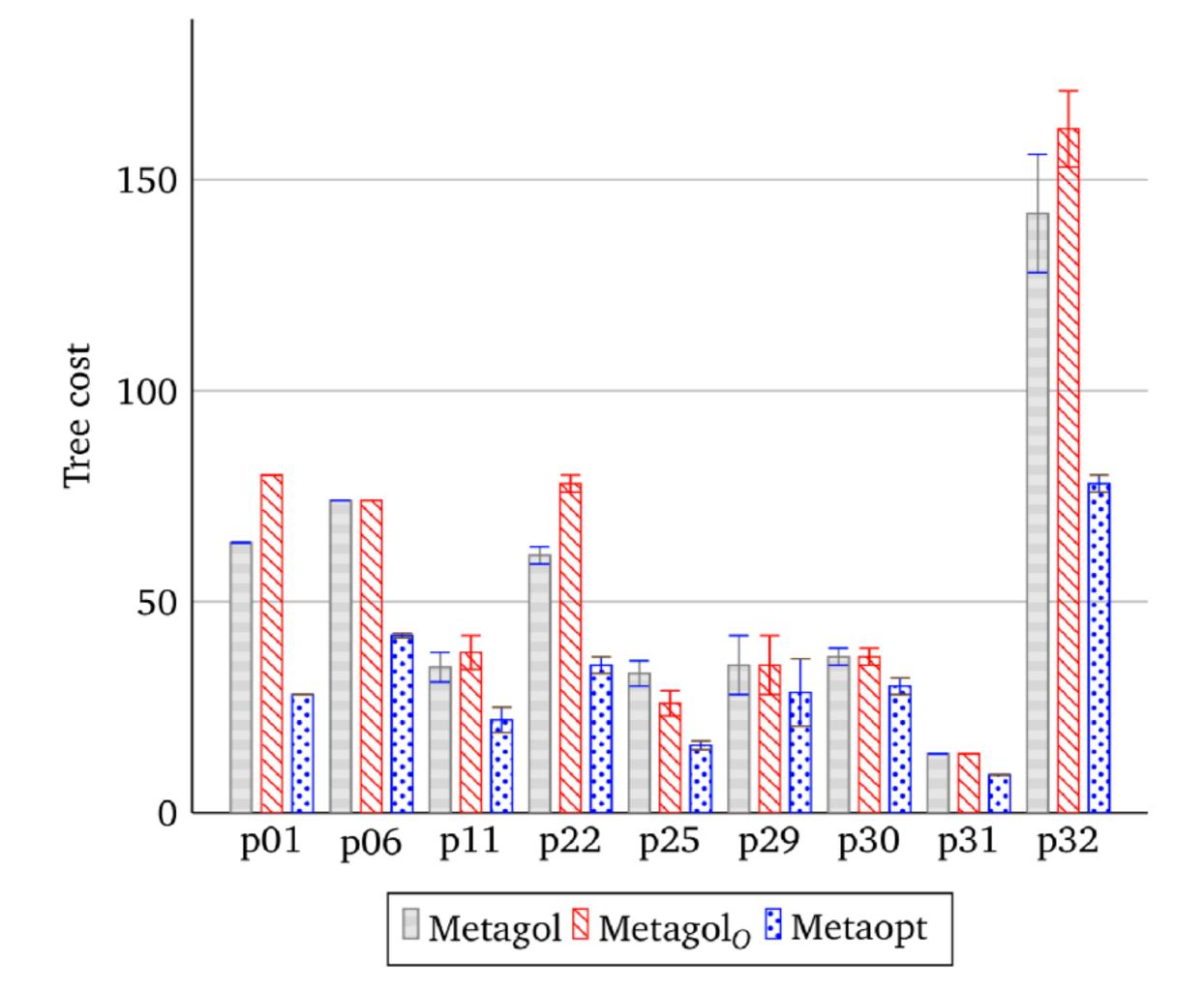


Initial state

Final state



Input	Output
My name is John.	John
My name is Bill.	Bill
My name is Josh.	Josh
My name is Albert.	Albert
My name is Richard.	Richard



```
%% metagol
f(A,B):-tail(A,C),f1(C,B).
f1(A,B):-dropLast(A,C),f2(C,B).
f2(A,B):-dropWhile(A,B,not_uppercase).
```

```
%% metagol unfolded
f(A,B):-
    tail(A,C),
    dropLast(C,D),
    dropWhile(D,B,not_uppercase).
```

```
%% metagol unfolded
f(A,B):-
    tail(A,C),
    dropLast(C,D),
    dropWhile(D,B,not_uppercase).
4n
```

```
% metagol0
f(A,B):-f1(A,C),f4(C,B).
f1(A,B):-f2(A,C),f3(C,B).
f2(A,B):-filter(A,B,is_letter).
f3(A,B):-dropWhile(A,B,is_uppercase).
f4(A,B):-dropWhile(A,B,not_uppercase).
```

```
% metagolO unfolded
f(A,B):-
   filter(A,C,is_letter).
   dropWhile(C,D,is_uppercase),
   dropWhile(D,B,not_uppercase).
```

```
% metagolO unfolded
f(A,B):-
    filter(A,C,is_letter).
4n    dropWhile(C,D,is_uppercase),
    dropWhile(D,B,not_uppercase).
4n
```

```
% learning f/2
% clauses: 1
% clauses: 2
% clauses: 3
% is better: 67
% is better: 57
% clauses: 4
% is better: 55
% clauses: 5
% is better: 53
% is better: 51
% is better: 49
% is better: 46
% clauses: 6
% is better: 41
% is better: 36
% is better: 31
f(A,B):-tail(A,C),f_1(C,B).
f_1(A,B):-f_2(A,C),dropLast(C,B).
f_2(A,B):-f_3(A,C),f_3(C,B).
f_3(A,B):-tail(A,C),f_4(C,B).
f_4(A,B):-f_5(A,C),f_5(C,B).
f_5(A,B):-tail(A,C),tail(C,B).
```

```
% metaopt
f(A,B):-tail(A,C),f1(C,B).
f1(A,B):-f2(A,C),dropLast(C,B).
f2(A,B):-f3(A,C),f3(C,B).
f3(A,B):-tail(A,C),f4(C,B).
f4(A,B):-f5(A,C),f5(C,B).
f5(A,B):-tail(A,C),tail(C,B).
```

```
% metaopt unfolded
f(A,B):-
    tail(A,C),
    tail(C,D),
    tail(D,E),
    tail(E,F),
    tail(F,G),
    tail(G,H),
    tail(H,I),
    tail(I,J),
    tail(J,K),
    tail(K,L),
    tail(L,M),
    dropLast(M,B).
```

```
% metaopt unfolded
f(A,B):-
    tail(A,C),
    tail(C,D),
    tail(D,E),
    tail(E,F),
    tail(F,G),
    tail(G,H),
    tail(H,I),
    tail(I,J),
    tail(J,K),
    tail(K,L),
    tail(L,M),
   dropLast(M,B).
```

does this last

## todo

- Study complexity of Metaopt variants
- Characterise complexity of learned programs
- Learning efficient functional programs
- Discover new efficient algorithms

## Todo (in general)

- Learning efficient programs
- Learning reusable higher-order abstractions
- Learning from minimal biases
- Never-ending / lifelong learning

## Learning abstractions

Input	Output
[dagstuhl,2017]	[dagstuh,201]
[alice,bob,charlie]	[alic,bo,charli]
[1234,12,564]	[123,1,56]
[ab,abc,abcd,abcde]	???

Input	Output
[dagstuhl,2017]	[dagstuh,201]
[alice,bob,charlie]	[alic,bo,charli]
[1234,12,564]	[123,1,56]
[ab,abc,abcd,abcde]	[a,ab,abc,abcd]

```
f(A,B):-map(A,B,f1).
f1(A,B):-f2(A,C),tail(C,D),f2(D,B).
f2(A,B):-reduceback(A,B,concat).
```

Input	Output
[dagstuhl,2017]	[dagstuh]
[alice,bob,charlie]	[alic,bo]
[1234,12,564]	[123,1]
[ab,abc,abcd,abcde]	???

Input	Output
[dagstuhl,2017]	[dagstuh]
[alice,bob,charlie]	[alic,bo]
[1234,12,564]	[123,1]
[ab,abc,abcd,abcde]	[a,ab,abc]

```
f(A,B):-map(A,C,f1),f1(C,B).
f1(A,B):-f2(A,C),tail(C,D),f2(D,B).
f2(A,B):-reduceback(A,B,concat).
```

## learn f(A,B):-map(A,C,f1),f1(C,B).f1(A,B):-f2(A,C),tail(C,D),f2(D,B). f2(A,B):-reduceback(A,B,concat).

learn