

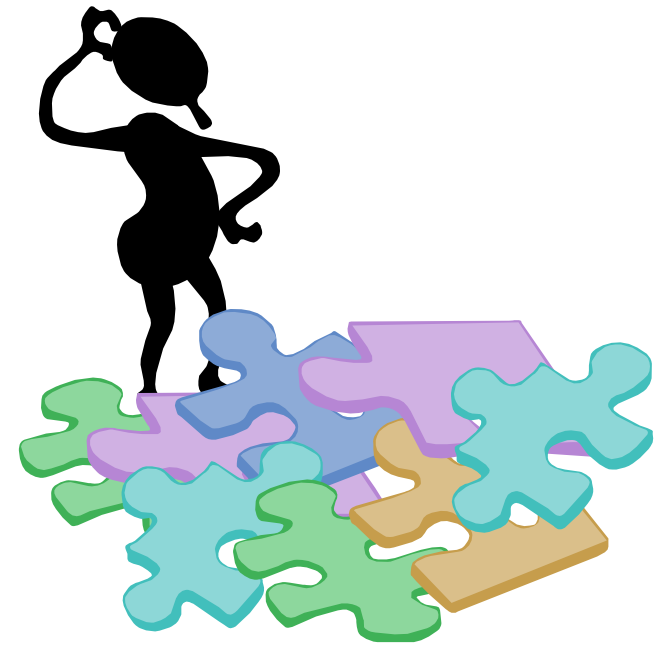


# Systematic search for functional programs

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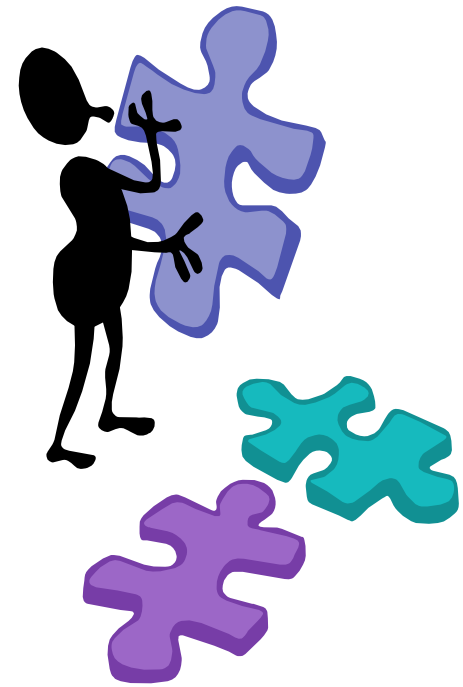


# *Exploiting strong typing*



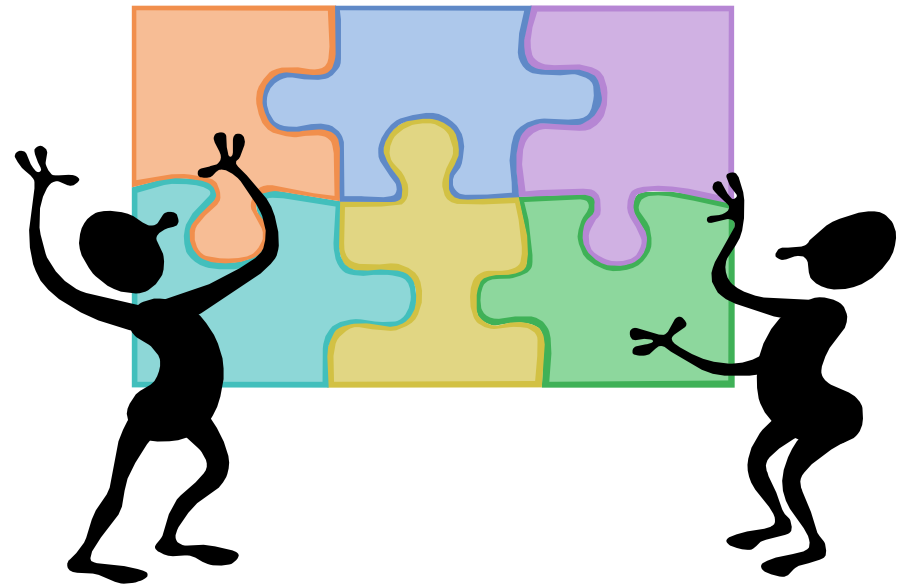


# *Exploiting strong typing*





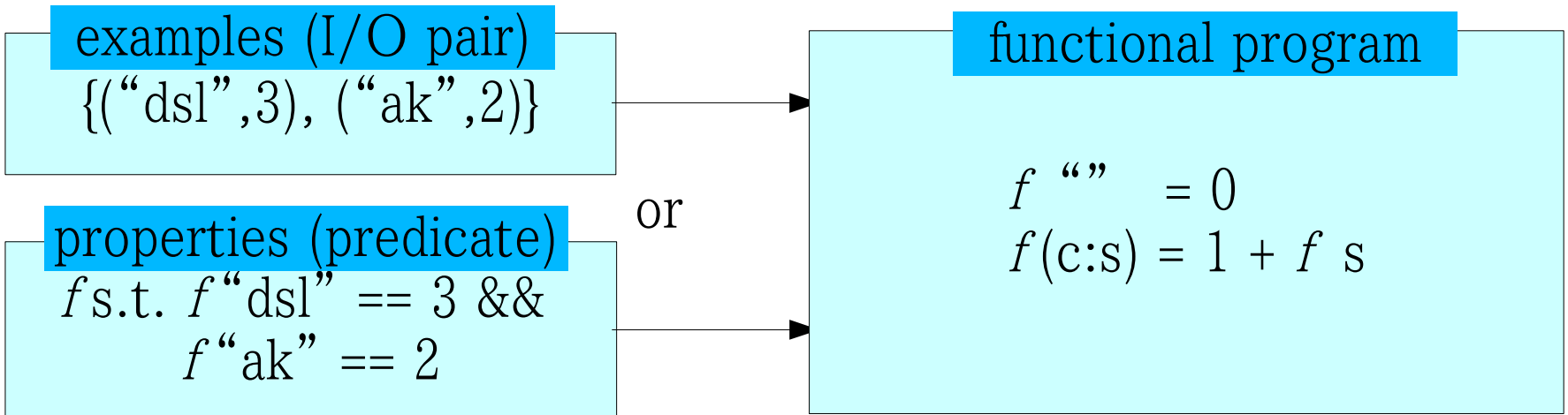
# *Exploiting strong typing*





# *Inductive synthesis of functional programs*

Inductive synthesis of functional programs (Inductive functional programming)  
**automatic** generation of **functional programs** from **examples/properties**.



\* Properties can be more flexible than examples.

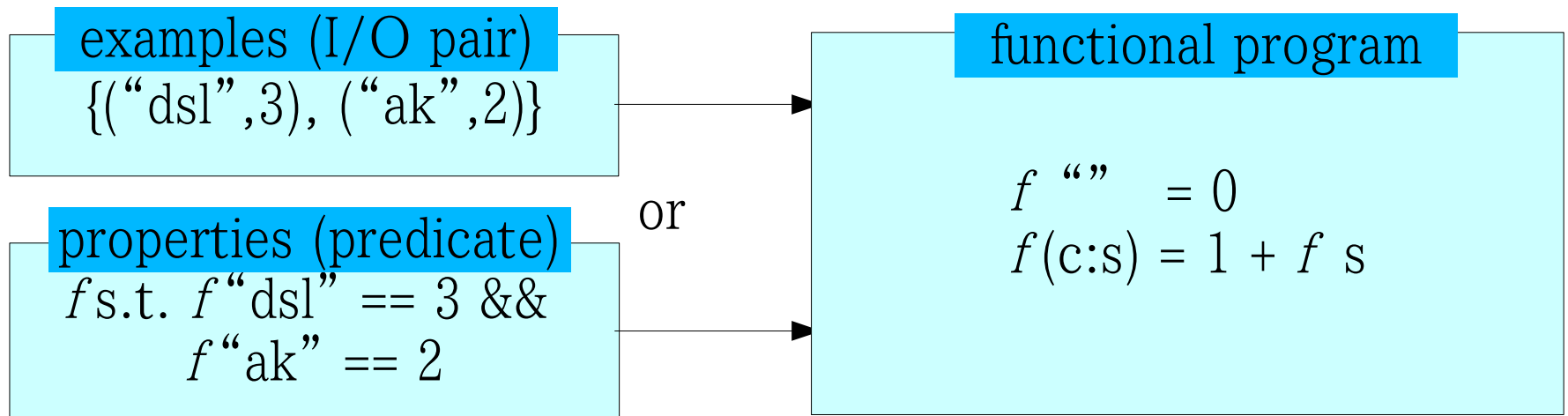


# *The implemented system*

Please watch the demo....

# Inductive synthesis of functional programs

Inductive synthesis of functional programs (Inductive functional programming):  
automatic generation of functional programs from examples/properties.



\* Properties can be more flexible than examples.

Human programmers usually:

Write a function  $\longrightarrow$  review it  $\longrightarrow$  test it

could be automated

You should still do that!



# *Outline*

- **Introduction**

motivation, policy, & conventional approaches

- **The implemented system**

rough specification & detailed implementation

- **Experiments**

results on some easy problems

- **Conclusions**





# *Placement*

- “Deductive” synthesis from complete spec.
- “Inductive” synthesis from incomplete spec.
  - by generating & folding **computational traces** from examples
  - **search-based**, or by trial and error
    - Genetic Programming (GP)
    - The proposed method

## *related work(1):*

# *Synthesis via computational traces*

- Step 1: Generate computational traces
  - by constraints on examples (Summers 1977)
  - by hand (a.k.a. programming by demonstration (PBD))
  - by a universal planner (Schmid 2001)
  - by genetic programming (Schmid 2004) } by search
- Step 2: Fold the traces into a recursive program
  - by pattern matching

Drawback: **unfolded traces are longer than programs**

search space bloats much earlier

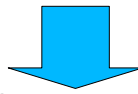
 (than direct search methods like GP)

*related work(2):*

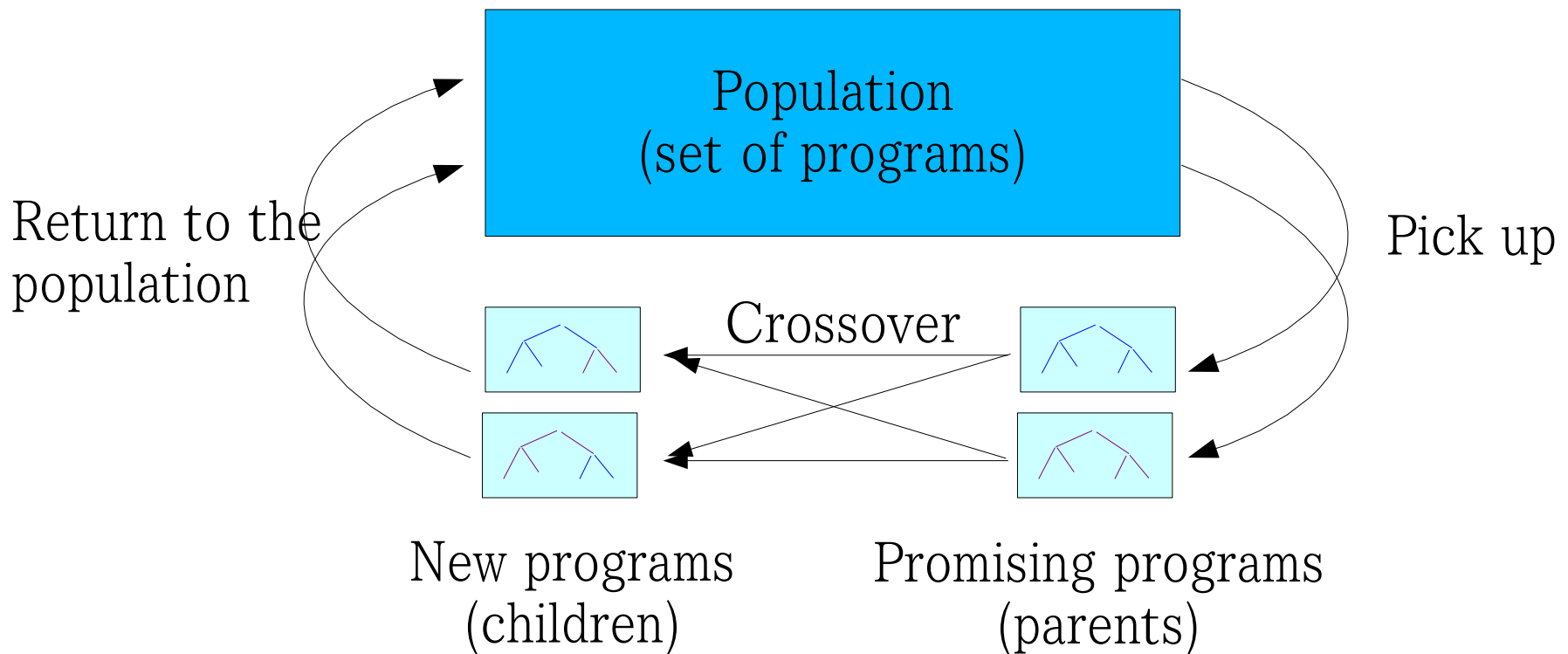
# *Genetic programming (1/3)*

**assumption:**

“useful programs consist of useful subexpressions”



## **Search by recombination of subexpressions**



 *related work(2):*

*Genetic programming (2/3)*

## **GP applied to inductive algorithm synthesis**

**our interest:**

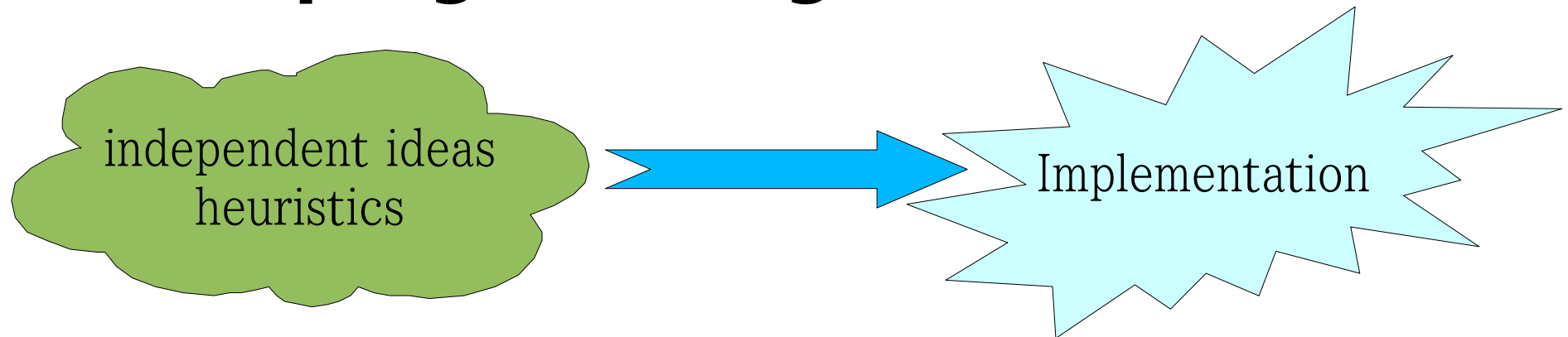
synthesis of **reusable** (often **typeful** and **recursive**) algorithms

**NB: typical use of GP: typeless CSP and function approx.**

- ADATE (Olsson 1995)
  - monomorphic first-order type system
  - requires a file with tens of lines written for each synthesis
  - interesting results reported, but **not reproducible**
- PolyGP (Yu 1998)
  - polymorphic higher-order type system
  - requires a file with tens of lines written for each synthesis
  - **exhaustive search is faster** (Katayama 2004)

*related work(2):*  
*Genetic programming (3/3)*

## Genetic programming research



- different scientists use different settings  
“100 researchers, 100 genetic algorithms
- no comparison with non-heuristic approaches  
It is often unclear if/which heuristic worked.



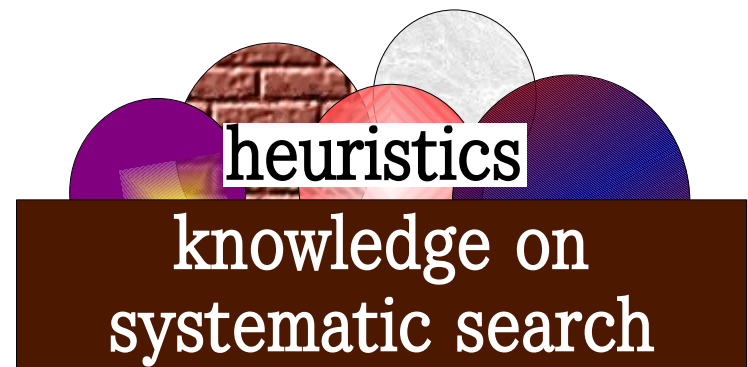
# *Motivation*

1) First we should investigate systematic search.

- How can we optimize its implementation?
- How efficient is it? Is it slower/faster than GP?
- etc.

2) Heuristics (incl' GP) should be added based on the above knowledge.

- better modularity
- steady improvement





# *Policy (1)*

- exploration of **exhaustive breadth-first search**

- construct a **basis** on which to build heuristic approaches like GP,
- control the **exponential bloat** in the search space.

*to be more concrete,*

- Limit to **(grammatically & type)-correct** expressions (of course!)
- Avoid multiple-count of **equivalent expressions**  
using known **reducible patterns** and (maybe) **transformation rules**  
e.g. *do not try both* `foldr foo bah []` *and* `bah`  
*do not try both* `(∀x -> foo x) bah` *and* `foo bah`
- **Optimize** the implementation - - - **memoization**, etc.



## *Policy (2)*

- **ease of use**

at least it should be easier than writing the programs directly

- just writing a boolean function as predicate  
invokes the search
- use general-purposed component combinator set,  
not tailor-made function set for each synthesis





# *The implemented system*

- **Introduction**

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rough specification & detailed implementation

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compared with polymorphic GP

- **Conclusions**



# Specification

keyboard

- property,  
e.g.,  $\backslash f. f \text{ "slkd" } == 4$
- type (optional)  
e.g.  $[a] \rightarrow \text{Int}$

\* The type is inferred when omitted.

component library file

```
zero :: Int
zero = 0
inc  :: Int -> Int
inc  = \ x -> x+1
nat_para :: Int -> a -> (Int -> a -> a) -> a
nat_para = \ i x f ->
    if i==0 then x else
    f (i-1) (nat_para (i-1) x f)
...
```

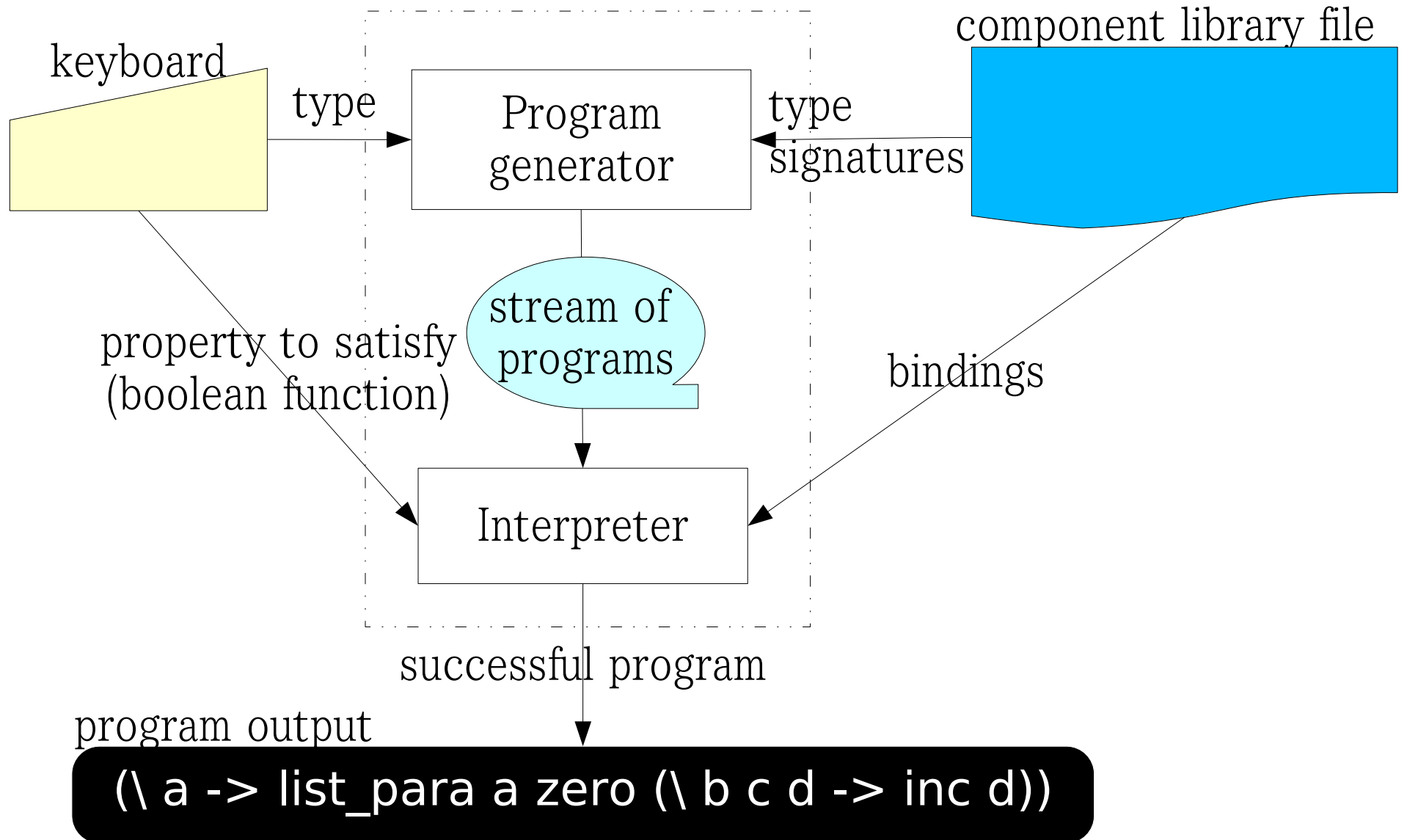
synthesizer

program output

```
(\ a -> list_para a zero (\ b c d -> inc d))
```



# *Internal structure*





# *Language*

The language: Haskell subset

- Hindley-Milner type system

**without functions in containers, such as  $[a \rightarrow b]$ ,  $(a, b \rightarrow c)$ ,**

...(for efficiency reasons)

- Frontend ... usual lambda calculus

Backend ... de Bruijn lambda calculus

Katayama (2004) used typed *SIBC* combinators, but

- Such combinatory expressions are redundant;
- Search tree branches a lot in the shallow stages;
- The complexity of involved types bloats.



# *Implementation*

## Story:

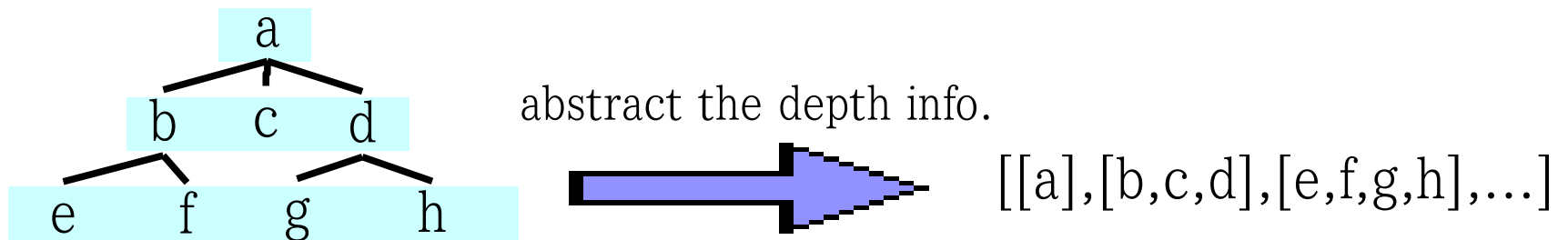
- Preparation
  - Monad for breadth-first search
  - Monad transformer for type inference
- Arity known cases **where functions never return type vars**
- Arity unknown cases
  - where there can be a function of the form “forall a. Foo → Bar → ... → a”
- Memoization & devices for making keys hit often



# Monad for breadth-first search

- Spivey (2000)'s monad:

Matrix  $a = \text{Stream} (\text{Bag } a)$  (actually defined as  $[[a]]$ )  
1<sup>st</sup> bag for depth 0, 2<sup>nd</sup> for depth 1, ... of the search tree



× for combinations and + for alternatives (defined as usual)

*Eases implementation a lot* but **gobbles the heap a lot**

- Recomputing variant:

Recomp  $a = \underline{\text{Int}} \rightarrow \text{Bag } a$  suppresses heap consumption  
└─ depth



## *Monad trans. for type inference*

- Typical implementation:

$\text{TI } a = \text{Subst} \rightarrow \text{Int} \rightarrow \text{Maybe } (a, \text{Subst}, \text{Int})$

states

Subst: “current” substitution

Int: fresh variable ID

- Monad transformer considering alternative states:

$\text{TI } \text{Recomp } a$

where  $\text{TI } m a = \text{Subst} \rightarrow \text{Int} \rightarrow m (a, \text{Subst}, \text{Int})$

# *Naïve implementation* (arity known)

```
data Expression = Lambda Expression      -- lambda abstraction
                | X      Int            -- de Bruijn variable
                | Expression :$ Expression -- function application

uniExprs :: [(Expression,Type)] -> [Type] -> Type -> TI Recomp Expression
uniExprs prims avails (t0:->t1) = do result <- uniExprs prims (t0:avails) t1
                                return (Lambda result)

uniExprs prims avails reqret
  = do (expr,typ) <- msum $ map return $ (zip (map X [0..]) avails ++ prims)
       typ' <- freshVariablesForForalls typ
       setMGU reqret (returnTypeOf typ')    -- setMGU :: Type -> Type -> TI m ()
       spine avails typ' expr
where spine :: Type -> Expression -> TI Recomp Expression
      spine (t:->ts) fun                -- (:->) represents function type constructor
      = lift delay (do subst <- getSubst
                    arg <- uniExprs prims (map (apply subst) avails) (apply subst t)
                    spine ts (fun :$ arg) )
      spine _avs _t      fun = return fun
```

**\*simplified to improve presentation rather than the efficiency**

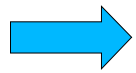
**e.g. Expressions with the same type are actually put together to [Expression]**



# Primitives returning type variables (arity unknown)

$\forall a. T_1 \rightarrow T_2 \rightarrow a$  can be specialized

$T_1' \rightarrow T_2' \rightarrow R, T_1'' \rightarrow T_2'' \rightarrow (A_1 \rightarrow R), T_1''' \rightarrow T_2''' \rightarrow (A_1 \rightarrow A_2 \rightarrow R), \dots$   
( $T_1', T_2', T_1'', T_2'', T_1''', T_2''', \dots$  are specialization of  $T_1$  and  $T_2$ )



## **Arity undecidable**

- solution 1: generate all alternatives

( $a \mapsto R$ ) case *implus* ( $a \mapsto b \rightarrow R$ ) case *implus* ( $a \mapsto b \rightarrow c \rightarrow R \dots$ )

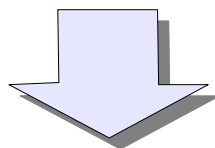
problem: create lots of equivalent programs

- solution 2(current): introduce new constructor representing direct product of type variables:  
 $b, (b,c), (b,c,d), \dots$



# Memoization

The same simple type is requested many times



**Memoize**  $Type \rightarrow [Type] \rightarrow TI \text{ Recomp } [Expression]$   
(or\*  $\underline{Type} \rightarrow [\underline{Type}] \rightarrow \underline{Int} \rightarrow [([Expression], Subst, Int)]$  )



Queries often hit.

This list becomes long.

\* Exactly,  $Type \rightarrow [Type] \rightarrow Subst \rightarrow Int \rightarrow Int \rightarrow [(Expression, Subst, Int)]$   
but you could apply the substitution and rename the type variables to normalize its numbering.



# *Reorganizing* (idea)

Point:

Only the set of available types matters  
× list

e.g. for  $X_0, X_1 :: Char$

$\dots X_0 \dots X_0 \dots$  is type correct  $\Leftrightarrow \dots X_0 \dots X_1 \dots$  is type correct  
 $\Leftrightarrow \dots X_1 \dots X_0 \dots$  is type correct  
 $\Leftrightarrow \dots X_1 \dots X_1 \dots$  is type correct

$available = \{X_0, X_1 :: Char\}$  case and

$available = \{X_0 :: Char\}$  case can share the same memo entry\*

\* *post processing required*



# Reorganizing (implementation)

Available variables:  $x_0 :: Char, x_1 :: Bool, x_2 :: Char$

Memo entry:  $x_0 :: Char, x_1 :: Bool$

Diagram illustrating the mapping of available variables to memo entries. Arrows point from  $x_0$  and  $x_1$  in the available variables to  $x_0$  and  $x_1$  in the memo entry. A diagonal arrow labeled "preprocess (sort & nub)" points from  $x_2$  in the available variables to  $x_0$  in the memo entry.

Retriever map:  $\{0 \mapsto [0, 2], 1 \mapsto [1]\}$

## Memo function wrapper

1. sort the argument types and assign **one variable name** for **one type**
2. invoke memo function
3. generate all cases by replacing variables (using retriever map).



## *Avoiding reducible expressions*

- Optimization rules suggest redundancy in the program space.
- Because expressions with the same type are put together, rules that cannot be detected from type info. might not help a lot.
- Current rule to identify reducible points:  
the strict argument of consumer functions  
(like case, cata, para) must not be constant  
(i.e. must include a free variable)  
(Note: quite limited)



# *Experiments*

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# *Problems*

Problems from the previous work (Katayama 2004) for comparison

- $\text{nth} :: \text{Int} \rightarrow [a] \rightarrow a$

$\text{nth } 5 \text{ "widjfgwi"} == \text{'f' \&\&}$

$\text{nth } 1 \text{ "wddidjfgwi"} == \text{'w'}$

- $\text{map} :: (b \rightarrow a) \rightarrow [b] \rightarrow [a]$

$\text{map } (==\text{'c'}) \text{ "stock"} == [\text{False}, \text{False}, \text{False}, \text{True}, \text{False}] \&\&$

$\text{map } (==\text{'e'}) \text{ "peeped"} == [\text{False}, \text{True}, \text{True}, \text{False}, \text{True}, \text{False}]$

- $\text{length} :: [a] \rightarrow \text{Int}$

$\text{length } \text{"hageho"} == 6 \&\& \text{length } \text{"hoge"} == 4$



# *The component combinators*

Previous work

- **S, K, I, B, and C**
- constructors
- curried paramorphisms
- head, tail and pred

New algorithm

- constructors
- curried paramorphisms
- head, tail and pred





# *Results*

Evaluation of proposed method:

Computation time (sec.)	nth	map	length
Old algorithm (real)	5.3	2.2	0.03
(user)	5.1	2.2	0.02
New algorithm (real)	0.8	1.9	0.03
(user)	0.6	1.2	0.02

- Improved for the problems used in the previous work.
- Greater programs (sized more than 12) still require more than a minute or cannot be synthesized.

# Discussion towards a usable system: # of equivalent programs

The results so far are happened-to-work toy examples!

Synthesis of  $take :: Int \rightarrow [a] \rightarrow [a]$  requires tens of seconds.

.... How much it might **potentially** be optimized?

➔ Do (very) light-weighted random testing

## **EXPERIMENT TO GIVE SOME ESTIMATE:**

for each  $f :: Int \rightarrow [a] \rightarrow [a]$  in the generated expressions,

- compute  $(f\ 2\ "12332", f\ 0\ "56789", f\ 2\ "k", f\ 0\ "")$   
> file
- compare `wc file` and `sort file | uniq | wc`

file ... 251940 lines  
sort file | uniq ... 514 lines

**LOTS OF ROOM FOR  
IMPROVEMENT**



# *Conclusions*

- **Introduction**

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## *Summary*

- investigated breadth-first exhaustive search for de Bruijn expressions
- easy problems sized around ten can be solved within seconds
- Still room for efficiency improvement



## *Future work*

- Controlling exponential bloat
  - Solution: **remove more equivalent programs**
    - or lower the priorities of seem-to-be-equivalent programs
  - transformation rules
  - use lightweighted random testing for small expressions
- Type classes: **dealing with contexts**
  - contexts “generic in nature” e.g. Eq, Show, Read, etc.
    - ... just ignore.
  - other “ad hoc” classes like Num ... needs implementation.
- Heuristics:
  - e.g. to prioritize subexpressions of expressions
  - that returned the correct outputs to some of the inputs