

# AUTOMAP:

## Inferring Rank-Polymorphic Function Applications with Integer Linear Programming

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## Example

```
def dotprod x y = sum (map (*) x y)
```

# Rank polymorphism

- $1 + 2 \Rightarrow 3$

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- $[1, 2, 3] + 4 \Rightarrow [5, 6, 7]$
- $\text{sqrt } [[1, 4, 9], [16, 25, 36]] \Rightarrow [[1, 2, 3], [4, 5, 6]]$

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`map (+) [1,2,3] [4,5,6]` vs. `[1,2,3] + [4,5,6]`

- Practically all rank polymorphic languages are **dynamic**:  
NumPy, APL, MATLAB, ...

- `map f xs` applies `f` to each element of `xs`:

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- You can `map` functions that take multiple arguments too:

`map (+) [x_0, ..., x_n] [y_0, ..., y_n]`  
`= [x_0 + y_0, ..., x_n + y_n]`

## map and rep

- **map**  $f$   $x_s$  applies  $f$  to each element of  $x_s$ :

**map**  $f$   $[x_0, x_1, \dots, x_n] = [f\ x_0, f\ x_1, \dots, f\ x_n]$

- You can **map** functions that take multiple arguments too:

**map**  $(+)$   $[x_0, \dots, x_n]$   $[y_0, \dots, y_n]$   
 $= [x_0 + y_0, \dots, x_n + y_n]$

- **rep**  $x$  makes an array of unspecified length whose elements are all  $x$ :

**rep**  $x = [x, x, \dots, x]$

- ▶ We'll ignore the question of how many elements are needed.



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f: []int -> [][]int -> int  
f xss xss
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**rep**s can often be eliminated

```
map (\xs -> f xs xss) xss
```

## Goal

For each function application, the compiler should automatically insert **map**s or **rep**s to make the application **rank-correct**.

$$f \ x \implies \begin{cases} \text{map} \ ( \dots \ (\text{map} \ f) \ \dots ) \ x \\ \text{or} \\ f \ (\text{rep} \ \dots \ (\text{rep} \ x)) \end{cases}$$



## Challenge: ambiguity

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length : []a -> int
xss : [][]int

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Many rank-correct elaborations:

1. `sum (rep (length xss))`
2. `sum (map length xss)`
3. `map sum (map (map length) (rep xss))`
4. ...

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An application can be **map**ped or **rep**ped (or neither) but **never both**.

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- **BAD:**

- ▶ `map f (rep x)`
- ▶ `(map (map g)) (rep x)`

- Never necessary to **map** and **rep** in the same application to obtain a rank-correct program.

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- Generally aligns with programmer's intent/simple mental model.
- Minimization over **all** the applications of a top-level definition:
  - ▶ Only have to choose from the set of minimal solutions.

`sum (length xss)` can be elaborated to:

1. `sum (rep (length xss))`
2. `sum (map length xss)`

## Challenge: elaboration is global

- `sum (map length xss)` is a **global minimal** elaboration of `sum (length xss)`.
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- `sum (map length xss)` is a **global minimal** elaboration of `sum (length xss)`.
  - ▶ Inserting the `map` for the inner `length` application requires considering the outer `sum`.
- To find all minimal elaborations, must consider all applications **simultaneously**.

## Challenge: type variables

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- A type variable can have any rank!
- How do we statically insert **maps** and **reps** in the presence of type variables, whose ranks aren't known?

# Constraints

- Suppose

f : p  $\rightarrow$  b

x : a

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- For example:  $|[\llbracket \rrbracket]_{\text{int}}| = 2$  and  $|\text{int}| = 0$ .



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- ▶ Example:

```
sqrt : int -> int  
[1,2,3] : []int
```

Application `sqrt [1,2,3]` gives the constraint

$$M + \underbrace{|int|}_0 = \underbrace{|[]int|}_1 \implies M = 1$$

$M$  is equal to the number of **map**s required: `map sqrt [1,2,3]`

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$$1 + |a| = R \implies R = 1, 2, 3, \dots$$

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$$|[[]a]| = R + |\text{int}|$$

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$R$  is equal to the number of **rep**s required:

- ▶ `length (rep 3)`
- ▶ `length (rep (rep 3))`
- ▶ ...

# Constraints

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$$M = 0 \text{ or } R = 0$$



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- Example: `sum (length xss)`

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- Rule 2: Minimize the number of **maps** and **reps**
- The or-constraints can be linearized to obtain an **Integer Linear Program (ILP)**.

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4. Type check elaborated program and continue with compilation.

Formalization

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$$\begin{array}{l} e ::= v \\ \quad | \quad x \quad (x \in V_p) \\ \quad | \quad [e, \dots, e] \\ \quad | \quad \mathbf{map} \ e \ e \\ \quad | \quad \mathbf{rep} \ e \\ \quad | \quad \boxed{e \ e \ \Delta \ (M, R)} \end{array}$$

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- The application  $e \ e \ \Delta \ (M, R)$  is a *flexible function application*.



# Flexible function applications

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- $f x \Delta (M, R)$  will ultimately be elaborated to

$$\mathbf{map}^{s_r(M)} f (\mathbf{rep}^{s_r(R)} x)$$

where  $s_r$  is the solution to the ILP and

$$\begin{array}{ll} \mathbf{map}^0 e = e & \mathbf{rep}^0 e = e \\ \mathbf{map}^{n+1} e = \mathbf{map} (\mathbf{map}^n e) & \mathbf{rep}^{n+1} e = \mathbf{rep} (\mathbf{rep}^n e) \end{array}$$

# Languages

- The formalization is split up into three languages:

	Source lang.		
Implicit <b>maps/refs</b>	✓		
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# Type checking

- Constraint-based type system; judgements of the form

$$\Gamma \vdash e :_S \sigma \parallel C$$

- Under environment  $\Gamma$ ,  $e$  has scheme  $\sigma$  with *frame*  $S$  when the constraints in  $C$  are satisfied.

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$$\Gamma \vdash \text{sqrt } [1, 2, 3] :_{[]}^M \text{int} \parallel \{[]^M \text{int} \stackrel{?}{=} [] \text{int}\}$$

- ▶ Frames are needed for proofs, improved ambiguity checking, and implementation optimizations.

# Type checking rules - C-APP

- The C-APP rule types flexible function applications.

$$\Gamma \vdash e_1 :_{S_1} \tau_1 \rightarrow \tau_2 \parallel C_1 \quad \Gamma \vdash e_2 :_{S_2} \tau_3 \parallel C_2$$

---

C-APP

# Type checking rules - C-APP

- The C-APP rule types flexible function applications.

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- The generated constraints are relaxed to rank constraints, which are used to form the ILP:

$$[ ]^M S_1 \tau_1 \stackrel{?}{=} [ ]^R S_2 \tau_3 \longrightarrow M + |S_1| + |\tau_1| \stackrel{?}{=} R + |S_2| + |\tau_3|$$

## Type checking rules - C-DEF

- The C-DEF rule types top-level polymorphic definitions and dispatches the constraint set for each top-level definition

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- ▶  $s$  also substitutes type variables.

# SOLVE: Constraint set solving algorithm

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7   if  $s'_r$  then  
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# Properties of SOLVE

- All satisfiers  $s$  can be decomposed into a type and rank substitution:

## Proposition

If  $s$  satisfies  $C$ , there exists a rank substitution  $s_r$  that satisfies  $|C|$  and there exists a closed type substitution  $s_t$  such that  $s|_{\text{ftv}(C) \cup \text{frv}(C)} = s_t \circ s_r$ .

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- You can always build a satisfier  $s_t \circ s_r$  of  $C$  from a satisfier  $s_r$  of  $|C|$ :

## Proposition

If  $C$  is satisfiable and  $s_r$  satisfies  $|C|$  then there is a closed type substitution  $s_t$  such that the substitution  $s = s_t \circ s_r$  satisfies  $C$ .

## Elaboration

- Reminder: a satisfying substitution  $s$  for a constraint set  $C$  (which SOLVE gives us) instantiates all rank variables with integral ranks:

$$s(\text{sqrt } [1, 2, 3] \Delta (M, R)) \longrightarrow \text{sqrt } [1, 2, 3] \Delta (1, 0)$$



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- The AM transformation converts from the internal language to the target language:

$$\begin{aligned} \text{AM}([e_1, \dots, e_m]) &= [\text{AM}(e_1), \dots, \text{AM}(e_m)] \\ \text{AM}(\text{def } f \ x = e ; p) &= \text{def } f \ x = \text{AM}(e) ; \text{AM}(p) \\ \text{AM}(e_1 \ e_2 \Delta (n_M, n_R)) &= \boxed{\text{map}^{n_M} \text{AM}(e_1) (\text{rep}^{n_R} \text{AM}(e_2))} \\ &\vdots \end{aligned}$$

# AM Properties

- The AM transformation preserves well-typedness of programs.

## Proposition: Well-typedness

If  $\Gamma \vdash e :_S \sigma \parallel C$  and  $s$  is a satisfier of  $C$ , then  $s(\Gamma) \vdash \text{AM}(s(e)) : s(S \sigma)$ .

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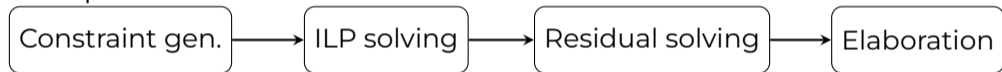
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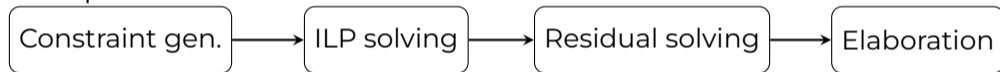
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- Aside from size inference, works very similarly to the SOLVE algorithm, with a couple of exceptions.

## Implementation - induced **reps**

- map ( $\lambda y. xs * y$ ) *ys* can be elaborated to

map ( $\lambda y. \mathbf{map} (*) xs (\mathbf{rep} y)$ ) *ys* (1) or map ( $\lambda y. \mathbf{map} (*) xs y$ ) ( $\mathbf{rep} ys$ ) (2)

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- map ( $\lambda y. \text{map} (\lambda x. x * y) xs$ )  $ys$  has size **1**.
- $\Rightarrow$  we can disambiguate by only counting **non-induced rep**s.



## Implementation - induced **reps**

- **Frames** track the number of **maps** in an application ( $S = []^M S_1$ ):

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- ILP objective becomes

$$\dots + M + \max(0, |R| - |S|) + \dots$$

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## Practical impact

- Difficult to quantify value of feature that is glorified syntax sugar.
- We (manually!) rewrote programs to take advantage of AUTOMAP when we judged it improved readability.

## Practical impact: before

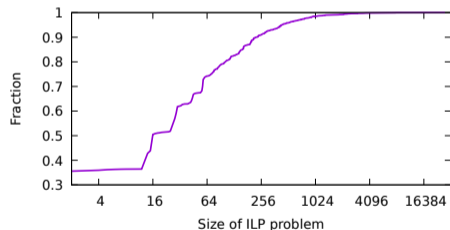
```
def main [nK][nX]
  (kx: [nK]f32) (ky: [nK]f32) (kz: [nK]f32)
  (x: [nX]f32) (y: [nX]f32) (z: [nX]f32)
  (phiR: [nK]f32) (phiI: [nK]f32)
  : ([nX]f32, [nX]f32) =
let phiM = map (\r i -> r*r + i*i) phiR phiI
let as = map (\x_e y_e z_e ->
  map (2*pi*)
    (map (\kx_e ky_e kz_e ->
      kx_e*x_e + ky_e*y_e + kz_e*z_e)
      kx ky kz))
  x y z
let qr = map (\a -> sum(map2 (*) phiM (map cos a))) as
let qi = map (\a -> sum(map2 (*) phiM (map sin a))) as
in (qr, qi)
```

## Practical impact: after

```
def main [nK][nX]
  (kx: [nK]f32) (ky: [nK]f32) (kz: [nK]f32)
  (x: [nX]f32) (y: [nX]f32) (z: [nX]f32)
  (phiR: [nK]f32) (phiI: [nK]f32)
  : ([nX]f32, [nX]f32) =
let phiM = phiR*phiR + phiI*phiI
let as = 2*pi*(kx*transpose (rep x)
  + ky*transpose (rep y)
  + kz*transpose (rep z))
let qr = sum (cos as * phiM)
let qi = sum (sin as * phiM)
in (qr, qi)
```

# Metrics from changing a benchmark suite

Proportion of ILP problems that have less than some given number of constraints.



Number of programs: 67

Lines of code: 8621  $\Rightarrow$  8515

Change in maps: 467  $\Rightarrow$  213

Largest ILP size: 28104 constraints

Median ILP size: 16 constraints

Mean ILP size: 116 constraints

Mean type checking slowdown: 2.50 $\times$

# Related work

- **Typed Remora:**

- ▶ Very general/powerful; binds shape variables in types:

$$\text{sum} : \forall S.S \text{ int} \rightarrow \text{int}$$

- ▶ Inference is very difficult.

# Related work

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- ▶ Cool rank polymorphism encoding in Haskell.
- ▶ Complicated function types (and potentially error messages).

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- **Single-assignment C:**

- ▶ Has *rank specialization* where functions have specialized definitions depending on the rank of the input.
- ▶ No parametric polymorphism or higher-order functions.

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- AUTOMAP is a conservative extension of/compatible with a Hindley-Milner type system for array programming.
- Anything inferred can also be inserted explicitly (much like classic type systems!)
- Type checking based on some heavy machinery (ILP), but we suspect of a fairly simple kind.
- Implemented in Futhark, but not really production ready yet.
  - ▶ Todo: quality of type errors, type checking speed, alternate ambiguity checking.

- Check out Futhark: <https://futhark-lang.org>
  - ▶ There's a blog post on AUTOMAP that covers much of this talk in more detail.
  - ▶ An AUTOMAP paper will be published at OOPSLA 2024, preprint available at <https://rschenck.com>.

