

Automating Kan composition

WG6 kick-off meeting: Syntax and Semantics of Type Theories

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A tactic for Cubical Agda

PhD project:

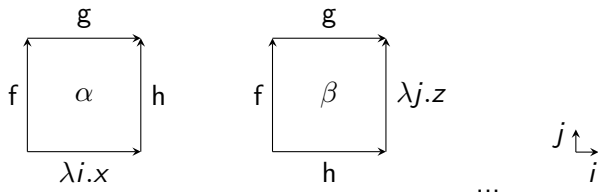
- ▶ Verify some results in computational topology: Use cubical type theory to study homotopy types of simplicial complexes.
- ▶ Central in many proofs: Use built-in Kan composition to show contractibility of certain types.

Steep learning curve when using hcomp's...

Goal: Automatic procedure to construct open cubes whose lids prove a goal under question.

Simplices to cubes

Example: We can map a triangle $\alpha : g \circ f \Rightarrow h$ between $f : x \rightarrow y$, $g : y \rightarrow z$ and $h : x \rightarrow z$ into cubical sets in different ways:



How can we turn α into β ?

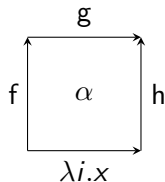
Syntax

We look at a fragment of Cubical Agda with PathP and \wedge , \vee .
Given infinite set of dimensions $D = \{i, j, \dots\}$, endpoints $0_{\mathcal{I}}$, $1_{\mathcal{I}}$:

$\Gamma ::= () \mid a : \Phi, \Gamma$	(contexts)
$r, s ::= i \mid (r \vee s) \mid (r \wedge s)$	(formulas)
$t, u, v ::= a \mid \lambda i. t \mid t r$	(terms)
$\Phi ::= \bullet \mid \text{PathP } (\lambda i. \Phi) u v$	(shapes)

Example:

$x : \bullet, y : \bullet, z : \bullet, f : \text{PathP } \bullet x y,$
 $g : \text{PathP } \bullet y z, h : \text{PathP } \bullet x z$
 $\alpha : \text{PathP } (\lambda j. \text{PathP } \bullet (f j) (h j)) \lambda i. x g$



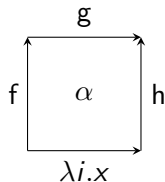
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Given $t : \Phi$, we will denote with $t|_{k=e}$, $\Phi|_{k=e}$ its $(k + e)$ -th face.

Example: $\alpha|_{1=0_{\mathcal{I}}} = \lambda i. x$, $\alpha|_{2=1_{\mathcal{I}}} = h$.

An algorithm for filling cubes

Input: Γ a context with faces of some type, n -dimensional cube Φ

Output: A term $t : \Phi$ formed over Γ with abs, app and hcomp.

Step 1: Make all $(a : \Psi)$ in Γ with $\dim(\Psi) = m$ to n -cubes:

$$\lambda i_1 \dots \lambda i_n. a \ r_1 \ \dots \ r_m \text{ where } r_i \in \text{FreeDL}(i_1, \dots, i_n)$$

Let S be the collection of all degeneracies for terms in Γ .

If for some $t \in S$ we have $t : \Phi$, return t .

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To fill the 2-dimensional square β from above, we generate

$$S = \{ \lambda i. \lambda j. x, \lambda i. \lambda j. y, \lambda i. \lambda j. z, \\ \lambda i. \lambda j. f \ i, \lambda i. \lambda j. f \ j, \lambda i. \lambda j. f \ (i \vee j), \lambda i. \lambda j. f \ (i \wedge j), \dots, \\ \lambda i. \lambda j. \alpha \ i \ i, \lambda i. \lambda j. \alpha \ i \ j, \lambda i. \lambda j. \alpha \ i \ (i \vee j), \dots, \lambda i. \lambda j. \alpha \ (i \wedge j) \ (i \wedge j) \}$$

Constructing an open $(n + 1)$ -dimensional cube

Step 2: Solve constraint satisfaction problem.

- ▶ Variables X_B and $X_{i_k,0_I}, X_{i_k,1_I}$ for $1 \leq k \leq n$.
- ▶ Domains $D_B = S$ and $D_{i_k,e} = \{t \in S : t|_{n=1_I} = \Phi|_{k=e}\}$ for $1 \leq k \leq n, e \in \{0_I, 1_I\}$.
- ▶ Constraints
 - ▶ $X_{i_k,e}|_{n=0_I} = X_B|_{k=e}$ for $k = 1, \dots, n, e \in \{0_I, 1_I\}$.
 - ▶ $X_{i_k,e}|_{k+e=e'} = X_{i_l,e'}|_{l-e'=e}$ for $1 \leq k < l \leq n, e, e' \in \{0_I, 1_I\}$.

If solution exists, return $hcomp(X_{i_1,0_I}, X_{i_1,1_I}, \dots, X_{i_n,0_I}, X_{i_n,1_I}) X_B$

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For an open 3-cube, constrain vars $X_B, X_{i,0_I}, X_{i,1_I}, X_{j,0_I}, X_{j,1_I}$:

$$X_{i_1,0_I}|_{2=0_I} = X_B|_{1=0_I} \qquad X_{i_1,1_I}|_{2=0_I} = X_B|_{1=1_I}$$

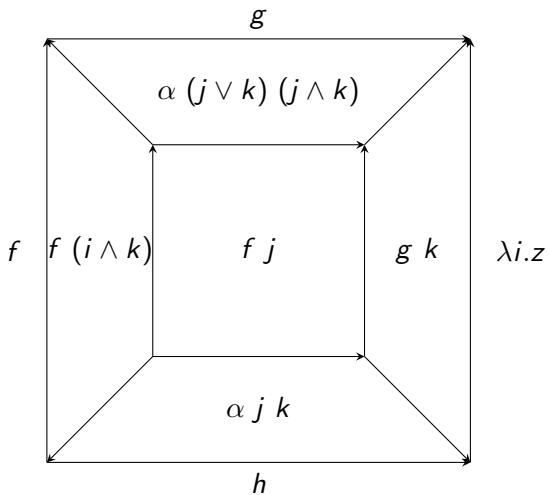
$$X_{i_2,0_I}|_{2=0_I} = X_B|_{2=0_I} \qquad X_{i_2,1_I}|_{2=0_I} = X_B|_{2=1_I}$$

$$X_{i_1,0_I}|_{1=0_I} = X_{i_2,0_I}|_{2=0_I} \qquad X_{i_1,0_I}|_{1=1_I} = X_{i_2,1_I}|_{1=0_I}$$

$$X_{i_1,1_I}|_{2=0_I} = X_{i_2,0_I}|_{2=1_I} \qquad X_{i_1,1_I}|_{2=0_I} = X_{i_2,1_I}|_{1=1_I}$$

Demo: finite-domain constraint solver to derive hcomps. Current implementation based on [Schrijvers et al., 2009].

<https://github.com/maxdore/csolver/>



Geometric proof search

Constraint satisfaction generates such derivation trees efficiently:

$$\frac{\frac{\dots}{u_{i_1} : \Theta_{i_1, 0\mathcal{I}}} \quad \frac{\dots}{v_{i_1} : \Theta_{i_1, 1\mathcal{I}}} \quad \dots \quad \frac{\dots}{u_{i_n} : \Theta_{i_n, 0\mathcal{I}}} \quad \frac{\dots}{v_{i_n} : \Theta_{i_n, 1\mathcal{I}}} \quad \frac{\dots}{w : \Psi}}{hcomp (u_{i_1}, v_{i_1}, \dots, u_{i_n}, v_{i_n}) w : \Phi}$$

To construct an open $(n + 1)$ -dimensional cube, the CSP has

- ▶ $1 + 2 \cdot n$ variables
- ▶ $2 \cdot n + 4 \cdot \binom{n}{2}$ constraints
- ▶ For any $a : \Psi$ in Γ , n -th Dedekind number ^{$\dim(\Psi)$} terms

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- ▶ For any $a : \Psi$ in Γ , n -th Dedekind number ^{$\dim(\Psi)$} terms

For $n = 5$, 7581 terms. For $n = 6$, 7828354 terms. For $n = 7$, 2414682040998 terms. For $n = 8$, ???

Next steps

- ▶ Search S more cleverly without generating full lattice.
- ▶ Implement *transp*. Also \sim , *hfill* and other non-primitives?
- ▶ Construct nested Kan composition and composed cubes.
- ▶ Support different flavours of cubical type theory.
- ▶ Implement as tactic, or combine with Agsy?

The role of cubical reasoning

We have a compelling story about path induction. Can a similarly appealing story be told for cubical Kan fillings?

- ▶ [Licata and Brunerie, 2015], [Mörtberg and Pujet, 2020], ... show that cubical reasoning allows for succinct proofs.
- ▶ Heterogeneous equality is naturally expressed with cubes.
- ▶ Most natural syntax for combinatorial reasoning about spaces?

Conclusions

First step towards mechanization of higher-dimensional type theory. Desirable for several reasons:

- ▶ Lower barrier of entry to Cubical Agda.
- ▶ Understand complexity of different type theories.
- ▶ Destill syntax agnostic to concrete type theory?

Cubical Agda is highly amenable to automation!

References



Licata, D. R. and Brunerie, G. (2015).

A cubical approach to synthetic homotopy theory.

In *2015 30th Annual ACM/IEEE Symposium on Logic in Computer Science*, pages 92–103. IEEE.



Mörtberg, A. and Pujet, L. (2020).

Cubical synthetic homotopy theory.

In *Proceedings of the 9th ACM SIGPLAN International Conference on Certified Programs and Proofs, CPP 2020*, pages 158–171, New York, NY, USA. Association for Computing Machinery.



Schrijvers, T., Stuckey, P., and Wadler, P. (2009).

Monadic constraint programming.

Journal of Functional Programming, 19(6):663–697.

Agda input

```
6 {-# OPTIONS --cubical #-}
1
2 module Cauto.Examples.Demo where
3
4 open import Cauto.Prelude
5
6 data Triangle : Set where
7   x y z : Triangle
8   f : x ≡ y
9   g : y ≡ z
10  h : x ≡ z
11  alpha : PathP (λ j → Path Triangle (f j) (h j)) (λ i → x) g
12
13 beta : PathP (λ j → Path Triangle (f j) z) h g
14 beta = ?
```

Agda output

```
clpc60[/home/scratch/maxore/csolver]$ ./csolver-exe ../Dropbox/Uni/experiments/Cauto/Examples/Demo.agda
λ i j → hcomp (λ k → λ {
  (i = i0) → alpha (j) (k)
; (i = i1) → alpha (k v j) (k ∧ j)
; (j = i0) → f (k ∧ i)
; (j = i1) → g (k)
}) (f (j))
```

Verbose output 1

```
c:\pc60[home/scratch/maxore/csolver]$ ./csolver-exe ../Dropbox/Uni/experiments/Cauto/Examples/Demo.agda --verbose
CONTEXT
x : Point
y : Point
z : Point
f : Path Point x y
g : Path Point y z
h : Path Point x z
alpha : Path (Path Point (f<[[1]]>) (h<[[1]]>)) \1.x g
GOAL
Path (Path Point (f<[[1]]>) z) h g
SHAPES
\2.\1.x : Path (Path Point x x) \1.x \1.x
\2.\1.y : Path (Path Point y y) \1.y \1.y
\2.\1.z : Path (Path Point z z) \1.z \1.z
\2.\1.(f<[[1]]>) : Path (Path Point x y) f f
\2.\1.(f<[[2]]>) : Path (Path Point (f<[[1]]>) (f<[[1]]>)) \1.x \1.y
\2.\1.(f<[[1,2]]>) : Path (Path Point x (f<[[1]]>)) \1.x f
\2.\1.(f<[[1],[2]]>) : Path (Path Point (f<[[1]]>) y) f \1.y
\2.\1.(g<[[1]]>) : Path (Path Point y z) g g
\2.\1.(g<[[2]]>) : Path (Path Point (g<[[1]]>) (g<[[1]]>)) \1.y \1.z
\2.\1.(g<[[1,2]]>) : Path (Path Point y (g<[[1]]>)) \1.y g
\2.\1.(g<[[1],[2]]>) : Path (Path Point (g<[[1]]>) z) g \1.z
\2.\1.(h<[[1]]>) : Path (Path Point x z) h h
\2.\1.(h<[[2]]>) : Path (Path Point (h<[[1]]>) (h<[[1]]>)) \1.x \1.z
\2.\1.(h<[[1,2]]>) : Path (Path Point x (h<[[1]]>)) \1.x h
\2.\1.(h<[[1],[2]]>) : Path (Path Point (h<[[1]]>) z) h \1.z
\2.\1.((alpha<[[1]]><[[1]]>) : Path (Path Point x z) \1.((alpha<[[1]]><[[1]]>) \1.((alpha<[[1]]><[[1]]>))
\2.\1.((alpha<[[1]]><[[2]]>) : Path (Path Point x (g<[[1]]>)) f h
\2.\1.((alpha<[[1]]><[[1,2]]>) : Path (Path Point x (g<[[1]]>)) f \1.((alpha<[[1]]><[[1]]>))
\2.\1.((alpha<[[1]]><[[1],[2]]>) : Path (Path Point x z) \1.((alpha<[[1]]><[[1]]>)) h
\2.\1.((alpha<[[2]]><[[1]]>) : Path (Path Point (f<[[1]]>) (h<[[1]]>)) \1.x g
\2.\1.((alpha<[[2]]><[[2]]>) : Path (Path Point ((alpha<[[1]]><[[1]]>) ((alpha<[[1]]><[[1]]>)) \1.x \1.z
\2.\1.((alpha<[[2]]><[[1,2]]>) : Path (Path Point (f<[[1]]>) ((alpha<[[1]]><[[1]]>)) \1.x g
\2.\1.((alpha<[[2]]><[[1],[2]]>) : Path (Path Point ((alpha<[[1]]><[[1]]>) (h<[[1]]>)) \1.x \1.z
\2.\1.((alpha<[[1,2]]><[[1]]>) : Path (Path Point x (h<[[1]]>)) \1.x \1.((alpha<[[1]]><[[1]]>))
\2.\1.((alpha<[[1,2]]><[[2]]>) : Path (Path Point x ((alpha<[[1]]><[[1]]>)) \1.x h
\2.\1.((alpha<[[1,2]]><[[1,2]]>) : Path (Path Point x ((alpha<[[1]]><[[1]]>)) \1.x \1.((alpha<[[1]]><[[1]]>))
\2.\1.((alpha<[[1,2]]><[[1],[2]]>) : Path (Path Point x (h<[[1]]>)) \1.x h
\2.\1.((alpha<[[1],[2]]><[[1]]>) : Path (Path Point (f<[[1]]>) z) \1.((alpha<[[1]]><[[1]]>)) g
\2.\1.((alpha<[[1],[2]]><[[2]]>) : Path (Path Point ((alpha<[[1]]><[[1]]>) (g<[[1]]>)) f \1.z
\2.\1.((alpha<[[1],[2]]><[[1,2]]>) : Path (Path Point (f<[[1]]>) (g<[[1]]>)) f g
\2.\1.((alpha<[[1],[2]]><[[1],[2]]>) : Path (Path Point ((alpha<[[1]]><[[1]]>) z) \1.((alpha<[[1]]><[[1]]>)) \1.z
NO DIRECT FIT FOUND, SEARCHING FOR HIGHER CUBES
```

Verbose output II

```
Path (Path Point (f<[[1]]>) z) h g
DOMAINS BEFORE CONSTRAINTS
i0: fromList [\2.\1.(h<[[2]]>),\2.\1.(h<[[1,2]]>),\2.\1.((alpha<[[2]]><[[1]]>),\2.\1.((alpha<[[2]]><[[1],[2]]>),\2.\1.((alpha<[[1,2]]><[[1]]>),\2.\1.((alpha<[[1,2]]><[[1],[2]]>),\2.\1.((alpha<[[1,2]]><[[1],[2]]>)]
j0: fromList [\2.\1.(f<[[2]]>),\2.\1.(f<[[1,2]]>)]
i1: fromList [\2.\1.(g<[[2]]>),\2.\1.(g<[[1,2]]>),\2.\1.((alpha<[[1]]><[[2]]>),\2.\1.((alpha<[[1]]><[[1,2]]>),\2.\1.((alpha<[[1],[2]]><[[2]]>),\2.\1.((alpha<[[1],[2]]><[[1,2]]>)]
j1: fromList [\2.\1.z,\2.\1.(g<[[1]]>),\2.\1.(g<[[1],[2]]>),\2.\1.(h<[[1]]>),\2.\1.(h<[[1],[2]]>),\2.\1.((alpha<[[1]]><[[1]]>),\2.\1.((alpha<[[1]]><[[1],[2]]>),\2.\1.((alpha<[[1],[2]]><[[1]]>),\2.\1.((alpha<[[1],[2]]><[[1],[2]]>)]
]
DOMAINS AFTER BACK CONSTRAINTS
fromList [\2.\1.x,\2.\1.(f<[[1]]>),\2.\1.(f<[[2]]>),\2.\1.(f<[[1,2]]>),\2.\1.(f<[[1],[2]]>),\2.\1.((alpha<[[1]]><[[1]]>),\2.\1.((alpha<[[1]]><[[1,2]]>),\2.\1.((alpha<[[2]]><[[1]]>),\2.\1.((alpha<[[2]]><[[1],[2]]>),\2.\1.((alpha<[[2]]><[[1],[2]]>),\2.\1.((alpha<[[1,2]]><[[1]]>),\2.\1.((alpha<[[1,2]]><[[1],[2]]>)]
]
i0: fromList [\2.\1.(h<[[2]]>),\2.\1.(h<[[1,2]]>),\2.\1.((alpha<[[2]]><[[1]]>),\2.\1.((alpha<[[2]]><[[1],[2]]>),\2.\1.((alpha<[[1,2]]><[[1]]>),\2.\1.((alpha<[[1,2]]><[[1],[2]]>)]
j0: fromList [\2.\1.(f<[[2]]>),\2.\1.(f<[[1,2]]>)]
i1: fromList [\2.\1.(g<[[2]]>),\2.\1.(g<[[1,2]]>),\2.\1.((alpha<[[1]]><[[2]]>),\2.\1.((alpha<[[1]]><[[1,2]]>),\2.\1.((alpha<[[1],[2]]><[[2]]>),\2.\1.((alpha<[[1],[2]]><[[1,2]]>)]
j1: fromList [\2.\1.z,\2.\1.(g<[[1]]>),\2.\1.(g<[[1],[2]]>),\2.\1.(h<[[1]]>),\2.\1.(h<[[1],[2]]>),\2.\1.((alpha<[[1]]><[[1]]>),\2.\1.((alpha<[[1]]><[[1],[2]]>),\2.\1.((alpha<[[1],[2]]><[[1]]>),\2.\1.((alpha<[[1],[2]]><[[1],[2]]>)]
]
DOMAINS AFTER SIDE CONSTRAINTS
fromList [\2.\1.(f<[[1]]>),\2.\1.(f<[[1,2]]>)]
i0: fromList [\2.\1.((alpha<[[2]]><[[1]]>),\2.\1.((alpha<[[1,2]]><[[1]]>)]
j0: fromList [\2.\1.(f<[[1,2]]>)]
i1: fromList [\2.\1.((alpha<[[1],[2]]><[[1,2]]>)]
j1: fromList [\2.\1.(g<[[1]]>),\2.\1.((alpha<[[1],[2]]><[[1]]>)]
[Comp \2.\1.(f<[[1]]>) [\2.\1.((alpha<[[2]]><[[1]]>),\2.\1.((alpha<[[1],[2]]><[[1,2]]>)), (\2.\1.(f<[[1,2]]>),\2.\1.(g<[[1]]>))]
λ i j → hcomp (λ k → λ {
  (i = i0) → alpha (j) (k)
  ; (i = i1) → alpha (k v j) (k λ j)
  ; (j = i0) → f (k λ i)
  ; (j = i1) → g (k)
}) (f (j))
```

Evaluation

$$r[i = 0_{\mathcal{I}}] \hookrightarrow \begin{cases} 0_{\mathcal{I}} & \text{if } i \in c \text{ for all } c \in r \\ \{c \mid i \notin c\} & \text{otherwise} \end{cases}$$

$$r[i = 1_{\mathcal{I}}] \hookrightarrow \begin{cases} 1_{\mathcal{I}} & \text{if } c = \{i\} \text{ for some } c \in r \\ \{c' \mid c' = c \setminus \{i\} \text{ for } c \in r\} & \text{otherwise} \end{cases}$$

$$a[i = e] \hookrightarrow a$$

$$(\lambda j. t)[i = e] \hookrightarrow \lambda j. t[i = e]$$

$(i \neq j)$

$$(t \ r)[i = e] \hookrightarrow \begin{cases} u & \text{if } r[i = e] \hookrightarrow 0_{\mathcal{I}} \text{ and } t : \text{PathP } (\lambda k. \Phi) \ u \ v \\ v & \text{if } r[i = e] \hookrightarrow 1_{\mathcal{I}} \text{ and } t : \text{PathP } (\lambda k. \Phi) \ u \ v \\ (t[i = e]) \ (r[i = e]) & \text{otherwise} \end{cases}$$

$(i \neq j)$