

# Relational Algebra for “Just Good Enough” Hardware

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# Motivation

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machine design hyd pne

**medical design**

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**R&D Notebook: The growing importance of software verification and validation in medical device design**

The third edition of IEC 60601-1 takes on a new role, bringing risk management into the very first stages of the product development process.

compared with...

# Motivation

## MITnews

engineering science management architecture + planning humanities, arts, and social sciences campus

### The surprising usefulness of sloppy arithmetic

A computer chip that performs imprecise calculations could process some types of data thousands of times more efficiently than existing chips.

Larry Hardesty, MIT News Office

**Sloppy** arithmetic useful?

Horror!

But there is more. . .

## “Just good enough” $h/w$

... coming from the land of the Swiss watch:

### “We should stop designing perfect circuits”



02.10.13 - Are integrated circuits "too good" for current technological applications? Christian Enz, the new Director of the Institute of Microengineering, backs the idea that perfection is overrated.

Message:

*Why perfection if (some) imperfection still meets the standards?*

## S/w for “just good enough” h/w

What about **software** running over “just good enough” hardware?

Ready to **take the risk**?

Nonsense to run **safety critical** software on **defective** hardware?

Uups! — it seems “it already runs”:

**medical design** *“IEC 60601-1 [brings] **risk management** into the very first stages of [product development]”*

**Risk** is everywhere — an inevitable (desired?) part of life.

# P(robabilistic)R(isk)A(nalysis)

NASA/SP-2011-3421 (Stamatelatos and Dezfuli, 2011):

*1.2.2 A PRA characterizes risk in terms of three basic questions: (1) What can **go wrong**? (2) How **likely** is it? and (3) What are the **consequences**?*

The PRA process

*answers these questions by systematically (...) identifying, modeling, and **quantifying** scenarios that can lead to undesired consequences*

Interestingly,

**medical**  
**design**

*"IEC 60601-1 [...] **very first** stages of [development]"*

## From the very first stage in development

Think of things that **can go wrong**:

$bad \cup good$

How **likely**?

$bad \text{ }_p\text{ } \diamond \text{ } good$  (1)

where

$bad \text{ }_p\text{ } \diamond \text{ } good = p \times bad + (1 - p) \times good$

for some **probability**  $p$  of *bad behaviour*, eg. the **imperfect** action

$top \text{ }_{(10^{-7})}\text{ } \diamond \text{ } pop$

leaving a stack unchanged with  $10^{-7}$  probability.



## Imperfect truth tables

Imperfect **negation**  $id_{0.01} \diamond neg$ :

$$\begin{aligned}
 & id_{0.01} \diamond neg \\
 = & 0.01 \times \left( \begin{array}{c|cc} & \mathbf{False} & \mathbf{True} \\ \hline \mathbf{False} & 1 & 0 \\ \mathbf{True} & 0 & 1 \end{array} \right) + 0.99 \times \left( \begin{array}{c|cc} & \mathbf{False} & \mathbf{True} \\ \hline \mathbf{False} & 0 & 1 \\ \mathbf{True} & 1 & 0 \end{array} \right) \\
 = & \left( \begin{array}{c|cc} & \mathbf{False} & \mathbf{True} \\ \hline \mathbf{False} & 0.01 & 0 \\ \mathbf{True} & 0 & 0.01 \end{array} \right) + \left( \begin{array}{c|cc} & \mathbf{False} & \mathbf{True} \\ \hline \mathbf{False} & 0 & 0.99 \\ \mathbf{True} & 0.99 & 0 \end{array} \right) \\
 = & \begin{array}{c|cc} & \mathbf{False} & \mathbf{True} \\ \hline \mathbf{False} & 0.01 & 0.99 \\ \mathbf{True} & 0.99 & 0.01 \end{array}
 \end{aligned}$$

## Functions? Relations? Yes: **matrices!**

Better than the “anything can happen” **relation**  $id \cup neg$ , **matrix**  $id \rho \diamond neg$  carries useful **quantitative** information.

Aside: fragment of **function**  $pres : President \rightarrow Country$  displayed as a **matrix** in the **Relational Mathematics** book (Schmidt, 2010).

**Relational** and **linear algebra** (LA) share a lot in common.

LA required when calculating **risk** of failure of **safety critical** s/w.

	US	French	German	British	Spanish
Clinton	1	0	0	0	0
Bush	1	0	0	0	0
Mitterand	0	1	0	0	0
Chirac	0	1	0	0	0
Schmidt	0	0	1	0	0
Kohl	0	0	1	0	0
Schröder	0	0	1	0	0
Thatcher	0	0	0	1	0
Major	0	0	0	1	0
Blair	0	0	0	1	0
Zapatero	0	0	0	0	1

# Linear algebra of programming

**Relational** / **KAT** algebra — a success story.

Linear algebra of programming (**LAoP**)  
— research track aiming at a  
quantitative extension of  
**heterogeneous** relational/KAT algebra.

Keeping the **pointfree** style!

Strategy: mild and pragmatic use of  
**categorical** techniques.

Main point — **Kleisli** categories matter!



Heinrich Kleisli  
(1930-2011)

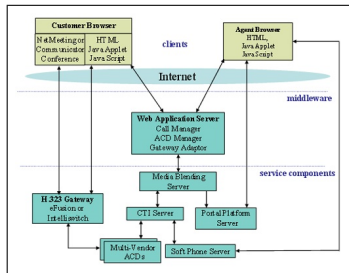
# Context

# Faults in CBS systems

Interested in reasoning about the risk of **faults propagating** in **component-based software (CBS)** systems.

Traditional CBS **risk analysis** relies on *semantically weak* CBS models, e.g. component **call-graphs** (Cortellessa and Grassi, 2007).

Our starting point is a **coalgebraic** semantics for s/w components modeled as **monadic Mealy machines** (Barbosa and Oliveira, 2006).



## Main ideas

**Component** = Monadic Mealy machine (MMM), that is, an  $\mathbb{F}$ -evolving transition structure of type:

$$S \times I \rightarrow \mathbb{F}(S \times O)$$

where  $\mathbb{F}$  is a **monad**.

**Method** = Elementary (single action) MMM.

**CBS design** = Algebra of MMM combinators.

**Semantics** = Coalgebraic, calculational.

To this framework we want to add analysis of

**Risk** = Probability of **faulty** (catastrophic) behaviour

# Mealy machines in various guises

$\mathbb{F}$ -transition structure:

$$S \times I \rightarrow \mathbb{F}(S \times O)$$

Coalgebra:

$$S \rightarrow (\mathbb{F}(S \times O))^I$$

State-monadic:

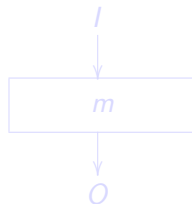
$$I \rightarrow (\mathbb{F}(S \times O))^S$$

All versions useful in  
component algebra.

Abstracting from internal  
state  $S$  and branching effect  
 $\mathbb{F}$ , machine

$$m : S \times I \rightarrow \mathbb{F}(S \times O)$$

can be depicted as



or as the **arrow**  $I \xrightarrow{m} O$ .

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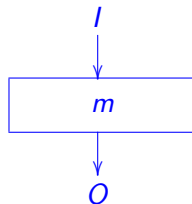
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or as the **arrow**  $I \xrightarrow{m} O$ .



## Example — stack component

From a **(partial)** algebra of finite lists (Haskell syntax)

<b>(partial) function</b>	<b>type</b>
$push\ s\ a = a : s$	$push :: ([a], a) \rightarrow [a]$
$pop = tail$	$pop :: [a] \rightarrow [a]$
$top = head$	$top :: [a] \rightarrow a$
$empty\ s = (length\ s = 0)$	$empty :: [a] \rightarrow \mathbb{B}$

to a collection of **(total)** methods (MMMs):

<b>method</b>	<b>type</b>
$push' = \eta \cdot (push \triangleleft !)$	$push' :: ([a], a) \rightarrow \mathbb{M}([a], 1)$
$pop' = (pop \triangleleft top \leftarrow (\neg \cdot empty)) \cdot fst$	$pop' :: ([a], 1) \rightarrow \mathbb{M}([a], a)$
$top' = (id \triangleleft top \leftarrow (\neg \cdot empty)) \cdot fst$	$top' :: ([a], 1) \rightarrow \mathbb{M}([a], a)$
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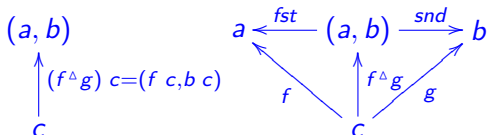
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where...

# Explanation

Pairing:



“Sink” (“bang”) function  $A \xrightarrow{!} 1$  onto singleton type  $1$

$\mathbb{M}$  : Monad with unit  $\eta$  and zero  $\perp$  (typically **Maybe**)

$\mathbb{M}$  -totalizer on given pre-condition:

$$\cdot \Leftarrow \cdot :: (a \rightarrow b) \rightarrow (a \rightarrow \mathbb{B}) \rightarrow a \rightarrow \mathbb{M} b$$

$$(f \Leftarrow p) a = \text{if } p a \text{ then } (\eta \cdot f) a \text{ else } \perp$$

# Component = $\sum$ methods

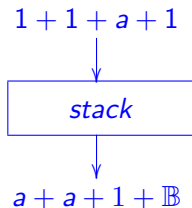
Define

$$\begin{aligned} \text{stack} &:: ([a], 1 + 1 + a + 1) \rightarrow \mathbb{M}([a], a + a + 1 + \mathbb{B}) \\ \text{stack} &= \text{pop}' \oplus \text{top}' \oplus \text{push}' \oplus \text{empty}' \end{aligned}$$

to obtain a **compound**  $\mathbb{M}$ -MM (stack **component**) with 4 methods, where

- **input** 1 means “DO IT!”
- **output** 1 means “DONE!”

Notation  $m \oplus n$  expresses the “coalesced” **sum** of two state-compatible MMMs (next slide).



# Machine sums

Note the **pointfree** definition

```

· ⊕ · :: (Functor F) =>
  -- input machines
  ((s, i) → F (s, o)) →
  ((s, j) → F (s, p)) →
  -- output machine
  (s, i + j) → F (s, o + p)
  -- definition
  m1 ⊕ m2 = (F dr◦) · Δ · (m1 + m2) · dr

```

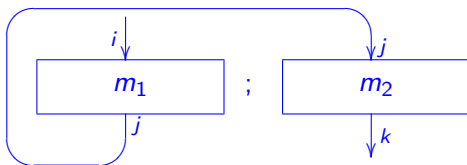
where  $dr^\circ$  is the converse of isomorphism

$$dr :: (s, i + j) \rightarrow (s, i) + (s, j)$$

and  $\Delta :: F a + F b \rightarrow F (a + b)$  is a kind of “cozip” operator.

# S/w system = component **composition**

## Forward **composition**



is central to component **communication**.

Abstracting from state, it means **composition** in a categorical sense:

$$i \xrightarrow{m_1} j \xrightarrow{m_2} k$$

$m_2 \cdot m_1$

The diagram illustrates the composition of two morphisms in a categorical sense. It shows a sequence of two arrows:  $i \xrightarrow{m_1} j$  followed by  $j \xrightarrow{m_2} k$ . A curved arrow above the sequence represents the composite morphism  $m_2 \cdot m_1$  from  $i$  to  $k$ .

# Exchange law

Formal definition of  $m ; n$  to be discussed shortly.

For suitably typed MMM  $m_1$  ,  $m_2$  ,  $n_1$  and  $n_2$  , mind the useful **exchange law**

$$(m_1 \oplus m_2) ; (n_1 \oplus n_2) = (m_1 ; n_1) \oplus (m_2 ; n_2) \quad (2)$$

expressing two alternative approaches to s/w system construction:

- $\cdot \oplus \cdot$  -first — “component-oriented”
- $\cdot ; \cdot$  -first — “method-oriented”

---

For several other combinators in the algebra see (Barbosa and Oliveira, 2006).

## Simulation (Haskell)

Let  $M$  instantiate to Haskell's **Maybe** monad:

- Running a **perfect** and **successful** composition:

```
> (pop' ; push') (([1], [2]), ())  
Just (([], [1, 2]), ())
```

- Running a **perfect** but **catastrophic** composition:

```
> (pop' ; push') (([], [2]), ())  
Nothing
```

(source stack empty)

What about **imperfect** machine communication?



## Imperfect components

Risk of  $pop'$  behaving like  $top'$  with **probability**  $1 - p$  :

$$pop'' :: \mathbb{P} \rightarrow ([a], 1) \rightarrow \mathbb{D} (\mathbb{M} ([a], a))$$

$$pop'' \ p = pop' \ p \diamond top'$$

Risk of  $push'$  not pushing anything, with probability  $1 - q$ :

$$push'' :: \mathbb{P} \rightarrow ([a], a) \rightarrow \mathbb{D} (\mathbb{M} ([a], 1))$$

$$push'' \ q = push' \ q \diamond !$$

---

Details:  $\mathbb{P} = [0, 1]$ ,  $\mathbb{D}$  is the (finite) **distribution** monad and

$$\cdot \diamond \cdot :: \mathbb{P} \rightarrow (t \rightarrow a) \rightarrow (t \rightarrow a) \rightarrow t \rightarrow \mathbb{D} \ a$$

$$(f \ p \diamond \ g) \ x = choose \ p \ (f \ x) \ (g \ x)$$

chooses between  $f$  and  $g$  according to  $p$  .

# Faulty components

Define

$$m_2 = \text{pop}'' \ 0.95 ;_D \ \text{push}'' \ 0.8$$

where  $\cdot ;_D \cdot$  is a **probabilistic** enrichment of **composition** and run the same simulations for  $m_2$  over the same state  $([1], [2])$ :

```
> m2 (([1], [2]), ())
Just (([], [1, 2]), ()) 76.0 %
  Just (([], [2]), ()) 19.0 %
  Just (([1], [1, 2]), ()) 4.0 %
  Just (([1], [2]), ()) 1.0 %
```

---

**Total risk** of faulty behaviour is 24% ( $1 - 0.76$ ) structured as:

- (a) 1% — both stacks misbehave; (b) 19% — target stack misbehaves;
- (c) 4% — source stack misbehaves.

## Faulty components

As expected, the behaviour of

```
> m2 ([], [2]), ()  
Nothing 100.0 %
```

is 100% **catastrophic** (popping from an empty stack).

---

Simulation details:

*Using the **PFP library** written in Haskell by Erwig and Kollmansberger (2006).*

## Central topic

Our MMMs have become **probabilistic**, acquiring the general shape

$$S \times I \rightarrow \mathbb{D} (\mathbb{F} (S \times O))$$

where the additional  $\mathbb{D}$  — (finite support) **distribution** monad — captures **imperfect** behaviour (fault propagation).

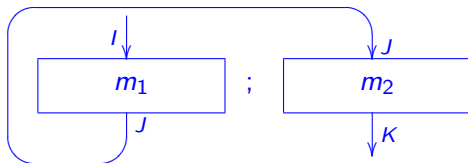
### Questions:

- Shall we compose  $\mathbb{D} \cdot \mathbb{F}$  and work over the **composite** monad?
- Or shall we try and find a way of working “as if  $\mathbb{D}$  wasn't there”?

Let us first see how MMM compose.

# MMM forward composition

## Combinator



is defined by **Kleisli** composition

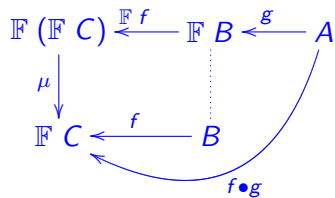
$$m_1 ; m_2 = (\psi m_2) \bullet (\phi m_1)$$

of two steps:

- $\phi m_1$  — run  $m_1$  “wrapped” with the state of  $m_2$
- $\psi m_2$  — run  $m_2$  “wrapped” with that of  $m_1$  for the output it delivers

# Kleisli composition

Let  $X \xrightarrow{\eta} \mathbb{F}X \xleftarrow{\mu} \mathbb{F}^2X$  be a **monad** in diagram



$f \bullet g$  denotes the so-called **Kleisli composition** of  $\mathbb{F}$ -resultric arrows, forming a monoid with  $\eta$  as identity:

$$f \bullet (g \bullet h) = (f \bullet g) \bullet h$$

$$f \bullet \eta = f = \eta \bullet f$$

# MMM composition — part I

Given  $I \xrightarrow{m_1} J$  build  $\phi m_1$  :

$$\begin{array}{ccccc}
 \mathbb{F}(S \times J) \times Q & \xleftarrow{m_1 \times id} & (S \times I) \times Q & \xleftarrow{xr} & (S \times Q) \times I \\
 \tau_r \downarrow & & & \swarrow \phi m_1 & \\
 \mathbb{F}((S \times J) \times Q) & & & & 
 \end{array}$$

where

- $xr : (S \times Q) \times I \rightarrow (S \times I) \times Q$  is the obvious **isomorphism** ensuring the compound state and input  $I$
- $\tau_r : (\mathbb{F} A) \times B \rightarrow \mathbb{F}(A \times B)$  is the right **strength** of monad  $\mathbb{F}$ , which therefore has to be a **strong** monad.

# MMM composition — part II

Given  $J \xrightarrow{m_2} K$  build  $\psi m_2$  :

$$\begin{array}{ccccc}
 S \times \mathbb{F}(Q \times K) & \xleftarrow{id \times m_2} & S \times (Q \times J) & \xleftarrow{x_l} & (S \times J) \times Q \\
 \downarrow \tau_l & & & \swarrow \psi m_2 & \\
 \mathbb{F}(S \times (Q \times K)) & & & & \\
 \downarrow \mathbb{F} a^\circ & & & & \\
 \mathbb{F}((S \times Q) \times K) & & & & 
 \end{array}$$

where

- $a^\circ$  is the converse of **isomorphism**  $a : (A \times B) \times C \rightarrow A \times (B \times C)$
- $x_l$  is a variant of  $x_r$
- $\tau_l : (B \times \mathbb{F} A) \rightarrow \mathbb{F}(B \times A)$  is the **left** strength of  $\mathbb{F}$ .



# MMM composition — part III

Finally build  $m_1 ; m_2 = (\psi m_2) \bullet (\phi m_1)$  :

$$\begin{array}{ccccc}
 \mathbb{F}(\mathbb{F}((S \times Q) \times K)) & \xleftarrow{\mathbb{F}(\psi m_2)} & \mathbb{F}((S \times J) \times Q) & \xleftarrow{\phi m_1} & \mathbb{F}((S \times Q) \times I) \\
 \downarrow \mu & & \vdots & & \downarrow \\
 \mathbb{F}((S \times Q) \times K) & \xleftarrow{\psi m_2} & (S \times J) \times Q & & \\
 & \swarrow & & \searrow & \\
 & & & & m_1 ; m_2
 \end{array}$$

This for **perfect**  $\mathbb{F}$ -monadic machines. What about the **imperfect** ones?

*What is the impact of adding probability-of-fault to the above construction? Does one need to rebuild the definition?*

## MMM composition — part III

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 \downarrow \mu & & \vdots & & \searrow \\
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 \end{array}$$

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*What is the impact of adding probability-of-fault to the above construction? Does one need to rebuild the definition?*

## Doubly-monadic machines

Recall Haskell simulations running combinator  $m_1 ;_D m_2$  for doubly-monadic machines of type

$$(S \times I) \rightarrow \mathbb{D} (\mathbb{M} (S \times O))$$

involving the Maybe ( $\mathbb{M}$ ) and (finite support) **distribution** ( $\mathbb{D}$ ) monads which generalize to

$$(S \times I) \rightarrow \mathbb{G} (\mathbb{F} (S \times O))$$

where, following the terminology of Hasuo et al. (2007):

- monad  $X \xrightarrow{\eta_F} \mathbb{F} X \xleftarrow{\mu_F} \mathbb{F}^2 X$  caters for **transitional effects** (how the machine evolves)
- monad  $X \xrightarrow{\eta_G} \mathbb{G} X \xleftarrow{\mu_G} \mathbb{G}^2 X$  specifies the **branching type** of the system.

# Going relational

# Doubly-monadic machines

Typical instance:

---

$\mathbb{G} = \mathbb{P}$  (powerset) and  $\mathbb{F} = \mathbb{M} = (1+)$  ('maybe'), that is,  
 $m : Q \times I \rightarrow \mathbb{P}(1 + Q \times J)$

is a reactive, **non-deterministic** finite state automaton  
 with explicit termination.

---

Such machines can be regarded as **binary** relations of (relational) type

$$(Q \times I) \rightarrow (1 + Q \times J)$$

and handled directly in **relational algebra**. (Details in the next slide)

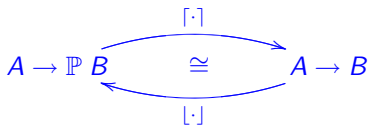
# Nondeterministic Maybe machines

The **power** transpose adjunction

$$R = [m] \quad \Leftrightarrow \quad \langle \forall b, a :: b R a = b \in m a \rangle$$

for trading between  $\mathbb{P}$ -**functions**  
and **binary relations**, in a way  
such that

$$[m \bullet n] = [m] \cdot [n]$$



where

- $m \bullet n$  — Kleisli composition of  $\mathbb{P}$ -functions
- $[m] \cdot [n]$  — **relational** composition

$$b (R \cdot S) a \quad \Leftrightarrow \quad \langle \exists c :: b R c \wedge c S a \rangle$$

of the corresponding *binary relations*.

# Composing relational $\mathbb{M}$ -machines

Transition monad on duty is  $\mathbb{M} = (1+)$  , ie.

$$X \xrightarrow{i_2} 1 + X \xleftarrow{[i_1, id]} 1 + (1 + X)$$

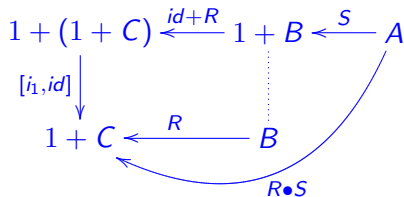
( $i_1$  ,  $i_2$  = binary sum injections).

**Lifting**: in the original definition

$$m_1 ; m_2 = (\psi m_2) \bullet (\phi m_1)$$

run **Kleisli** composition **relationally**:

$$\begin{aligned} R \bullet S &= [i_1, id] \cdot (id + R) \cdot S = [i_1, R] \cdot S \\ &= i_1 \cdot i_1^\circ \cdot S \cup R \cdot i_2^\circ \cdot S \end{aligned}$$



## Composing relational $\mathbb{M}$ -machines

Pointwise:  $y (R \bullet S) a$  holds iff

$$(y = *) \wedge (* S a) \vee \langle \exists c :: (y R c) \wedge ((i_2 c) S a) \rangle$$

where  $* = i_1 \perp$

In words:

*$R \bullet S$  doomed to fail if  $S$  fails;*

*Otherwise,  $R \bullet S$  will fail where  $R$  fails.*

*For the same input,  $R \bullet S$  may both succeed or fail.*

---

**Summary:** Nondeterministic  $\mathbb{M}$ -machines are  $\mathbb{M}$ -**relations** and original (deterministic) definition is “reused” in the relational setting:

$$R_1 ; R_2 = (\psi R_2) \bullet (\phi R_1) = [i_1, \psi R_2] \cdot (\phi R_1)$$



# Going linear

# Probabilistic branching ( $\mathbb{D}$ instead of $\mathbb{P}$ )

Again, instead of working in  $\text{Set}$ ,

$$\begin{array}{ccc} & \mathbb{D}(\mathbb{F} B) & \xleftarrow{g} A \\ & \vdots & \\ \mathbb{D}(\mathbb{F} C) & \xleftarrow{f} & B \end{array}$$

we seek to implement  $\mathbb{F}$ -Kleisli-composition in the Kleisli category of  $\mathbb{D}$ , that is

$$\begin{array}{ccc} & & \mathbb{F} B \xleftarrow{[g]} A \\ & \text{[f]}\bullet\text{[g]} & \searrow \\ & & \vdots \\ \mathbb{F} C & \xleftarrow{[f]} & B \end{array}$$

thus “abstracting from” monad  $\mathbb{D}$ .

**Question:**  $\text{Kleisli}(\mathbb{D}) = ??$

## Probabilistic monadic machines

It turns out to be the (monoidal) category of column-stochastic (CS) **matrices**, cf. adjunction

$$A \rightarrow_{\text{Set}} \mathbb{D}B \begin{array}{c} \xrightarrow{[\cdot]} \\ \cong \\ \xleftarrow{[\cdot]} \end{array} A \rightarrow_{\text{CS}} B$$

such that

$$M = [f] \quad \Leftrightarrow \quad \langle \forall b, a :: b M a = (f a) b \rangle$$

where  $A \rightarrow_{\text{CS}} B$  is the **matrix type** of all matrices with  $B$ -indexed rows and  $A$ -indexed columns all adding up to 1 (100%).

**Important:**

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$\text{CS}$  represents the **Kleisli** category of  $\mathbb{D}$

---

## Probabilism versus matrix algebra

Recall probabilistic **negation** function

$$f = id_{0.1} \diamond (\neg)$$

which corresponds to matrix

$$[f] = \begin{array}{cc} & \begin{array}{cc} \text{True} & \text{False} \end{array} \\ \begin{array}{c} \text{True} \\ \text{False} \end{array} & \begin{pmatrix} 0.1 & 0.9 \\ 0.9 & 0.1 \end{pmatrix} \end{array}$$

where **probabilistic choice** is immediate on the matrix side,

$$[f \diamond g] = p [f] + (1 - p) [g]$$

where (+) denotes **addition** of matrices of the same **type**.

# Typed linear algebra

In general, category of matrices over a semi-ring  $(\mathbb{S}; +, \times, 0, 1)$ :

- **Objects** are types  $(A, B, \dots)$  and **morphisms**  $(M : A \rightarrow B)$  are matrices whose columns have finite support.
- **Composition:**

$$\begin{array}{ccccc}
 B & \xleftarrow{M} & A & \xleftarrow{N} & C \\
 & \searrow & \swarrow & \nearrow & \\
 & & C=M \cdot N & & 
 \end{array}$$

that is:

$$b(M \cdot N)c = \langle \sum a :: (rMa) \times (aNc) \rangle$$

- **Identity:** the diagonal Boolean matrix  $id : A \rightarrow A$ .

# Typed linear algebra

## Matrix **coproducts**

---


$$(A + B) \rightarrow C \cong (A \rightarrow C) \times (B \rightarrow C)$$


---

where  $A + B$  is disjoint union, cf. **universal property**

$$X = [M|N] \Leftrightarrow X \cdot i_1 = M \wedge X \cdot i_2 = N$$

where  $[i_1|i_2] = id$ .

$[M|N]$  is one of the basic matrix **block** combinators — it puts  $M$  and  $N$  side by side and is such that

$$[M|N] = M \cdot i_1^\circ + N \cdot i_2^\circ$$

as in **relation algebra**.

# Typed linear algebra

## Matrix **direct sum**

$$M \oplus N = \left[ \begin{array}{c|c} M & 0 \\ \hline 0 & N \end{array} \right]$$

is an (endo,bi)functor, cf.

$$\begin{aligned} (id \oplus id) &= id \\ (M \oplus N) \cdot (P \oplus Q) &= (M \cdot P) \oplus (N \cdot Q) \\ [M|N] \cdot (P \oplus Q) &= [M \cdot P|N \cdot Q] \end{aligned}$$

as in **relation algebra** — etc, etc.

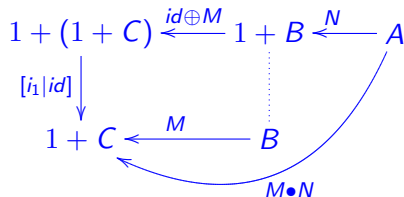
The Maybe monad in the category is therefore given by

$$\mathbb{M} = (id \oplus \cdot)$$

## Another “Kleisli shift”

As we did for relations representing  $\text{Kleisli}(\mathbb{P})$ , let us encode  $M$ -Kleisli composition in matrix form:

$$M \bullet N = [i_1 | M] \cdot N$$



Thus  $M \bullet N = i_1 \cdot i_1^\circ \cdot N + M \cdot i_2^\circ \cdot N$  leading into the pointwise

$$y (M \bullet N) a = (y = *) \times (* N a) + \langle \sum b :: (y M b) \times ((i_2 b) N a) \rangle$$

— compare with the relational version and example (next slide).



## Another “Kleisli shift”

Example:

Probabilistic  $\mathbb{M}$ -Kleisli  
composition  $M \bullet N$  of matrices  
 $N : \{a_1, a_2, a_3\} \rightarrow 1 + \{c_1, c_2\}$   
and  
 $M : \{c_1, c_2\} \rightarrow 1 + \{b_1, b_2\}$ .

Injection  $i_1 : 1 \rightarrow 1 + \{b_1, b_2\}$   
is the leftmost column vector.

				a1	a2	a3
			*	0.5	0	0
			c1	0.5	1	0.7
		*	c1	0	0	0.3
*	1	0.2	0	0.6	0.2	0.14
b1	0	0	0.6	0	0	0.18
b2	0	0.8	0.4	0.4	0.8	0.68

Example: for input  $a_1$  there is 60% probability of  $M \bullet N$  failing = either  $N$  fails (50%) or passes  $c_1$  to  $M$  (50%) which fails with 20% probability.

## Probabilistic MMM (=pMMM) as matrices

Similarly to relations before, we can think of **probabilistic**  $\mathbb{M}$ -monadic Mealy machines as CS matrices which communicate (as matrices) as follows

$$N ; M = [i_1 | (id \oplus a^\circ) \cdot \tau_l \cdot (id \otimes M) \cdot x_l] \cdot \tau_r \cdot (N \otimes id) \cdot x_r \quad (3)$$

where

- **functions** are represented matricially by **Dirac** distributions;
- relational product becomes matrix **Kronecker product**

$$(y, x)(M \otimes N)(b, a) = (yMb) \times (xNa)$$

**NB:** Haskell implementation of pMMM composition follows (3).

# Kleisli shift

## Monad-monad lifting

For the above to make sense for machines of **generic** type  $Q \times I \rightarrow \mathbb{G} (\mathbb{F} (Q \times J))$  make sure that

---

*The lifting of monad  $\mathbb{F}$  by monad  $\mathbb{G}$  still is a monad in the Kleisli category of  $\mathbb{G}$ .*

---

Recall:

- $\mathbb{F}$  — **transition** monad
- $\mathbb{G}$  — **branching** monad

Mind their different roles:

---

**Branching** monad “hosts” **transition** monad.

---

# Monad-monad lifting

In general, given two monads

$$X \xrightarrow{\eta_G} \mathbb{G}X \xleftarrow{\mu_G} \mathbb{G}^2X \quad (\text{the host})$$

$$X \xrightarrow{\eta_F} \mathbb{F}X \xleftarrow{\mu_F} \mathbb{F}^2X \quad (\text{the guest})$$

in a category  $\mathbf{C}$  :

- let  $\mathbf{C}^b$  denote the Kleisli category induced by host  $\mathbb{G}$ ;
- let  $B \xleftarrow{f^b} A$  be the morphism in  $\mathbf{C}^b$  corresponding to  $\mathbb{G}B \xleftarrow{f} A$  in  $\mathbf{C}$  ;
- define

$$f^b \cdot g^b = (f \bullet g)^b = (\mu_G \cdot \mathbb{G} f \cdot g)^b$$

## Monad-monad lifting

For *any* morphism  $B \xleftarrow{f} A$  in  $\mathbf{C}$  define its lifting to  $\mathbf{C}^b$  by

$$\bar{f} = (\eta_G \cdot f)^b \quad (4)$$

As in (Hasuo et al., 2007), assume **distributive law**

$$\lambda : FG \rightarrow GF$$

Lift the **guest** endofunctor  $\mathbb{F}$  from  $\mathbf{C}$  to  $\mathbf{C}^b$  by defining  $\bar{\mathbb{F}}$  as follows, for  $G B \xleftarrow{f} A$  :

$$\bar{\mathbb{F}}(f^b) = (\lambda \cdot \mathbb{F} f)^b$$

cf. diagram

$$GFB \xleftarrow{\lambda} FGB \xleftarrow{\mathbb{F} f} FA$$

## Monad-monad lifting

For  $\overline{\mathbb{F}}$  to be a **functor** in  $\mathbf{C}^b$  two conditions must hold (Hasuo et al., 2007):

$$\lambda \cdot \mathbb{F} \eta_G = \eta_G$$

$$\lambda \cdot \mathbb{F} \mu_G = \mu_G \cdot \mathbb{G} \lambda \cdot \lambda$$

We need to find extra conditions for *guest*  $\mathbb{F}$  to lift to a **monad** in  $\mathbf{C}^b$ ; that is,

$$X \xrightarrow{\overline{\eta_{\mathbb{F}}} = (\eta_G \cdot \eta_{\mathbb{F}})^b} \overline{\mathbb{F}} X \xleftarrow{\overline{\mu_{\mathbb{F}}} = (\eta_G \cdot \mu_{\mathbb{F}})^b} \overline{\mathbb{F}}^2 X$$

should be a monad in  $\mathbf{C}^b$ .

The standard monadic laws, e.g.  $\overline{\mu_{\mathbb{F}}} \cdot \overline{\eta_{\mathbb{F}}} = id$ , hold via lifting (4) and Kleisli composition laws.

# Monad-monad lifting

The remaining **natural** laws,

$$(\overline{\mathbb{F}} f^b) \cdot \overline{\eta}_{\mathbb{F}} = \overline{\eta}_{\mathbb{F}} \cdot f^b$$

$$(\overline{\mathbb{F}} f^b) \cdot \overline{\mu}_{\mathbb{F}} = \overline{\mu}_{\mathbb{F}} \cdot (\overline{\mathbb{F}}^2 f^b)$$

are ensured by two “monad-monad” compatibility conditions:

$$\lambda \cdot \eta_{\mathbb{F}} = \mathbb{G}\eta_{\mathbb{F}}$$

$$\lambda \cdot \mu_{\mathbb{F}} = \mathbb{G}\mu_{\mathbb{F}} \cdot \lambda \cdot \mathbb{F}\lambda$$

that is:

$$\begin{array}{ccccc}
 \mathbb{G}X & \xrightarrow{\eta_{\mathbb{F}}} & \mathbb{F}\mathbb{G}X & \xleftarrow{\mu_{\mathbb{F}}} & \mathbb{F}^2(\mathbb{G}X) \\
 \searrow \mathbb{G}\eta_{\mathbb{F}} & & \downarrow \lambda & & \downarrow \mathbb{F}\lambda \\
 & & \mathbb{G}\mathbb{F}X & \xleftarrow{\mathbb{G}\mu_{\mathbb{F}}} & \mathbb{G}(\mathbb{F}^2X) \xleftarrow{\lambda} & \mathbb{F}\mathbb{G}\mathbb{F}X
 \end{array}$$

(Details in the paper.)



# Pairing!

## Not yet done!

There is a price to pay for the “hosting” process.

Definition of  $\llbracket m_1 \rrbracket ; \llbracket m_2 \rrbracket$  is **strongly** monadic.

Question:

---

*Do **strong monads lift** to strong monads?*

---

Recall the types of the two **strengths**:

$$\tau_l : (B \times \mathbb{F} A) \rightarrow \mathbb{F} (B \times A)$$

$$\tau_r : (\mathbb{F} A \times B) \rightarrow \mathbb{F} (A \times B)$$

The basic properties, e.g.  $\mathbb{F} \text{ lift} \cdot \tau_r = \text{lift}$  and

$\mathbb{F} a^\circ \cdot \tau_r = \tau_r \cdot (\tau_r \times \text{id}) \cdot a^\circ$  are preserved by their liftings (e.g.  $\overline{\tau_r}$ )

by construction.

## Naturality vs lifting

So, what may fail is their **naturality**, e.g.

$$\bar{\tau}_l \cdot (N \otimes \bar{\mathbb{F}} M) = \bar{\mathbb{F}} (N \otimes M) \cdot \bar{\tau}_l$$

where  $M$  and  $N$  are arbitrary CS matrices and  $\cdot \otimes \cdot$  is Kronecker product.

---

**Naturality** *is essential to pointfree proofs!*

---

Example: for  $\mathbb{F} = \mathbb{M} = (1+)$  we have e.g.  $\bar{\tau}_l = (\bar{!} \oplus id) \cdot \bar{dr}$ , that is

$$1 + A \times B \xleftarrow{! \oplus id} (1 \times B) + (A \times B) \xleftarrow{dr} (1 + A) \times B$$

dropping the  $\bar{f}$  bars over functions for easier reading.

## Naturality which lifts

Is  $\bar{!} \oplus id$  natural? We check:

$$(id \oplus N) \cdot (! \oplus id) = (! \oplus id) \cdot (M \oplus N)$$

$$\Leftrightarrow \{ \text{bifunctor } \cdot \oplus \cdot \}$$

$$! \oplus N = (! \cdot M) \oplus N$$

$$\Leftrightarrow \{ ! \cdot M = ! \text{ because } M \text{ is a CS matrix} \}$$

*true*

---

**Note:** matrix  $M$  is CS iff  $! \cdot M = !$  holds. (Thus composition is closed over CS-matrices.)

## Naturality which does not lift

Is the diagonal function  $\delta = id \triangle id$  — that is

$$\delta x = (x, x)$$

still natural once lifted to matrices?

**No!** Diagram

$$\begin{array}{ccc}
 A \times A & \xleftarrow{\delta} & A \\
 M \otimes M \downarrow & & \downarrow M \\
 B \times B & \xleftarrow{\delta} & B
 \end{array}$$

does not commute for every CS matrix  $M : A \rightarrow B$  — counter-example in the next slide.

# Naturality which does not lift

Given probabilistic  $f$

$f$	a	b
F	0.3	1
T	0.7	0

evaluate  $\delta \cdot f$

		$f$	a	b	
		F	0.3	1	
		T	0.7	0	
$\delta$	F				
(F,F)	1	0	0.3	1	(F,F)
(F,T)	0	0	0	0	(F,T)
(T,F)	0	0	0	0	(T,F)
(T,T)	0	1	0.7	0	(T,T)

$\delta * f$

Then evaluate  $(f \otimes f) \cdot \delta$

		$\delta$	a	b		
	(a,a)	1	0			
	(a,b)	0	0			
	(b,a)	0	0			
	(b,b)	0	1			
$(f \times f)$	(a,a)	(a,b)	(b,a)	(b,b)		
(F,F)	0.09	0.3	0.3	1	0.09	1
(F,T)	0.21	0	0.7	0	0.21	0
(T,F)	0.21	0.7	0	0	0.21	0
(T,T)	0.49	0	0	0	0.49	0

$(f \times f) * \delta$

where  $\delta : \{a, b\} \rightarrow \{a, b\} \times \{a, b\}$

where  $\delta : \mathbb{B} \rightarrow \mathbb{B} \times \mathbb{B}$

## Probabilistic pairing

This happens because the Kleisli-lifting of **pairing**

$$(f \triangle g) x = (f x, g x)$$

is a **weak**-product for column stochastic matrices:

$$X = M \triangle N \Rightarrow \begin{cases} fst \cdot X = M \\ snd \cdot X = N \end{cases} \quad (5)$$

ie. ( $\Leftarrow$ ) is not guaranteed

So  $(fst \cdot X) \triangle (snd \cdot X)$  differs from  $X$  in general.

---

In LA,  $M \triangle N$  is known as the **Khatri-Rao** matrix product.

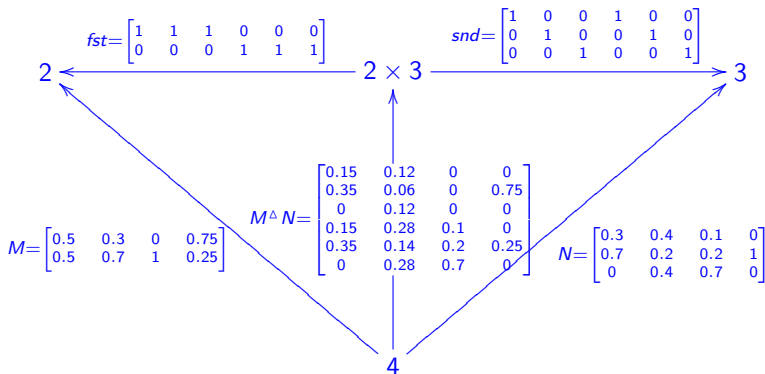
In RA,  $R \triangle S$  is known as the **fork** operator.

# Probabilistic pairing

In summary: weak product (5) still grants the **cancellation** rule,

$$fst \cdot (M \triangle N) = M \wedge snd \cdot (M \triangle N) = N$$

cf. e.g.





# Probabilistic pairing

... but **reconstruction**

$$X = (fst \cdot X) \triangle (snd \cdot X)$$

doesn't hold in general, cf. e.g.

$$\begin{array}{l}
 X : 2 \rightarrow 2 \times 3 \\
 X = \begin{bmatrix} 0 & 0.4 \\ 0.2 & 0 \\ 0.2 & 0.1 \\ 0.6 & 0.4 \\ 0 & 0 \\ 0 & 0.1 \end{bmatrix}
 \end{array}
 \quad
 (fst \cdot X) \triangle (snd \cdot X) = \begin{bmatrix} 0.24 & 0.4 \\ 0.08 & 0 \\ 0.08 & 0.1 \\ 0.36 & 0.4 \\ 0.12 & 0 \\ 0.12 & 0.1 \end{bmatrix}$$

( $X$  is not recoverable from its projections — Khatri-Rao not surjective).

This is not surprising (cf. RA) but creates difficulties and needs attention.

# Closing

# Research proposal

Need to quantify software (un)reliability in presence of faults.

Need for **weighted** nondeterminism, e.g. **probabilism**.

Relation algebra  $\rightarrow$  Matrix algebra

Usual strategy:

---

*“Keep category (sets), change definition”*

---

Proposed strategy:

---

*“Keep definition, change category”*

---

## Change category

Possible wherever semantic models are structured around a pair  $(\mathbb{F}, \mathbb{G})$  of monads:

Monad	$\mathbb{F}$	$\mathbb{G}$
Effect	Transition	Branching
Role	Guest	Host
Strategy	Lifted	“Kleislified”

Works nicely for those  $\mathbb{G}$  for which well-established Kleisli categories are known, for instance (aside):

$\mathbb{G}$	Kleisli
$\mathbb{P}$	Relation algebra
$\mathit{Vec}$	Matrix algebra
$\mathbb{D}$	Stochastic matrices
$\mathit{Giry}$	Stochastic relations

cf. (Panangaden, 2009) etc.

## Future work

- **LAoP** in its infancy — really a lot to do!
- Relation to **quantum** physics — cf. remarks by Coecke and Paquette, in their *Categories for the Practising Physicist* (Coecke, 2011):

*Rel [the category of relations] possesses more 'quantum features' than the category Set of sets and functions [...]  
The categories FdHilb and Rel moreover admit a categorical matrix calculus.*

- **Final** (behavioural) **semantics** of pMMM calls for infinite support distributions.
- **Measure** theory — Kerstan and König (2012) provide an excellent starting point.
- Case studies!

# Verification of IBM 4765

Marić and Sprenger (2014) rely on MMM of type

$$(Q \times A) \rightarrow \mathbb{P}((2 + V) \times Q)$$

for verifying a persistent memory manager (in IBMs 4765 secure coprocessor) in face of **restarts** and **hardware failures**, where

- $V$  - (normal) return values
- $2$  - exceptions (either “regular” or “restarts”)

Interested in scaling up  $\mathbb{P}$  to  $\mathbb{D}$  and do the proofs using (pointfree!) matrix algebra where they use explicit monad **transformers** etc, etc (Isabelle).

## The monadic “curse”

*“Monads [...] come with a curse. The monadic curse is that once someone learns what monads are and how to use them, they lose the ability to explain it to other people”*


(Douglas Crockford: Google Tech Talk on how to express monads in JavaScript, 2013)



Douglas Crockford (2013)

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