

An approach to computational category theory

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Includes joint work with David Spivak (operadic normal forms)

Outline

I'll discuss

- Overall design concept for a computer algebra and modelling system based on computational category theory
- Some necessary component algorithms:
 - ▶ Determining equality for morphism terms,
 - ▶ Related aspects of normal forms and rewriting (such as finding tree decompositions), and
 - ▶ Categorically-formulated belief propagation, a common generalization of algorithms used for many types of models.
- Status of *Cateno*: pre-alpha software, looking for “customers” and collaborators

Motivation

Goal: build a practical computer algebra system for computational (monoidal) category theory. It should be able to:

- 1 **manipulate** abstract categorical quantities such as **morphism terms** in a REPL.
- 2 **compile**/lower code expressed categorically **to an efficient implementation** in a particular category (e.g. numerical linear algebra, (probabilistic) databases, quantum simulation, belief networks).
- 3 **scale** to be useful for practical computational problems in modeling uncertainty in data analysis, statistics, physics, computer science.

Motivation (modelers)

Goal: Abstractions and engine to quickly build performant, modular models of uncertainty. It should be able to:

- Ingest qualitative domain knowledge expressed with flow charts/diagrams from non-programmers
- Attach multiple competing quantitative explanations (e.g. probabilistic, differential equation, discrete dynamical) and test them
- Port algorithms, using best solution in each component (e.g. tree decomposition)
- Scale to useful size
- Pay only a small abstraction cost in final assembly program.

Want something like REPLs for linear or commutative algebra

In a computer algebra systems such as Macaulay2, Singular, Maple, Mathematica, we:

- tell the computer the context, e.g. polynomial ring $R = \mathbb{Q}[x, y]$, and
- type in an expression such as $x^2 + 3xy + (x + y)^2$.
- The system performs some simplification according to the axioms of a free polynomial ring (or more generally, a Gröbner basis computation in a quotient ring $R = \mathbb{Q}[x, y]/I$), and
- displays something like $2x^2 + 5xy + y^2$, element of R .
- Even something this trivial requires some thought for computational category theory!

Want something like MATLAB, NumPy, R

But that:

- allows a higher level of abstraction in describing algorithms,
- can handle more types of “data” than matrices of floats or probability distributions, and
- treats morphism expressions as first class to enable rewriting, syntax tricks

Absurd Claim: Computational category theory is the numerical linear algebra of the 21st century.

- Now: reducing an applied math problem to numerical linear algebra means you can solve it using BLAS, LAPACK primitives, matrix decompositions, etc.
- Future: reduce your uncertainty/information-processing applied math problem to (computational) category theory, solve it using generic engines and libraries with matching abstractions.

Design Principles

- Today: some **design principles** toward such a system.
- Can be implemented in any programming language with suitable features.
 - ▶ Needed are typeclasses/traits/interfaces and, for performance, a means for zero or low cost abstraction
 - ▶ so JIT and AOT languages with modern type systems are preferred
 - ▶ building in Julia, exploring Scala, Rust, maybe Python (what would you use?)
- Typeclasses/Traits represent *doctrines*: monoidal category, compact closed category, well-supported compact closed category, . . . and describe the *common interface* available to manipulate terms in any particular category

Design Principles

Everything the computer does is represented as one of five modular components:

- a **doctrine** (e.g. “compact closed category”)
- an instance or implementation of a doctrine (e.g. matrices or relations as CCC):
 - ▶ a **morphism term** or word (e.g. $f \otimes (g \circ \delta_A \circ h)$) in a free language, or
 - ▶ a concrete **value** in an implementation (e.g. $\begin{pmatrix} 1 & 3 \\ 4 & 5 \end{pmatrix}$)
- a **representation** (an X -functor) between implementations (usu. free \rightarrow concrete, e.g. a binding $f = \begin{pmatrix} 1 & 3 \\ 4 & 5 \end{pmatrix}$ for each symbol)
- **algorithms** are expressed in terms of the defining methods of the doctrine (e.g. $\otimes, \delta, \mu, \text{bind}, \text{return}$) or operadically

From the modeling perspective:

- **Morphism Term**: a human-readable **qualitative model**, captured by a labeled generalized graph; fixes the relationships, *suggests* qualitative rules and syntax of the model
- **Doctrine**: formal **categorical syntax** constraining the quantitative models of uncertainty that can be attached, rewrite rules, available constructions
- **Value**: a machine-processed **quantitative model** in which the graph is interpreted and the data summarized, e.g. probabilistically as a Bayesian network, in Hilbert space for a quantum circuit, or with rate constants in a chemical reaction network
- **Representation**: the **interpretation** assigning quantitative meaning to the qualitative description (generalizing the mathematical idea of a representation of a quiver or algebra)
- **Algorithms**, categorically expressed, for processing and analyzing data. Make quantitative predictions, choose the model which best explains a given system (often a variant of belief propagation).

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- displays something like $2x^2 + 5xy + y^2$, element of R .

REPL/Computer algebra system for computational category theory

In a computational category theory REPL, we

- tell it the context, e.g. the tensor signature $f : A \rightarrow A$, $g : A \rightarrow B$, with doctrine “compact closed category,” and
- type in an expression such as $\text{ev}_A \circ (\text{id}_{A^*} \otimes f) \circ \text{coev}_A$.
- The system performs some simplification according to the axioms of a free compact closed category, and
- displays $\text{tr}(f)$.
- Even easier: write $\text{ev}_A \circ (\text{id}_{A^*} \otimes f) \circ \text{coev}_A == \text{tr}(f)$ and have the system return “True”
- This easier problem is complete for the complexity class Graph Isomorphism (GI). There is a partial solution that doesn't require solving graph isomorphism: pick a good interpretation functor.

Monoidal languages

- Given object variables $O = \{A_1, \dots, A_n\}$, get monoid O^\otimes of words such as $(A_5 \otimes A_3) \otimes A_1$.
- A **tensor signature** \mathcal{T} comprises finite sets $\text{Ob}(\mathcal{T})$ of object variables, and $\text{Mor}(\mathcal{T})$ of morphism variables, and functions $\text{dom}, \text{cod} : \text{Mor}(\mathcal{T}) \rightarrow \text{Ob}(\mathcal{T})^\otimes$.
- \mathcal{T} defines a monoidal category $\mathcal{M}_X(\mathcal{T})$,
 - ▶ augmenting $\text{Ob}(\mathcal{T})$ with a monoidal unit $I_{\mathcal{T}}$ and
 - ▶ $\text{Mor}(\mathcal{T})$ with a finite set of parameterized structure morphisms $\text{PSM}(\mathcal{T}, X)$ depending on the doctrine X .
 - ▶ Here $X = \text{“compact closed category”}$, PSM are e.g. ev_A for any A , etc.
- The language $\mathcal{T}_{\text{CCC}}^{\otimes, \circ}$ is all valid morphism words that can be formed from $\text{Mor}(\mathcal{T}) \cup \text{PSM}(\mathcal{T}, \text{CCC})$, so generates the free compact closed category over \mathcal{T} .
- Q: **When do two words define the same morphism? NF?**

Constructively

Constructively, $\mathcal{T}_{CCC}^{\otimes, \circ}$ is as follows.

- Each $f \in \text{Mor}(\mathcal{T}) \cup \text{PSM}(\mathcal{T}, CCC)$ is a word.
- Given words u, u' , $u \otimes u'$ is a word with domain $\text{dom}(u) \otimes \text{dom}(u')$ and codomain $\text{cod}(u) \otimes \text{cod}(u')$.
- Given words w, w' with $\text{dom}(w') = \text{cod}(w)$, $w \circ w'$ is a word.

Mod the relations for a compact closed category, and imposing strictness, this gives a presentation of the free strict compact closed category over the generating set of object variables and morphisms.

Complexity

- The easy part, Solving word problems such as $\text{ev}_A \circ (\text{id}_{A^*} \otimes f) \circ \text{coev}_A \stackrel{?}{=} \text{tr}(A)$ is **GI-complete**.
- For some reasonable choices of normal form, finding the normal form (fixing tensor signature and doctrine = compact closed category):
 - ▶ given $\text{ev}_A \circ (\text{id}_{A^*} \otimes f) \circ \text{coev}_A$, output $\text{tr}(f)$ is **NP-complete** (allows optimal contractions).
- This is not such bad news. The **normal form for polynomials** in a quotient ring, using Gröbner bases and Buchberger's algorithm, is worst case **doubly exponential**. Yet they are still extremely useful and practical for many computations.
- If we are willing to accept an imperfect but pretty good normal form, or can control the term complexity in various ways, we can get a good, fast normal form giving near-optimal contractions.

Computational category theory problems

- *term*: equivalence class of monoidal words over a finite tensor scheme, usually with certain additional properties X .
- *representation*: an X -monoidal functor “assigning values”

Questions of a term represented in a particular quantitative category:

- 1 compute a (possibly partial) contraction,
- 2 solve the word problem (are two terms equivalent, i.e. do they have the same representation) or compute a normal form for a term,
- 3 solve the implementability problem (construct a word equivalent to a target using a library of allowed morphisms), and
- 4 choose morphisms in a term to best approximate a more general term (possibly allowing the approximating term itself to vary).

Computational category theory is hard

- *term*: equivalence class of monoidal words over a finite tensor scheme, usually with certain additional properties X .
- *representation*: an X -monoidal functor “assigning values”

Questions of a term represented in a particular category:

- 1 compute a (possibly partial) contraction, (**#P-hard**)
- 2 solve the word problem (are two terms equivalent, i.e. do they have the same representation) or compute a normal form for a term, (**undecidable**)
- 3 solve the implementability problem (construct a word equivalent to a target using a library of allowed morphisms) (**undecidable**)
- 4 choose morphisms in a term to best approximate a more general term (possibly allowing the approximating term itself to vary). (**NP-hard**)

Recap

- Every doctrine has a free language that can be used to express morphism terms.
- This is how the machine represents domain-expert diagrams (e.g. gene interactions, high-school science test).
- A morphism term can be rewritten and simplified efficiently independently of interpretation/value.
- The **normal form problem** for morphism term subsumes query planning, finding an efficient way to contract a network, etc.
- Part of this problem can be solved “internally” by applying a suitable functor to a value category that removes distinctions between equal morphism terms.

Obvious approach: rewriting

- View a term as a grid of partial terms
- Look for tiles of any shape in the grid and rewrite them to something simpler
- Problem: dead ends, have to choose how to associate
- No known confluent terminating rewriting system
- Will need to solve parts of this eventually

S-expression

The value of the expression

`(foo bar baz)`

is the result of applying function `f` to arguments `bar` and `baz` (which may themselves be expressions).

Morphism expressions

- A morphism term such as $\text{ev}_A \circ (\text{id}_{A^*} \otimes f) \circ \text{coev}_A$ in a monoidal category can be represented as an expression tree (AST):

$$(\circ (\text{ev } A) (\circ (\otimes (\text{id } A) f) (\text{coev } A)))$$

here **ev**, **coev**, **id**, are unary functions, \otimes , and \circ are 2-ary and $(\otimes g f)$ means “apply function \otimes to arguments g and f .”

- Note: more than one expression tree can represent the same term (e.g. associate from left or right), and
- more than one morphism term can represent the same morphism in the free compact closed category (e.g. $(f_2 \otimes g_2) \circ (f_1 \otimes g_1)$ vs. $(f_2 \circ f_1) \otimes (g_2 \circ g_1)$).

Philosophy

- Fix a doctrine X .
- A morphism term (parenthesis and all) in an X -category is a **program** for obtaining a value.
- Don't rewrite!
- Instead, find good bindings for the symbols in another X -category
- such that when the program is executed, a normal form is obtained as that value.
- (or something from which a normal form can be extracted)

Operad approach

- Now we have not two (\otimes, \circ) but one notion of composition: function composition.
- There are different ops or functions, taking different numbers of arguments, and we impose some associativity constraints.
- This matches the setting of operads.
- We represent each function in the expression (such as \otimes), and each primitive (e.g. f , ev_A) as an op in an operad. For the CCC doctrine, the relevant operad is 1-cobordisms.
- 1-D composition in that operad is associative. So different terms, and different trees with the same morphism give the same answer. (up to GI issue).

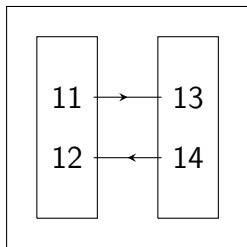
k-ary ops

The ops include

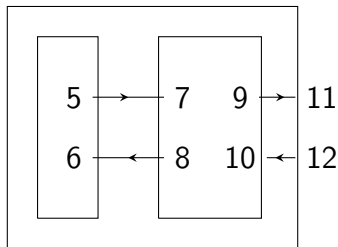
- 2-ary ops: \otimes, \circ
- 1-ary ops: (partial) traces, (partial) transposes, and
- **0-ary ops**: $\text{ev}_A, \text{coev}_A, \text{id}_A, \sigma_{A,B}$.

Note that 0-ary ops and 2-ary ops again form a monoidal category equivalent to the original. Each op has a representation in 1-Cob.

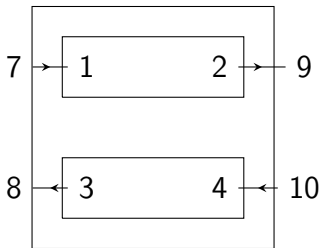
This depends on the type information, e.g. if $f : A \rightarrow B$ and $g : A \otimes A \rightarrow I$, the \otimes in $f \otimes f$ and $g \otimes g$ technically represent two different ops (although they are implemented by the same function).



ϕ_3



ϕ_2



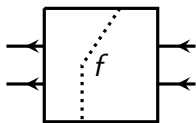
ϕ_1

The ops ϕ_3 and ϕ_2 correspond to compositions, while ϕ_1 corresponds to a tensor product.

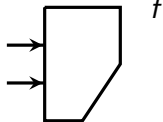
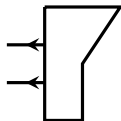
Representing in / Code lowering to 1-Cob

- Even when we have many objects and their duals, we can typecheck this at the top level (when the morphism term is formed).
- Represent object variables as wires, morphism variables as labeled identity ops, and structure morphisms $ev, coev$, etc. as special 0-ary ops.
- Lowered morphism term is just a 1-Cob
- Representing in 1-Cob, then evaluating, gives a (typed) morphism of 1-Cob.
- This typed morphism gets rid of all relations for CCC, except those requiring the solution of a graph isomorphism problem.

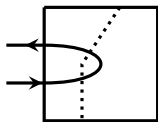
0-ary ops from morphisms: morphism variable f



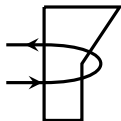
$\Downarrow \rho$



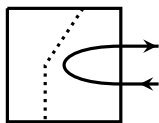
0-ary ops from morphisms: $\text{coev} : I \rightarrow A \otimes A^*$



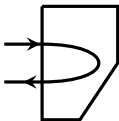
$\Downarrow \rho$



0-ary ops from morphisms: $ev : A^* \otimes A$



$\Downarrow \rho$



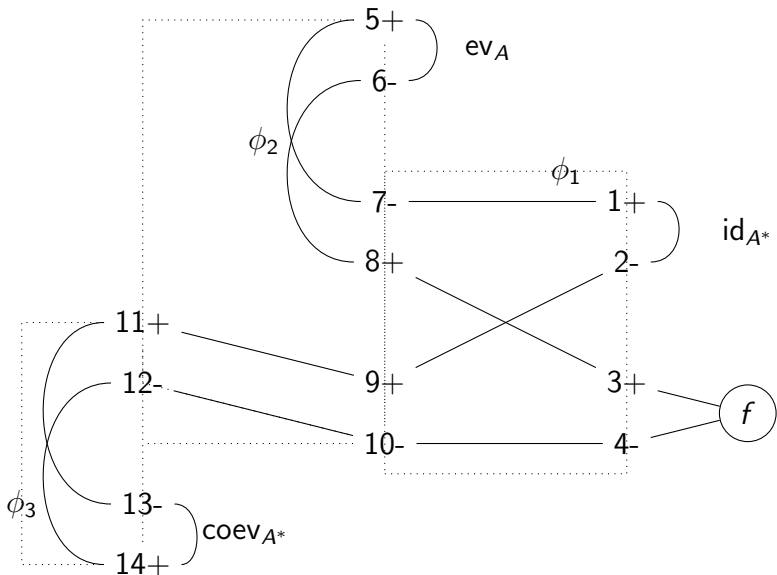
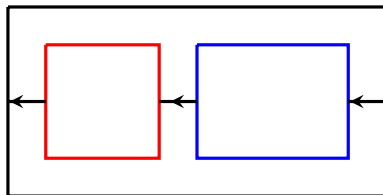
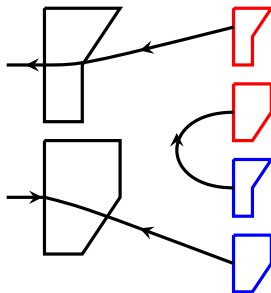


Figure : Operad tree and abstract syntax tree for $(\circ, ev_A, (\otimes, id_{A^*}, f), coev_{A^*})$ with cobordisms.

2-ary ops from composition: \circ



$\Downarrow \rho$



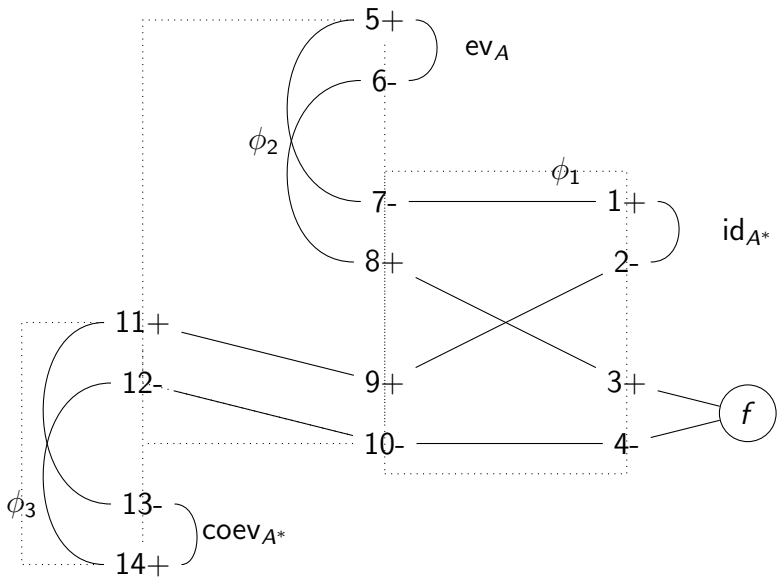
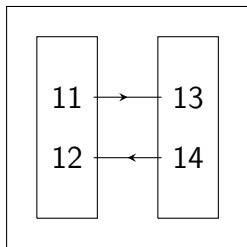
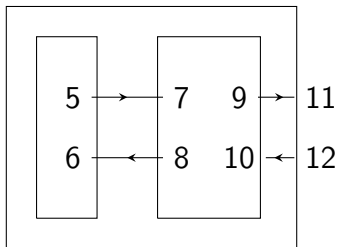


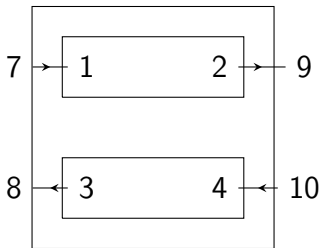
Figure : Operad tree and abstract syntax tree for $(\circ(\circ, \text{ev}_A, (\otimes, \text{id}_{A^*}, f)), \text{coev}_{A^*})$ with cobordisms.



ϕ_3



ϕ_2



ϕ_1

The ops ϕ_3 and ϕ_2 correspond to compositions, while ϕ_1 corresponds to a tensor product.

Graph isomorphism issue

A hierarchy

- A morphism expression is what is stored in the machine
 - ▶ A strict morphism word is an equivalence class of expressions
- Evaluating a morphism expression in terms of 1-Cob as we've described yields a labelled 1-Cob.
 - ▶ If there is only one morphism variable, labelled 1-Cobs are equal iff the morphism terms are equal
 - ▶ If there are > 1 morphism variables, a labelled 1-Cob may have **nontrivial automorphisms** under permutation of the morphism variables
 - ▶ A labelled 1-Cob is an equivalence class of words (and so of morphism expressions)
- A morphism of the category is a (GI)-equivalence class of labelled 1-Cobs, **up to these automorphisms.**

Obtaining a normal form from a labeled 1-Cob

- Any stable procedure for obtaining a morphism expression from a labeled 1-Cob yields a normal form.
- Probably the most interesting are efficient ones wrt some measure of the cost of each 2-ary op.
- This is where the NP-hardness appears, in finding an optimal tree decomposition
- There are many deeply developed algorithms for this problem, and heuristics work quite well

Algorithms authored axiomatically

- When modeling, algorithms should be expressed in terms of categorical primitives.
- The right primitive for many algorithms is the well-supported compact closed category [RSW05].
- Belief propagation (and its many variants) can be expressed this way [Morton 2014].

(Limited) practical algorithms, even though the problems are hard in general

- Many practical questions are instances of one of these problems
 - ▶ quantum programming and logic
 - ▶ probabilistic graphical models,
 - ▶ tensor network state approach to quantum condensed matter,
 - ▶ computational complexity theory: circuits, CSP, #CSP
 - ▶ even databases
- So many tractable special cases, approximate algorithms, and heuristics exist
- Let's **turn these into categorical algorithms**, and formalize analogies among procedures.

Example: belief propagation

- A message-passing algorithm (Pearl 1982), for contraction, marginalization, and optimization problems
- Many extensions, analogs (survey propagation, turbo coding)
- These should be the same abstract **categorical** algorithm, varying the category (e.g. prob. graphical models vs. sets and relations).

To make this precise, first recall the set-up

Monoidal languages

- Given object variables $O = \{A_1, \dots, A_n\}$, get monoid O^\otimes of words such as $(A_5 \otimes A_3) \otimes A_1$.
- A **tensor signature** \mathcal{T} comprises finite sets $\text{Ob}(\mathcal{T})$ of object variables, and $\text{Mor}(\mathcal{T})$ of morphism variables, and functions $\text{dom}, \text{cod} : \text{Mor}(\mathcal{T}) \rightarrow \text{Ob}(\mathcal{T})^\otimes$.
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- Q: **When do two words define the same morphism? NF?**

I -valued points: the messages

- Important word problem for Belief Propagation: equality of morphisms of type $\text{Mor}(I, A)$ for objects A (I -valued points).
- Why?
 - ▶ Want to generalize algorithms (e.g. belief propagation in the category of vector spaces and linear transformations)
 - ▶ **Can't** assume objects A are **sets** with points (such as probability distributions in the classical belief propagation algorithm).
- But, messages are still **morphisms** of type $\text{Mor}(I, A)$ for each object A ; equate these for belief propagation equations
 - ▶ Deciding if **two vectors are equal** up to numerical tolerance becomes **deciding a word problem in $\text{Mor}(I, A)$** .
 - ▶ These messages must also be stored somehow.

Word problems in monoidal languages

- Coherent graphical languages for some types of monoidal categories means those word problems can be reduced to e.g. graph isomorphism, and produces normal forms by $\text{word} \mapsto \text{graph} \mapsto \text{word}$.
- Hence the word problem for the free closed category and free compact closed category over a finite tensor scheme are in LOGSPACE and P [Luk82] respectively.
- Adding adjectives (X -monoidal categories) and relations, or fixing values by applying a functor F , so that the category is no longer free may make it easier or harder.

Proposition

The word problem and implementability problem in a monoidal category over a finite tensor scheme are undecidable.

I -valued points: the messages

- For an efficient algorithm, need **representation** and **word problem** for I -valued points to be **efficient**.
- Classical belief propagation: have a monoid homomorphism, $\text{size} : \text{Ob}(\mathcal{T})^{\otimes} \rightarrow \mathbb{N}$, from the free monoid generated by the objects of our tensor scheme to the natural numbers.
- Monoidal product \mapsto multiplication of vector space dimensions
- Then words in $\text{Mor}(I, A)$ can be stored and compared in $O(\text{size}(A))$.

Now look at type of category BP will work in. Need something like variables: objects in a well-supported compact closed category accomplish this.

Sum-product and belief propagation for contraction

- The **sum product algorithm** [KFL01]: if a term is a tree, can perform contraction according to the tree.
- If not, use a **tree decomposition** [Hal76] to force it to be a tree, then run sum-product.
 - ▶ Tree decompositions can be computed at the symbolic level with a cost function (e.g. dimension of each vector space); best is NP-hard but many good approximation strategies exist.
 - ▶ The result is the *junction tree algorithm* [LS88], also extended to the quantum case [MS08].
- Can improve on the abstract sum-product algorithm by using an **optimized message-passing version**, which among other benefits permits parallelization.

this is belief propagation

Belief propagation in factor graphs

- The algorithm operates on a *factor graph*, a bipartite graph with
 - ▶ one part **discrete random variables** $v \in V$ and
 - ▶ one part **factors** $u \in U$.
- Each **factor** (potential) assigns a real number to each combination of states of the variables it is connected to.
- Multiplying factors and normalizing if needed gives a joint probability distribution.
- Belief propagation is a **message passing algorithm**.
 - ▶ Each **message** is a probability distribution over the states one variable v can take: a vector in the associated vector space V_v .
- Each **factor** f_u at node u is a tensor in $\otimes_{v \in \text{nbhd}(u)} V_v$, defines $\text{valence}(u)$ reshaped linear maps

$$f_{u,v} : \otimes_{i \in \text{nbhd}(u) \setminus v} V_i \rightarrow V_v,$$

one for each $v \in \text{nbhd}(u)$.

Messages at variables.

- Compute the pointwise (Hadamard) product of the incoming messages, and output it as the outgoing message along e .
- In a probabilistic category, Hadamard product rescales so the out message is a probability distribution.
- If there are no incoming messages, output the uniform message.

Messages at factors.

- Compute the tensor product of the incoming messages,
- apply reshaped $f_{u,v} : \otimes_{i \in \text{nbhd}(u) \setminus v} V_i \rightarrow V_v$, and output the result as the outgoing message along the edge to v .

Resulting algorithm.

- *BP equations* describe fixed points of the update rules.
- Initial messages can be uniform distributions.
- Tree factor graph: done in two “passes,” leaves to root then root to leaves, updating messages only as they change.
- Belief propagation is exact on trees

Messages at spiders.

- Apply the reshaped spider to incoming messages, and output the result as the outgoing message.
- If there are no incoming messages, treat the spider as a Frobenius unit.

Messages at “factor” morphisms.

- Compute the monoidal product of the incoming messages,
- apply the reshaped f ,
- output the result as the outgoing message.






Resulting algorithm.

- System of BP equations are equalities of I -valued points describing the fixed points of the update rules.
- Initial messages can be chosen to be units at the spiders.
- Nice behavior on trees preserved

A spider is just a special kind of morphism. To get the **general bipartite** version, replace the message procedure at spiders with another copy of the factor message procedure.

To solve a problem, just reduce it to computational category theory

- **Goal:** general tools that work for any category with suitable properties
 - ▶ specialize automatically by giving a **monoidal category interface**
- Rapidly expanding universe of applied problems given categorical representations
 - ▶ a problem-solving abstraction with the potential to be as useful as convex programming or numerical linear algebra.

-  Bob Coecke and Ross Duncan, *Interacting quantum observables: categorical algebra and diagrammatics*, *New Journal of Physics* **13** (2011), no. 4, 043016.
-  Rudolf Halin, *S-functions for graphs*, *Journal of Geometry* **8** (1976), no. 1-2, 171–186.
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