Belief Propagation in Monoidal Categories

Jason Morton

Penn State

June 5, 2014 QPL 2014

Computational category theory with diagrams

- *diagram*: equivalence class of monoidal words over a finite tensor scheme, usually with certain additional properties *X*.
- interpretation: an X-monoidal functor "assigning values"

Questions of a diagram interpreted in a particular category:

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

Computational category theory is hard

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Questions of a diagram interpreted in a particular category:

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

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(Limited) practical algorithms anyway

- 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

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- So many tractable special cases, approximate algorithms, and heuristics exist
- Let's **turn these into categorical algorithms** (see also [MT13]). 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).

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To make this precise, first describe set-up

Tensor schemes and monoidal languages

Definition ([JS91, Sel09])

A (finite) tensor scheme ${\mathscr T}$ (or monoidal signature) is

- a finite set $\mathrm{Ob}_V(\mathcal{T})$ of object variables (including a monoidal identity object I),
- ullet a finite set $Mor(\mathcal{T})$ of morphism variables, and
- functions dom, cod : $Mor(\mathscr{T}) \to Ob(\mathscr{T}) = \otimes$ -words in obj vars.

May also add relations.

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May also add relations. The monoidal language $\mathscr{T}^{\otimes,\circ}$ comprises all valid morphism words built from $\mathsf{Mor}(\mathscr{T})$, and identity morphisms. It generates the free monoidal category over \mathscr{T} . Constructively:

- For all $A \in \mathsf{Ob}(\mathscr{T})$, id_A is a word.
- **2** Each $f \in Mor(\mathscr{T})$ is a word.
- **③** Given words u, u', u ⊗ u' is a word with domain dom(u) ⊗ dom(u') and codomain cod(u) ⊗ cod(u').
- Given words w, w' with dom(w') = cod(w), $w \circ w'$ is a word.

Words in a monoidal language

- Interpret a word to define a morphism is a particular category.
 - Factors through free monoidal category, which imposes equivalences such as $\mathrm{id}_A \circ f = f$ and $(f \otimes g) \circ (f' \otimes g') = (f \circ f') \otimes (g \circ g')$,
 - Words equivalent if represent same morphism in the free (X-) monoidal category.

Definition

An equivalence class of words in the free (X-) monoidal category over a tensor scheme is called a *diagram*.

- Further notions of equivalence arise with additional relations.
- So far, no normal form for words



I-valued points: the messages

- Important word problem for Belief Propagation: equality of morphisms of type 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).

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- 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 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 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 [DK13], and produces normal forms by word→graph→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.

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Proposition

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

Word problems, term rewriting, normal forms

- Preferable to have a confluent terminating rewriting system that attached a direction to the equalities of the X-category.
- Term rewriting and computing normal forms in monoidal categories is a field in its infancy [Kis12, Mim14]
- Or, exploit completeness
 - Finite dimensional vector spaces over a field of characteristic zero are complete for traced symmetric monoidal categories [HHP08] and finite dimensional Hilbert spaces are complete for dagger compact closed categories [Sel11].

I-valued points: the messages

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

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Now look at type of category BP will work in. Need something like variables.

Spiders: generalized variables

Definition

A *spidered category* is a strict symmetric monoidal category equipped with a special commutative (†) Frobenius structure [CPV08] $(A, m, u, \delta, \epsilon, \sigma^F)$ on each object A.

- Note: morphisms of a spidered category not generally monoid or comonoid homomorphisms.
- Now add duals for objects to obtain a compact closed category with additional structure.

Dungeon category

Call a compact closed category spidered in a compatible way a

Definition

A dungeon category is a compact closed category (C, σ^C, i, e) s.t.

- (i) Each object has a special commutative Frobenius structure $(A, m, u, \delta, \epsilon, \sigma^F)$ with $\sigma_{A^{(*)}, A^{(*)}}^F = \sigma_{A^{(*)}, A^{(*)}}^C$, and
- (ii) Any two morphisms which are
 - Constructed from the identity id_A , the symmetric braiding $\sigma_{A,A}$, the Frobenius morphisms, and the dualizing cup and cap morphisms i_A , e_A for A, and
 - Have the same domain (tensor product of zero or more or copies of A and A^*) and the same codomain (another such tensor product)
 - Are equal.

Dungeon category

So a directed spider morphism depends only on

- the number of inward and outward directed arrows,
- which way they point,
- and their order

Good setting for generalized belief propagation because we can

- bend wires to choose inputs and outputs of any morphism and
- have spiders that play the role of variables in the probabilistic setting for belief propagation.

Sum-product and belief propagation for contraction

- The sum product algorithm [KFL01]: if a diagram 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.
 - ▶ This is the *junction tree algorithm* [LS88], also extended to the quantum case [MS08].

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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 $\bigotimes_{v \in \mathsf{nbhd}(u)} V_v$, defines valence(u) reshaped linear maps

$$f_{u,v}: \bigotimes_{i \in \mathsf{nbhd}(u) \setminus v} V_i \to V_v,$$

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



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 $f_{u,v}: \otimes_{i\in \mathsf{nbhd}(u)\setminus v} V_i \to V_v$, 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 to 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 interpretations
 - a problem-solving abstraction with the potential to be as useful as convex programming or numerical linear algebra.

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