

Simulations between contextual resources



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Contextuality as a resource in quantum computation
Oxford, 4th July 2019

Overview of the talk

- ▶ Simulations
- ▶ Main results:
 - ▶ Categorical characterizations of contextuality
 - ▶ Equivalence with free operations
 - ▶ No-cloning
- ▶ Further topics:
 - ▶ More expressive simulations?
 - ▶ No-catalysis?
 - ▶ Further questions

Structure of resources

Two perspectives:

1. **Resource theories** (coming from Physics):

Algebraic theory of '**free operations**' that do not introduce more of the resource in question.

Resource B can be obtained from resource A if it can be built from A using free operations.

'Contextual fraction as a measure of contextuality', Abramsky, B, Mansfield, PRL, 2017.

'Noncontextual wirings', Amaral, Cabello, Terra Cunha, Aolita, PRL, 2018.

2. **Simulations or reducibility** (coming from Computer Science):

Notion of **simulation** between behaviours of systems.

One resource can be reduced to another if it can be simulated by it.

Cf. (in)computability, degrees of unsolvability, complexity classes.

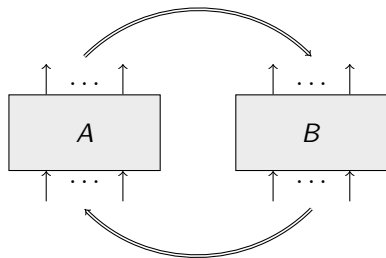
- ▶ Intuitively, some correlations can be used to simulate others. Here's some examples:
 - ▶ Any two-outcome bipartite box can be simulated with PR boxes.
'Popescu-Rohrlich correlations as a unit of nonlocality'
Barrett, Pironio, PRL 2005.
 - ▶ An explicit two-outcome three-partite box that cannot be simulated with PR boxes (same source).
 - ▶ No finite set of bipartite boxes can simulate all of them
'No nonlocal box is universal'
Dupuis et al, Journal of Mathematical Physics 2007.

Morphisms as simulations

- ▶ Think of a measurement scenario as a concrete experimental setup, where for each measurement there is a grad student responsible for it.
- ▶ The grad student responsible for measuring $x \in X$, should have instructions which measurement $\pi(x) \in Y$ to use instead.
- ▶ Given a result for those measurements, should be able to determine the outcome to output.
- ▶ The outcome statistics should be identical to those of e .

We first define basic simulations. Dependencies on multiple measurements and stochastic processing added as a comonadic effect.

Basic simulations



To simulate B using A :

- ▶ map inputs of B (measurements) to inputs of A
- ▶ run A
- ▶ map outputs of A (measurement outcomes) to outputs of B

Basic simulations formally

A morphism of scenarios $(\pi, h) : \langle X, \Sigma, O \rangle \rightarrow \langle Y, \Theta, P \rangle$ is given by:

- ▶ A simplicial map $\pi : \Theta \rightarrow \Sigma$.
- ▶ For each $y \in Y$, a map $h_y : O_{\pi(y)} \rightarrow P_y$.

Simpliciality of π means that contexts in Θ are mapped to contexts in Σ .

Basic simulations formally

A morphism of scenarios induces a natural action on empirical models:

Given $e: \langle X, \Sigma, O \rangle$, then $(\pi, h)^* e: \langle Y, \Theta, P \rangle$ is given by, for $\tau \in \Delta$:

$$((\pi, h)^* e)_\tau = \text{Prob}(\gamma)(e_{\pi(\tau)})$$

the push-forward of the probability measure $e_{\pi(\tau)}$ along the map

$$\gamma: \prod_{x \in \pi(\tau)} O_x \longrightarrow \prod_{y \in \tau} P_y$$

given by $\gamma(s)_y = h_y(s_{\pi(y)})$.

This gives a category **Emp**, with:

- ▶ objects are empirical models $e: \langle X, \Sigma, O \rangle$,
- ▶ morphisms $e \rightarrow d$ are simulations $(\pi, h): \langle X, \Sigma, O \rangle \rightarrow \langle Y, \Theta, P \rangle$ such that $(\pi, h)^* e = d$.

Basic simulations

This is a simple notion of simulation, but already covers several things:

- ▶ It allows for relabelling of measurements and outcomes (**isomorphisms**).
- ▶ Since the map on measurements needn't be surjective, it allows for simulating B using only part of A (**restrictions**).
- ▶ Since the map on outcomes needn't be injective, it allows for **coarse-graining** of outcomes.

Note that mappings of inputs go backward, of outputs forward:

- ▶ Akin to the Hom functor being **contravariant** in its first argument, **covariant** in its second.
- ▶ Logically, to reduce one implication to another, one must **weaken** the antecedent and **strengthen** the consequent.

Beyond deterministic maps

Basic simulations are useful, but limited.

To allow **adaptive** use of the resource, we introduce **measurement protocols**.

These protocols proceed iteratively by first performing a measurement over the given scenario, and then conditioning their further measurements on the observed outcome.

Note that different paths can lead into different, **incompatible** contexts.

Thus they incorporate adaptive classical processing, of the kind used e.g. in Measurement-Based Quantum Computing.

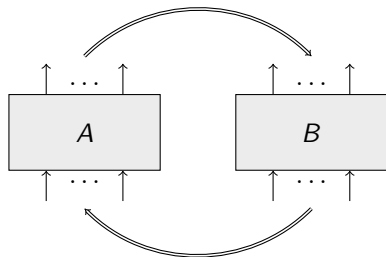
Previously considered in:

'A combinatorial approach to nonlocality and contextuality'

Acin, Fritz, Leverrier, Sainz, Communications in Mathematical Physics, 2015.

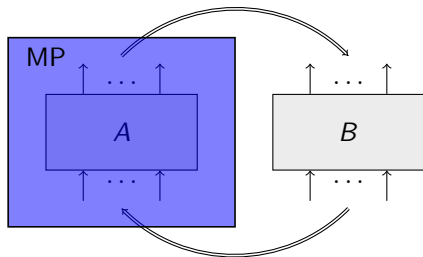
Formally, we construct a **comonad** MP on the category of empirical models, where $\text{MP}(e: \langle X, \Sigma, O \rangle)$ is the model obtained by taking all measurement protocols over the given scenario.

Basic simulations

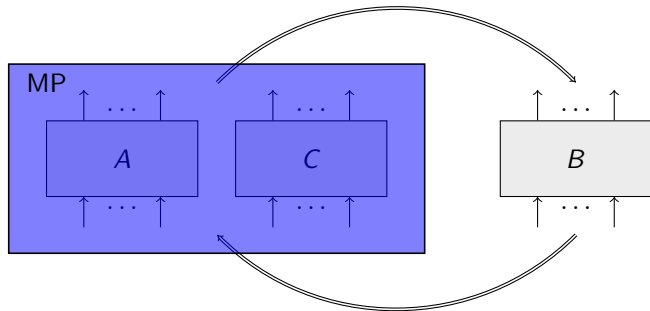


For more expressive simulations, increase the power of A **classically**

CoKleisli simulations



Cokleisli simulations with shared randomness



Requirement: C is noncontextual.

The MP construction

Given a scenario $\mathbf{X} = \langle X, \Sigma, O \rangle$ we build a new scenario $\text{MP}(\mathbf{X})$, where:

- ▶ measurements are the measurement protocols on \mathbf{X}

$$\text{MP}(\langle X, \Sigma, O \rangle) ::= \emptyset \mid (x, f) \quad \text{where } x \in X \text{ and } f: O_x \rightarrow \text{MP}(\langle X \setminus \{x\}, \text{lk}_x \Sigma, O \rangle).$$

- ▶ outcomes are the joint outcomes observed during a run of the protocol
- ▶ measurement protocols are compatible if they can be combined consistently
 - ▶ A run is a sequence $\bar{x} = (x_i, o_i)_{i=1}^l$ with $x_i \in X$, $o_i \in O_{x_i}$
 - ▶ $\sigma_{\bar{x}} = \{x_1, x_2, \dots, x_l\} \in \Sigma$.
 - ▶ Two runs (of different protocols) are consistent if they agree on common measurements
 - ▶ Protocols $\{Q_1, \dots, Q_n\}$ are compatible if for any choice of pairwise consistent runs \bar{x}_i from Q_i , we have $\bigcup_i \sigma_{\bar{x}_i} \in \Sigma$

The MP construction

Empirical models in \mathbf{X} are then naturally lifted to this scenario $\text{MP}(\mathbf{X})$.

Proposition

MP defines a comonoidal comonad on the category \mathbf{Emp} of empirical models.

Roughly: comultiplication $\text{MP}(\mathbf{X}) \rightarrow \text{MP}^2(\mathbf{X})$ by “flattening”, unit $\text{MP}(\mathbf{X}) \rightarrow \mathbf{X}$,
and $\text{MP}(\mathbf{X} \otimes \mathbf{Y}) \rightarrow \text{MP}(\mathbf{X}) \otimes \text{MP}(\mathbf{Y})$

General simulations

Given empirical models e and d , a **simulation** of e by d is a map

$$d \otimes c \rightarrow e$$

in \mathbf{Emp}_{MP} , the coKleisli category of MP, i.e. a map

$$\text{MP}(d \otimes c) \rightarrow e$$

in \mathbf{Emp} , for some noncontextual model c .

The use of the noncontextual model c is to allow for classical randomness in the simulation.

We denote the existence of a simulation of e by d as $d \rightsquigarrow e$, read “ d simulates e ”.

Results

In the resulting category/resource theory,

- ▶ Contextual fraction is a monotone
- ▶ an empirical model is contextual iff it cannot be simulated from the trivial model
- ▶ an empirical model is logically noncontextual iff it cannot be simulated possibilistically from the trivial model
- ▶ an empirical model e is strongly contextual iff no model e' contained in the support of e can be simulated from the trivial model

Moreover, no-cloning holds: a simulation $e \rightarrow e \otimes e$ exists iff e is noncontextual.

The viewpoints agree

Theorem

Let $e : \mathbf{X}$ and $d : \mathbf{Y}$ be empirical models. Then $d \rightsquigarrow e$ if and only if there is a typed term $a : \mathbf{Y} \vdash t : \mathbf{X}$ such that $t[d/a] \simeq e$.

Proof.

(Sketch) If $d \rightsquigarrow e$, then e can be obtained from $\text{MP}(d \otimes x)$ by a combination of a coarse-graining and a measurement translation. There is a term representing x and MP can be built by repeated controlled measurements.

For the other direction, build a simulation $d \rightarrow t[d/a]$ inductively on the structure of t . □

Properties of NCF

- ▶ $\text{NCF}(e) = 1$ iff e is noncontextual
- ▶ $\text{NCF}(\text{MP}(e)) = \text{NCF}(e)$
- ▶ $\text{NCF}(e \otimes d) = \text{NCF}(e)\text{NCF}(d)$
- ▶ $d \rightsquigarrow e$ implies $\text{NCF}(d) \leq \text{NCF}(e)$

No-cloning

Theorem (No-cloning)

$e \rightsquigarrow e \otimes e$ if and only if e is noncontextual.

Proof.

The direction " \Leftarrow " is easy as there is an obvious deterministic simulation $\text{MP}(e \otimes e) \rightarrow (e \otimes e)$. For the other direction, we rule out $0 < \text{NCF}(e) < 1$ and $\text{NCF}(e) = 0$. So assume $0 < \text{NCF}(e) < 1$ and $e \rightsquigarrow e \otimes e$. Then

$$\text{NCF}(e) \leq \text{NCF}(e \otimes e) = \text{NCF}(e)^2 < \text{NCF}(e),$$

a contradiction. □

Cont.

Remains to prove $\text{NCF}(e) > 0$. Now $e \rightsquigarrow e \otimes e$ implies $e \rightsquigarrow e^{\otimes n}$ for any n . Moreover, we may assume that the noncontextual resource gets measured first. But then it merely randomizes which deterministic simulation $\text{MP}(\mathbf{X}^{\otimes n}) \rightarrow \text{MP}(X)$ to use, and there is only finitely many of these. Thus, for large enough n and some possible value of the random data, too many of the components end up using the same deterministic simulation. As the underlying map $\Sigma^{*n} \rightarrow \Sigma$ is simplicial, they must thus use a compatible (and hence noncontextual) part of X . This gives $\text{NCF}(e) > 0$. □

The talk so far is more or less based on

'A comonadic view of simulation and quantum resources'

Abramsky, Barbosa, Karvonen, Mansfield, LiCS 2019.

Let's do something new!

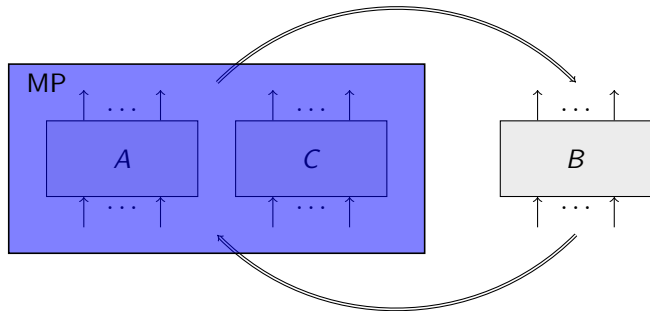
More expressive simulations?

When asking if d simulates e , several further relaxations are possible:

- ▶ does (π, h) take d to (a subset of) e on the level of supports?
- ▶ does d simulate e approximately?
- ▶ does $d^{\otimes n}$ simulate e for some n ?
- ▶ is there a sequence of approximate simulations $d^{\otimes n} \rightarrow e$ converging to e ?
- ▶ what if we allow the free resource c to come from some larger class of models?

We will discuss the last one.

More expressive simulations?



Requirement: ~~C is noncontextual.~~ C is in some fixed class \mathcal{F} .

More expressive simulations?

Let \mathcal{F} be a class of models closed under \otimes that contains noncontextual models. Given empirical models d and e , an \mathcal{F} -simulation $d \rightarrow e$ consists of a deterministic simulation $\text{MP}(d \otimes f) \rightarrow e$ for some $f \in \mathcal{F}$. If d \mathcal{F} -simulates e , we write $d \rightsquigarrow_{\mathcal{F}} e$

One could ask if $\rightsquigarrow_{\mathcal{F}}$ allows for catalysis: a situation where e doesn't \mathcal{F} -simulate f yet $d \otimes e \rightsquigarrow_{\mathcal{F}} d \otimes f$.

Theorem (No-catalysis)

If $d \otimes e \rightsquigarrow_{\mathcal{F}} d \otimes f$, then $e \rightsquigarrow_{\mathcal{F}} f$

NB: The case when $d = f$ and e is non-contextual is no-cloning.

Contrast with entanglement, where catalysis is possible:

'Entanglement-assisted local manipulation of pure quantum states'

Jonathan, Plenio, PRL 1999.

Further questions

- ▶ Study the preorder induced by $d \rightsquigarrow e$.
- ▶ Study various relaxed simulations
- ▶ Comparison with other approaches to contextuality.
- ▶ Multipartite non-locality
- ▶ Graded structure on the comonad?
- ▶ MBQC?
- ▶ Generating all empirical models?
- ▶ Bell inequalities?

Summary

- ▶ Intraconversions of contextual resources formalized in terms of
 - ▶ free operations
 - ▶ simulations
- ▶ These viewpoints agree and capture known examples
- ▶ A no-cloning result
- ▶ No-catalysis
- ▶ Several avenues for further work