

A unified approach to contextuality and violations of macrorealism

Shane Mansfield



THE UNIVERSITY *of* EDINBURGH
informatics

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Overview

1. Unified approach to nonlocality & contextuality
2. Example of experiment with non-trivial causal structure: double-slit
3. Causal measurement scenarios and causal empirical models
4. Unified description of
{nonlocality, contextuality, violations of macrorealism}

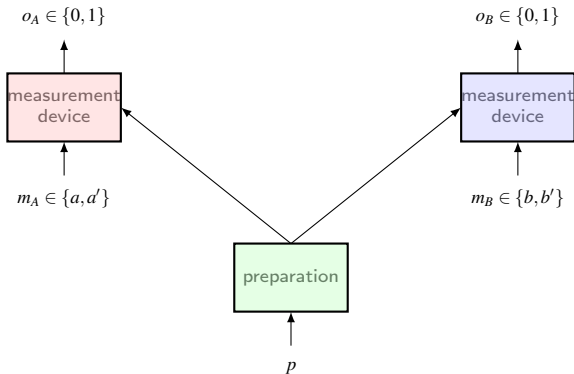
Remark

- This work extends the *sheaf-theoretic framework* for contextuality initiated by Abramsky & Brandenburger
- Key additional ingredient: causal structure
- Alternative unified approach in the *Contextuality-by-Default framework*: Kujala-Dzhafarov-Larsson [PRL 115, 150401 (2015)]

Nonlocality & contextuality recap

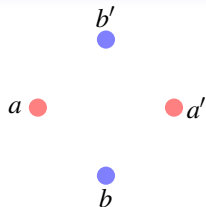
Empirical Data (e.g. CHSH experiment)

	(0,0)	(0,1)	(1,0)	(1,1)
(a,b)	1/2	0	0	1/2
(a,b')	3/8	1/8	1/8	3/8
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Measurement Scenarios: CHSH

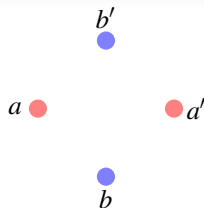
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A *measurement scenario* is a triple $\langle X, \mathcal{M}, O \rangle$ where:

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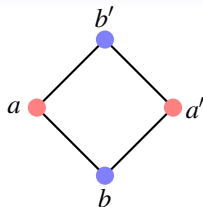
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$$X = \{a, a', b, b'\}$$

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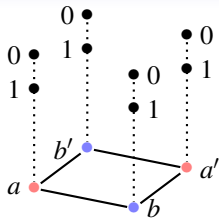
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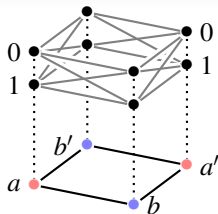
O a finite set of outcomes — e.g.

$$O = \{0, 1\}$$

Empirical Models

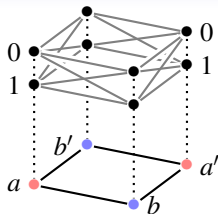
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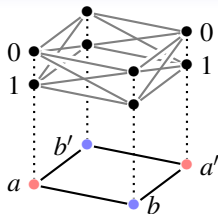


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- *Empirical model*: family $\{e_C\}_{C \in \mathcal{M}}$ where each $e_C \in \text{Prob}(\mathcal{O}^C)$
- i.e. a distribution for each context:

$$e_{\{a,b\}} = \text{prob}(o_a, o_b | a, b), \quad \dots, \quad e_{\{a',b'\}} = \text{prob}(o_{a'}, o_{b'} | a', b')$$

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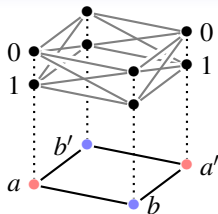
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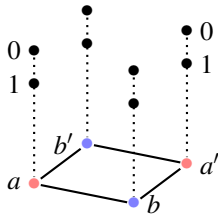
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Classical Correlations

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Classical data arises as a convex combination of *global assignments*:

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 (a, a', b, b') &\mapsto (0, 0, 0, 0), \\
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 &\dots, \\
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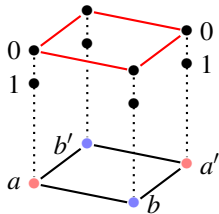


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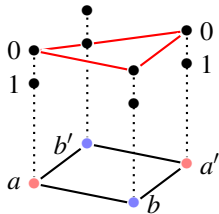
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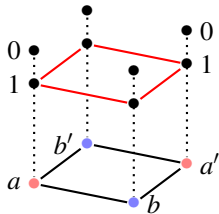


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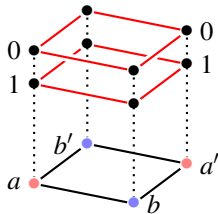


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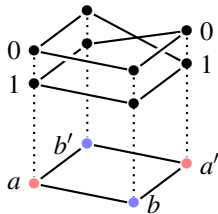


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However, the above correlations *cannot* be obtained as a convex combination of global assignments!

PR box bundle diagram

Logical contextuality:

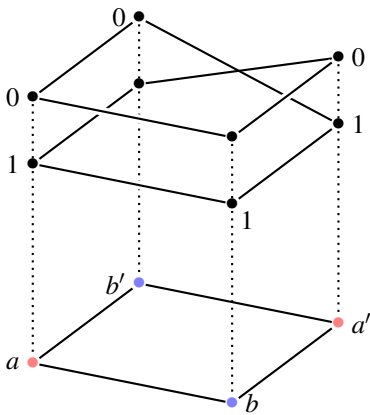
present at the level of possibilities

Strong contextuality:

no event can be extended to a global assignment

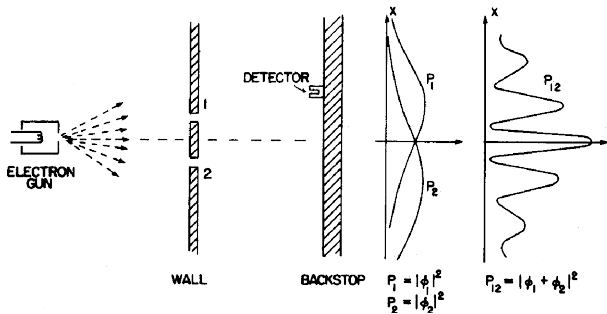
PR box possibility table:

	(0,0)	(0,1)	(1,0)	(1,1)
(a,b)	✓	×	×	✓
(a,b')	✓	×	×	✓
(a',b)	✓	×	×	✓
(a',b')	×	✓	✓	×



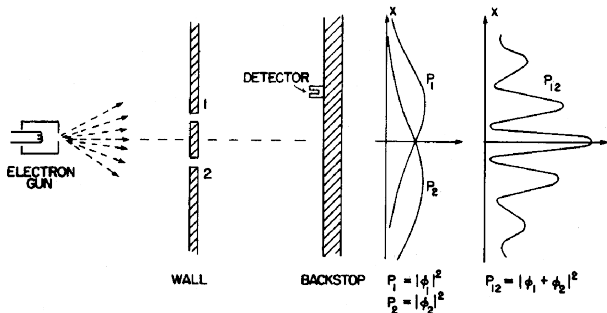
Example: the double-slit experiment

A classic experiment



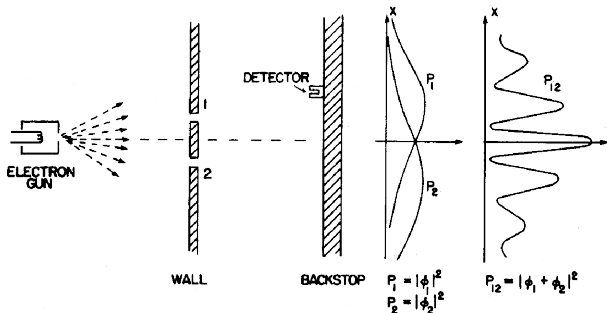
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- Turn down intensity to shoot one electron at a time
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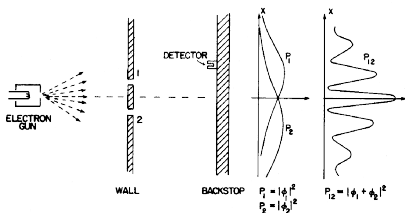
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- Wave-particle duality mysticism, etc. ...

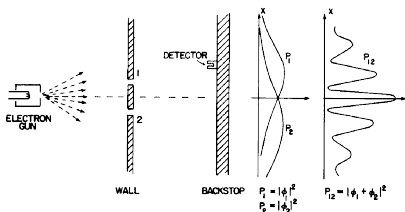
A classic experiment — empirical model?

Position detector in dark fringe



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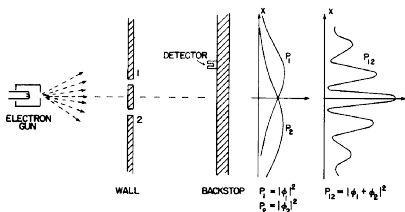


Measurements $X = \{x_e, x_w, x_d\}$,

1. x_e — *emission*,
values in $\{0, 1\}$
2. x_w — *which way*
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values in $\{0, 1, 2\}$
3. x_d — *detection*,
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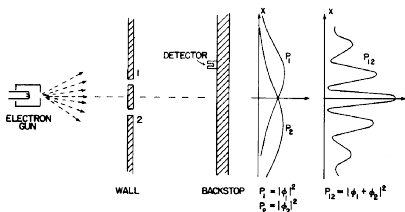
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Contexts,

$$\mathcal{M} = \{ \{x_e, x_w, x_d\}, \{x_e, x_d\} \}$$

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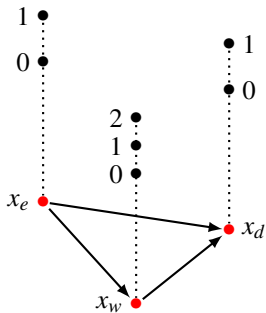
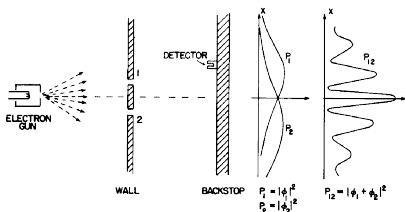
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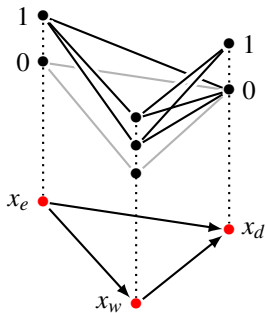
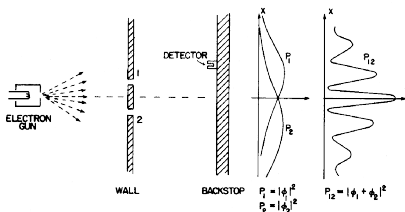
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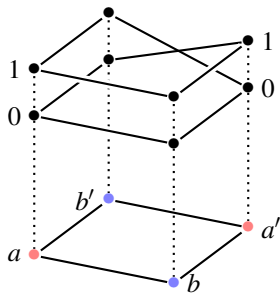
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Trouble with marginals

- Empirical models satisfy 'local' consistency:

$$\text{prob}(o_1|a,b) = \text{prob}(o_1|a,b') = \text{prob}(o_1|a), \text{ etc.}$$

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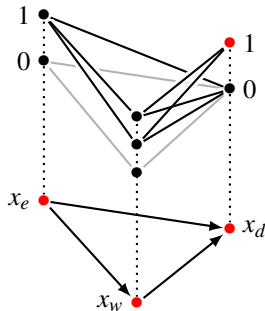
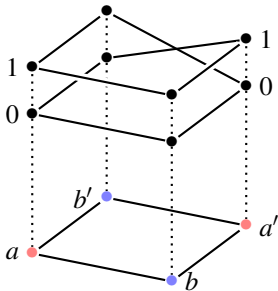
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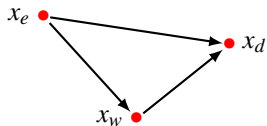
- Not the case here:

$$\text{prob}(1_d | x_d, x_e) \neq \text{prob}(1_d | x_d, x_w, x_e)$$



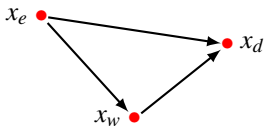
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- Marginals only make sense if they respect causal structure



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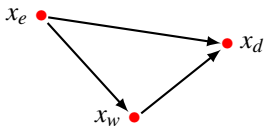
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$$A \leq B \quad \text{iff} \quad \downarrow A \cap B = A$$

i.e. B reduces to A via successive removals of maximal elements

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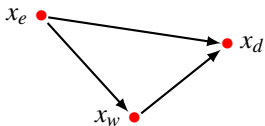
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- e.g. $\{x_e, x_w\} \leq \{x_e, x_w, x_d\}$ but $\{x_e, x_d\} \not\leq \{x_e, x_w, x_d\}$

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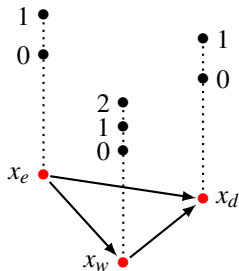
i.e. B reduces to A via successive removals of maximal elements

- e.g. $\{x_e, x_w\} \leq \{x_e, x_w, x_d\}$ but $\{x_e, x_d\} \not\leq \{x_e, x_w, x_d\}$
- Intuition: *later* measurements cannot affect *earlier* outcomes
Earlier measurements can affect *later* outcomes

Causal empirical models

Causal measurement scenarios

A *causal measurement scenario* is a tuple $\langle X, \leq, \mathcal{M}, O \rangle$ where:



- (X, \leq) a partially ordered set of measurements — e.g.

$$X = \{x_e, x_w, x_d\}$$

- \mathcal{M} the maximal contexts (wrt ' \leq ') — e.g.

$$\mathcal{M} = \{\{x_e, x_d\}, \{x_e, x_w, x_d\}\}$$

- O a finite set of outcomes — e.g.

$$O = \{0, 1, 2\}$$

Causal empirical models

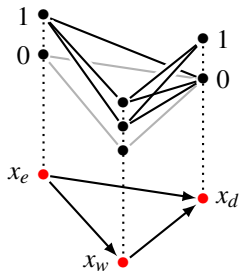
Fix a causal measurement scenario

$\langle X, \leq, \mathcal{M}, \mathcal{O} \rangle$

Causal empirical model: family $\{e_C\}_{C \in \mathcal{M}}$ where each $e_C \in \text{Prob}(\mathcal{O}^C)$

- i.e. a distribution for each context:

$$e_{\{x_e, x_w, x_d\}} = \text{prob}(o_e, o_w, o_d \mid x_e, x_w, x_d), \quad e_{\{x_e, x_d\}} = \text{prob}(o_e, o_d \mid x_e, x_d)$$

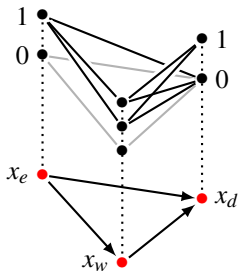


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- 'Local' consistency of causal marginals:

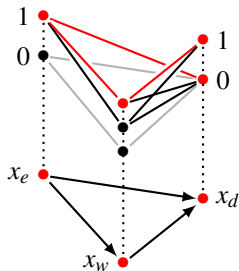
$$\text{prob}(o_e \mid x_e, x_w, x_d) = \text{prob}(o_e \mid x_e, x_w) = \text{prob}(o_e \mid x_e), \text{ etc.}$$

generalised no signalling (outside of forward light-cone)

Contextuality and macrorealism

Double-slit model is logically contextual

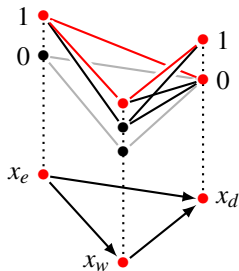
i.e. cannot be obtained as convex combination of global assignments



Contextuality and macrorealism

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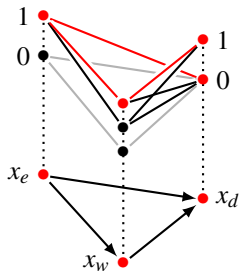
For totally ordered scenarios, *macrorealism* (Leggett-Garg) has 3 requirements (Maroney-Timpson):

1. Macrorealism *per se* — mixtures of deterministic hidden states
2. Non-invasiveness — parameter-independence (forwards in time)
3. Induction — parameter-independence (backwards in time)

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Macrorealism \longleftrightarrow non-contextuality

Special causal scenarios

1. (Standard/static) empirical models:
' \leq ' is trivial *or* each maximal context contains only pairwise incomparable elements
2. Leggett-Garg-type models:
' \leq ' is a total order
3. Temporal Bell-type models (Fritz):
 X can be partitioned into time steps as $(X_t)_{t \in \mathbb{N}}$, such that:
 - $\mathcal{M} = \prod X_t$
(contexts contain one measurement per time step)
 - For all $x \in X_t, y \in X_{t'}$ we have $x \leq y$ iff $t \leq t'$
(causal order is induced by the partition)

Conclusion

- Unified approach to

{nonlocality, contextuality, violations of macrorealism}

Representative examples:

{Bell, Kochen-Specker, Leggett-Garg}

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Future directions (relating to other talks):

- Quantifying 'contextuality' in causal scenarios (Rui)
- Process calculus (Rui)
- Advantages over classical causal models in informatic tasks (Samson)
- Extension to indefinite causal structures (Philip)
- Clarify relation to unified approach in CbD (Ehtibar)