Monogamy of non-locality and macroscopic averages

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Royal Holloway, University of London 5th February 2015



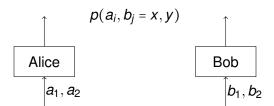
- Average macro correlations from micro models are local (Ramanathan, Paterek, Kay, Kurzyński & Kaszlikowski 2011: multipartite guantum models)
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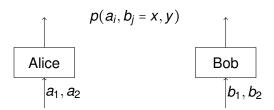
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- ► This talk: we mainly consider a simple illustrative example.

Non-locality

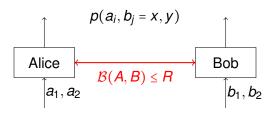


Non-locality

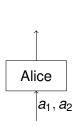


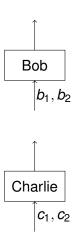
	00	01	10	11
a_1b_1	1/2	0	0	1/2
a_1b_2	3/8	1/8	1/8	3/8
a_2b_1	3/8	1/8	1/8	3/8
a_2b_2	1/8	3/8	3/8	1/8

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Empirical model: no signalling probabilities

$$p(a_i,b_j,c_k=x,y,z)$$

where x, y, z are possible outcomes.

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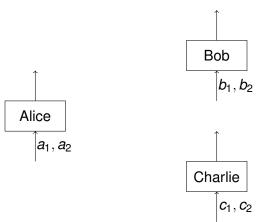
Consider the subsystem composed of A and B only, given by marginalisation (in QM, partial trace):

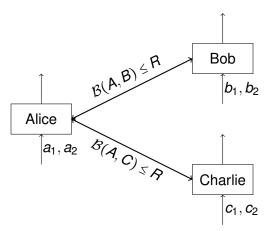
$$p(a_i, b_j = x, y) = \sum_{z} p(a_i, b_j, c_k = x, y, z)$$

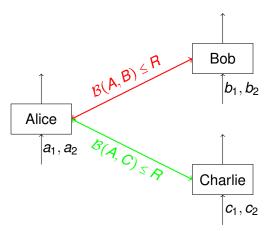
(this is independent of c_k due to no-signalling).

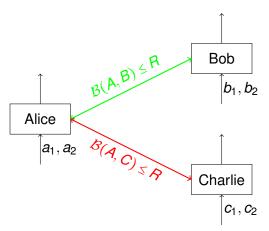
Similarly define $p(a_i, c_k = x, z)$. (A and C)



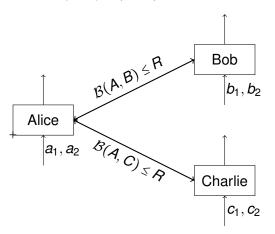








Given a Bell inequality $\mathcal{B}(-,-,) \leq R$,



Monogamy relation: $\mathcal{B}(A, B) + \mathcal{B}(A, C) \leq 2R$



Macroscopic average behaviour

Macroscopic measurements

- (Micro) dichotomic measurement: a single particle is subject to an interaction a and collides with one of two detectors: outcomes 0 and 1.
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- ► The interaction is probabilistic: p(a = x), x = 0, 1.
- Consider beam (or region) of N particles, differently prepared.
- Subject each particle to the interaction a: the beam gets divided into 2 smaller beams hitting each of the detectors.
- Outcome represented by the intensity of resulting beams: $I_a \in [0,1]$ proportion of particles hitting the detector 1.
- We are concerned with the mean, or expected, value of such intensities.



Macroscopic average behaviour

This mean intensity can be interpreted as the average behaviour among the particles in the beam or region: if we would randomly select one of the N particles and subject it to the microscopic measurement a, we would get outcome 1 with probability I_a:

$$I_a = \sum_{i=1}^{N} p_i(a=1)$$
.

The situation is analogous to statistical mechanics, where a macrostate arises as an averaging over an extremely large number of microstates, and hence several different microstates can correspond to the same macrostate.

Macroscopic average behaviour: multipartite

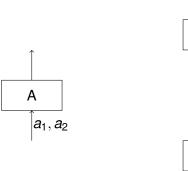
- Multipartite macroscopic measurements:
 - several 'macroscopic' sites consisting of a large number of microscopic sites/particles;
 - several (macro) measurement settings at each site.
- Average macroscopic Bell experiment: the (mean) values of the macroscopic intensities indicate the behaviour of a randomly chosen tuple of particles: one from each of the beams, or sites.

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- Average macroscopic Bell experiment: the (mean) values of the macroscopic intensities indicate the behaviour of a randomly chosen tuple of particles: one from each of the beams, or sites.
- We shall show that, as long as there are enough particles (microscopic sites) in each macroscopic site, such average macroscopic behaviour is always local no matter which no-signalling model accounts for the underlying microscopic correlations.

Macroscopic average behaviour: tripartite example

Consider again the tripartite scenario:



В

Macroscopic average behaviour: tripartite example

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- We regard sites B and C as forming one 'macroscopic' site, M, and site A as forming another.
- In order to be 'lumped together', B and C must be symmetric/of the same type: the symmetry identifies the measurements $b_1 \sim c_1$ and $b_2 \sim c_2$, giving rise to 'macroscopic' measurements m_1 and m_2 .

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- Given an empirical model p(a_i, b_j, c_k = x, y, z), the 'macroscopic' average behaviour is a bipartite model (with two macro sites A and M) given by the following average of probabilities of the partial models:

$$p_{a_i,m_j}(x,y) = \frac{p_{a_i,b_j}(x,y) + p_{a_i,c_j}(x,y)}{2}$$



Example: W state

Z and X measurements on the W state:

	000	001	010	011	100	101	110	111
$a_1b_1c_1$	9	1	1	1	1	1	1	9
$a_1b_1c_2$	8	2	0	2	0	2	8	2
$a_1b_2c_1$	8	0	2	2	0	8	2	2
$a_1b_2c_2$	4	4	4	0	4	4	4	0
$a_2b_1c_1$	8	0	0	8	2	2	2	2
$a_2b_1c_2$	4	4	4	4	4	0	4	0
$a_2b_2c_1$	4	4	4	4	4	4	0	0
$a_2b_2c_2$	0	8	8	0	8	0	0	0
(every entry should be divided by 24)								

Example: W state

	00	01	10	11
$a_1 m_1$	10	2	2	10
$a_1 m_2 a_2 m_1$	8	4	8	4
	8	8	4	4
a_2m_2	8	8	8	0

(every entry should be divided by 24)

This is local!

Another example model

	000	001	010	011	100	101	110	111
$a_1b_1c_1$	1	1	0	0	0	0	1	1
$a_1b_1c_2$	1	1	0	0	0	0	1	1
$a_1b_2c_1$	1	1	0	0	0	0	1	1
$a_1b_2c_2$	1	1	0	0	0	0	1	1
$a_2b_1c_1$	1	1	0	0	0	0	1	1
$a_2b_1c_2$	1	1	0	0	0	0	1	1
$a_2b_2c_1$	0	0	1	1	1	1	0	0
$a_2b_2c_2$	0	0	1	1	1	1	0	0
(every entry should be divided by 4)								

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	00	01	10	11		00	01	10	11
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a_1b_2					a_1c_2	1	1	1	1
a_2b_1					a_2c_1	1	1	1	1
a_2b_2	0	2	2	0	a_2c_2	1	1	1	1
(divided by 4)					(divid	ed b	y 4)	

maximally non-local

local

	00	01	10	11
$\overline{a_1 m_1}$	3	1	1	3
$a_1 m_1$	3	1	1	3
$a_1 m_1$	3	1	1	3
$a_1 m_1$	1	3	3	1

(every entry should be divided by 8)

Again, this is local!

Monogamy of non-locality and macroscopic averages

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$$\mathcal{B}(A, M) \leq R$$

$$\Leftrightarrow \sum_{i,j,x,y} \alpha(i,j,x,y) p(a_i, m_j = x,y) \leq R$$

$$\Leftrightarrow \sum_{i,j,x,y} \alpha(i,j,x,y) \frac{p(a_i, b_j = x,y) + p(a_i, c_j = x,y)}{2} \leq R$$

$$\Leftrightarrow \sum_{i,j,x,y} \alpha(i,j,x,y) p(a_i, b_j = x,y) + \sum_{i,j,x,y} \alpha(i,j,x,y) p(a_i, c_j = x,y) \leq 2R$$

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The average model p_{a_i,m_j} satisfies the inequality if and only if in the microscopic model Alice is **monogamous** with respect to violating it with Bob and Charlie.



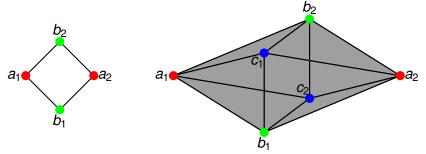
- In the two examples above, the average models were local.
- Equivalently, the examples satisfied the monogamy relation for any Bell inequality.
- This is true for all no-signalling empirical models on the tripartite scenario under consideration, with two measurement settings per site.
- We give a structural explanation for this...
- ... which generalises well beyond this particular scenario.

Vorob'ev's theorem

Abramsky-Brandenburger framework

Measurement scenarios:

- a finite set of measurements X;
- a cover U of X (or an abstract simplicial complex Σ on X), indicating joint measurability or compatibility of measurements.

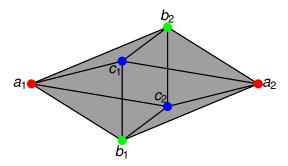


Examples: Bell-type scenarios, KS configurations, and more.

Abramsky–Brandenburger framework

E.g. the tripartite scenario:

$$X = \{a_1, a_2, b_1, b_2, c_1, c_2\}$$
$$\mathcal{U} = \{\{a_i, b_j, c_k\} \mid i, j, k \in \{1, 2\}\}$$



Abramsky-Brandenburger framework

No-signalling **empirical model**:

- ▶ a family $(p_C)_{C \in \mathcal{U}}$, where p_C is a probability distribution on the outcomes of measurements in context C.
- compatibility condition: p_C and $p_{C'}$ marginalise to the same distribution on the outcomes of measurements in $C \cap C'$. (on multipartite scenarios: no-signalling)

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An empirical model admits a **local/non-contextual hidden variable** explanation (in the sense of Bell's theorem) iff there exists a **global distribution** p_X (i.e. for all measurements at the same time) that marginalises to all the p_C .

Obstructions to such extensions are witnessed by violations of general Bell inequalities.



Vorob'ev's theorem

For which measurement compatibility structures Σ is it so that **any** no-signalling empirical model admits a global extension, i.e. is local/non-contextual?

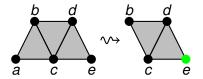
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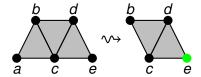
Vorob'ev(1962) derived a **necessary and sufficient** combinatorial condition on Σ for this to be the case.

 Turns out to be equivalent to the notion of acyclicity of a database schema.

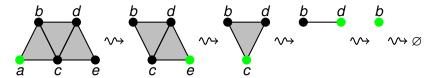
Graham reduction step: delete a measurement that belongs to only one maximal context.



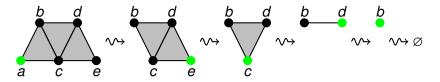
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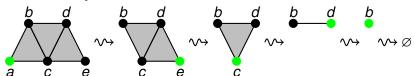
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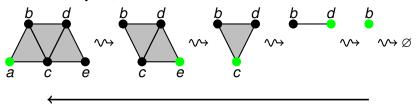
Theorem (Vorob'ev 1962, adapted)

All empirical models on Σ are extendable iff Σ is acyclic

If Σ is acyclic,



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then construct a global distribution by glueing

Given distributions P_{ab} over $\{a,b\}$ and P_{bc} over $\{b,c\}$ compatible on b,

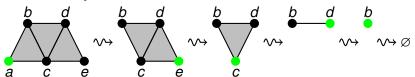
$$\sum_{x\in O}P(a,b=x,y)=\sum_{z\in O}P(b,c=y,z)\;,$$

we can define an extension:

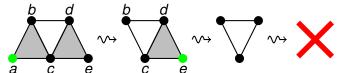
$$P(a,b,c=x,y,z) = \frac{P(a,b=x,y)P(b,c=y,z)}{P(b=y)}$$
.



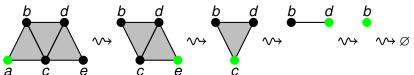
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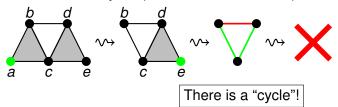
• If Σ is not acyclic (Graham reduction fails).



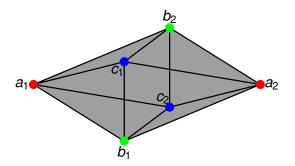
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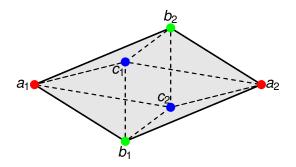
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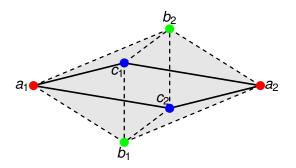
A structural explanation



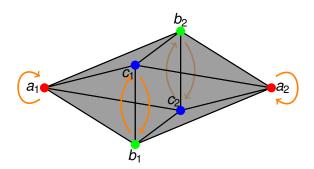
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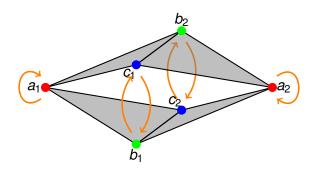


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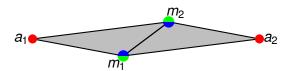
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- The macro scenario arises as a quotient.



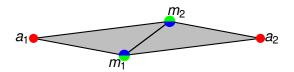


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- We identify *B* and *C*: $b_1 \sim c_1$, $b_2 \sim c_2$.
- ▶ The macro scenario arises as a quotient.



- This quotient complex is acyclic.
- Therefore, no matter from which micro model p_{a_i,b_j,c_k} we start, the averaged macro correlations p_{a_i,m_i} are local.
- In particular, they satisfy any Bell inequality. Hence, the original tripartite model also satisfies a monogamy relation for any Bell inequality.

- ▶ n macroscopic sites A, B, C, . . .
- k_i measurement settings at site i
- take r_i copies of each site i, or r_i micro sites constituting i. For a macro site A:
 - copies / micro sites: $A^{(1)}, \ldots, A^{(r_1)}$
 - measurement settings at $A^{(m)}$: $a_1^{(m)}, \ldots, a_{k_a}^{(m)}$

Scenario:
$$\Sigma_{n,\vec{k},\vec{r}} := \mathfrak{D}_{k_1}^{(*r_1)} * \cdots * \mathfrak{D}_{k_n}^{(*r_n)}$$
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$$\Sigma_{n,\vec{k},\vec{r}} := \mathfrak{D}_{k_1}^{(*r_1)} * \cdots * \mathfrak{D}_{k_n}^{(*r_n)}$$
.

Symmetry by $S_{r_1} \times \cdots \times S_{r_n}$ identifies the copies at each macro site.

$$a_j^{(1)} \sim \cdots \sim a_j^{(r_A)} \quad (\forall j \in \{1, \dots, k_A\}),$$

 $b_j^{(1)} \sim \cdots \sim a_j^{(r_A)} \quad (\forall j \in \{1, \dots, k_A\}),$
etc.



Proposition

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The quotient of the measurement scenario $\Sigma_{n,\vec{k},\vec{r}}$ by the symmetry above is acyclic iff one of the following holds:

- each site has at least as many microscopic sites or copies as it has measurement settings, i.e. $\forall_{i \in \{1,...,n\}}$. $k_i \leq r_i$;
- one of the sites has a single copy and the condition above is satisfied by all the other sites, i.e.

$$\exists_{i_0}.\ \left(r_{i_0}=1\ \land\ \forall_{i\in\{1,\ldots\widehat{i_0}\ldots,n\}}.\ k_i\leq r_i\right).$$

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We get monogamy relations

$$\sum_{m_B=1}^{r_B} \sum_{m_C=1}^{r_C} \cdots \mathcal{B}(A, B^{(m_B)}, C^{(m_C)}, \ldots) \leq r_B r_C \cdots R$$



Conclusions

Summary

- A symmetry (G-action) on Σ identifies measurements.
- A model satisfies a G-monogamy relation for a Bell inequality iff the macro average correlations (quotient model by G) satisfy the Bell inequality.
- So, if the quotient scenario is acyclic, then any no-signalling empirical model is G-monogamous wrt to all Bell inequalities (since the average correlations, being defined in this quotient scenario, must be local/non-contextual).

Summary

- In particular, we proved that this is the case for multipartite Bell-type scenarios provided the number of parties being identified as belonging to each 'macro' site is larger than the number of measurement settings available to each of them.
- Our approach highlights the reason why monogamy relations for general multipartite Bell inequalities follow from no-signalling alone, generalising the result of Pawłowski and Brukner (2009) from bipartite to multipartite. It also shows that what Ramanathan et al. proved holds not only for QM but for any no-signalling theory.
- The approach is not restricted to multipartite Bell-type scenarios. More generally, we can apply the same ideas to derive monogamy relations for contextuality inequalities.

Questions...

