

# Miscellaneous Comments on (Quantum) Contextuality

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## Preamble

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  - because the term comes from QM historically,
  - and because in QM we have the traditional examples of contextual systems.
- Otherwise, contextuality is not about QM specifically.
  - (One could even say, it has nothing to do with QM.)
- Contextuality is all about random variables, more specifically, about **identities** of random variables.

What are random variables, mathematically?



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- Mathematically, a random variable is (identified by) a measurable mapping  $D \xrightarrow{\text{measurable}} C$ ,
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- Mathematically, elements of a set  $\mathbb{X}$  of random variables are jointly distributed,  $\text{JD}(\mathbb{X})$ , iff they have the same domain probability space  $(D, \Sigma_D, \mu)$ .
- Ergo: if  $\mathbb{A}$  and  $\mathbb{B}$  are sets of random variables with  $\mathbb{A} \cap \mathbb{B} \neq \emptyset$ ,

$$\text{JD}(\mathbb{A}) \ \& \ \text{JD}(\mathbb{B}) \Rightarrow \text{JD}(\mathbb{A} \cup \mathbb{B}) \quad (\text{“agglutinativity”}).$$

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  - in particular, random variables  $A$  and  $B$  generated by Hermitian operators  $\overline{A}, \overline{B}$  with  $[\overline{A}, \overline{B}] \neq 0$  may very well be measured “together” (e.g., in succession).



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  - and if  $[\overline{A}, \overline{B}] = 0$ , the  $A$  and  $B$  they generate may very well be recorded in different, mutually exclusive contexts.

A system of random variables: contents and contexts

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★	★			$c = 1$
	★	★		$c = 2$
		★	★	$c = 3$
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# A system of random variables: contents and contexts

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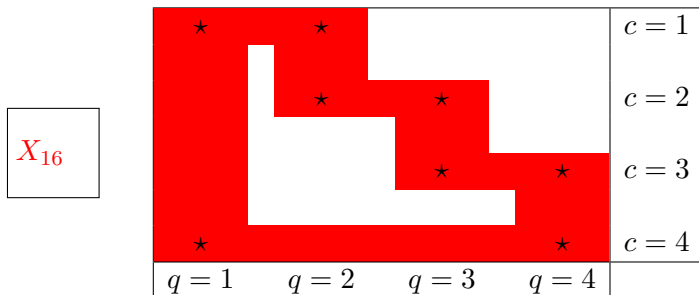
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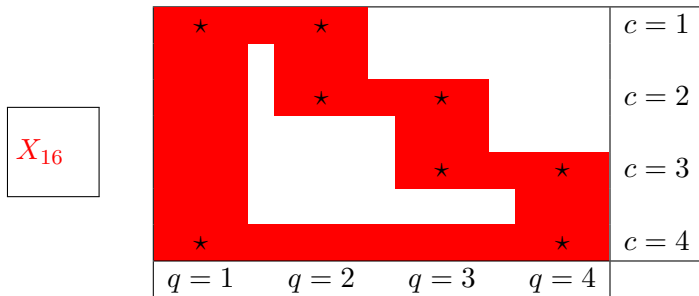
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## Contextuality of a system of random variables



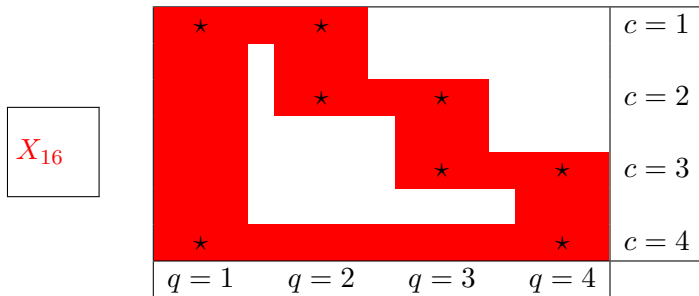
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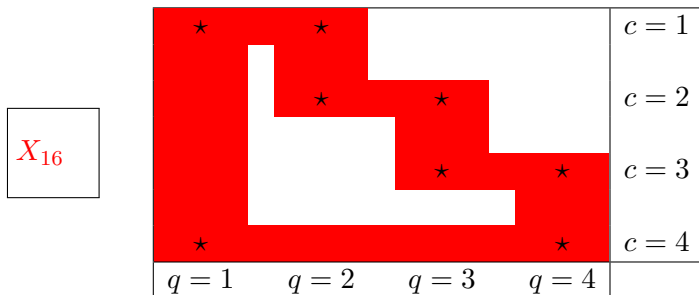
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- Each  $R_i^j = f_i(X_{16}) \Rightarrow 9$  linear equations with 16 unknown probabilities  $\Rightarrow$  may or may not have a (nonnegative) solution.
- If no solution exists, we have to reject the assumption (the only one made) that the content-sharing variables are the same.

# Kochen-Specker system (Cabello-Estebarez-Alcaine)

$C_9$										*		*		*	*			
$C_8$					*					*			*	*	*			
$C_7$				*					*				*	*				
$C_6$								*	*		*	*						
$C_5$		*			*					*	*							
$C_4$						*	*		*	*								
$C_3$			*				*	*	*									
$C_2$	*				*	*	*											
$C_1$	*	*	*	*														
	$q_{0001}$	$q_{0010}$	$q_{1100}$	$q_{1100}$	$q_{0100}$	$q_{1010}$	$q_{1010}$	$q_{1111}$	$q_{1111}$	$q_{0011}$	$q_{1111}$	$q_{0101}$	$q_{1001}$	$q_{1001}$	$q_{0110}$	$q_{1111}$	$q_{1111}$	$q_{1111}$

- We have to reject the assumption that the content-sharing variables are the same, because it is incompatible with any assignment of values that complies with the within-context distributions.

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- 3 we **derive from this** that all random variables in the system are jointly distributed;
- 4 we arrive at a contradiction;
- 5 we refute the assumption at step 3 and conclude: (at least some of the) random variables measuring the same thing in different contexts are not the same, even if they have the same distribution.



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- In practice, some of us prefer to “conclude” things more imaginative, such as  $JD(A, B)$  and  $JD(B, C)$  and  $JD(A, C)$  but not  $JD(A, B, C)$  — which is impossible in probability theory.

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- There is a much better way of conceptualizing the situation.



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*T*

*H*

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- Do not ask why and how the G's and F's create different variables: this is true by construction, there are no causes involved.
- There are, however, other questions one can meaningfully ask, the answers to which will cover every aspect of traditional contextuality analysis (and more).

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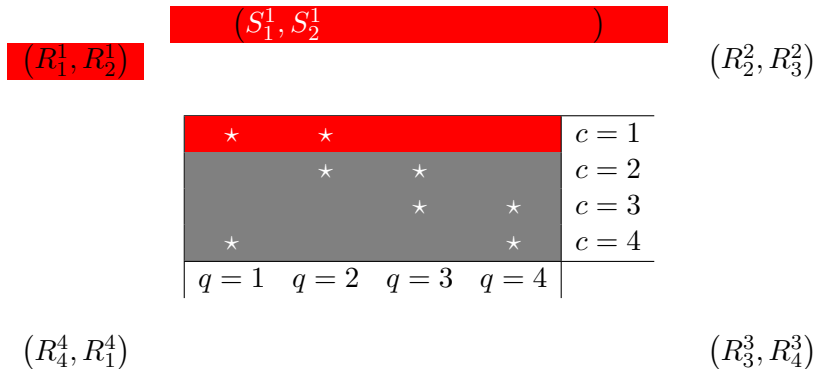
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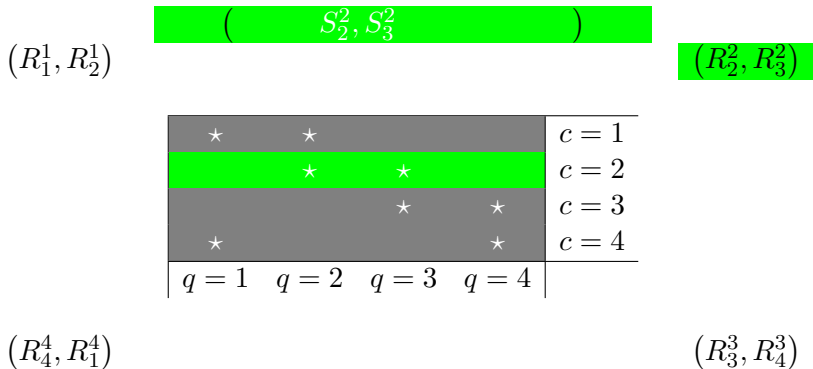
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Generally, there is an infinity of such couplings. In CbD, we are interested in the existence among them of certain, “special” couplings.

## Maximal couplings for content-sharing pairs

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	+	-
<i>A</i>	$p$	$1 - p$
<i>B</i>	$q$	$1 - q$

## Maximal couplings for content-sharing pairs

	+	-
$A$	$p$	$1 - p$
$B$	$q$	$1 - q$

	$B' = +$	$B' = -$	
$A' = +$	$\min(p, q)$	$p - \min(p, q)$	$p$
$A' = -$	$q - \min(p, q)$	$\min(1 - p, 1 - q)$	$1 - p$
	$q$	$1 - q$	

*equivalently,*

$$A' = B'$$

with max possible probability

## Contextuality-by-Default: Logic of contextuality analysis

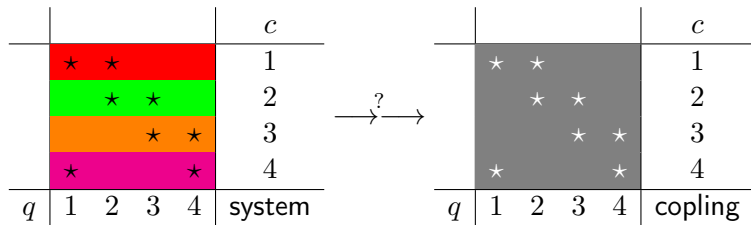
## Contextuality-by-Default: Logic of contextuality analysis

					$c$
	*	*			1
		*	*		2
			*	*	3
	*			*	4
$q$	1	2	3	4	system

- Given: A system of dichotomous random variables (each random variable is uniquely identified by its content and its context).

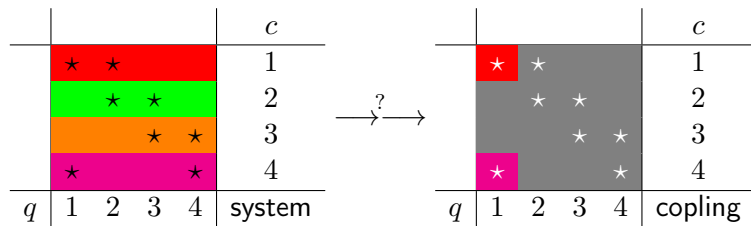


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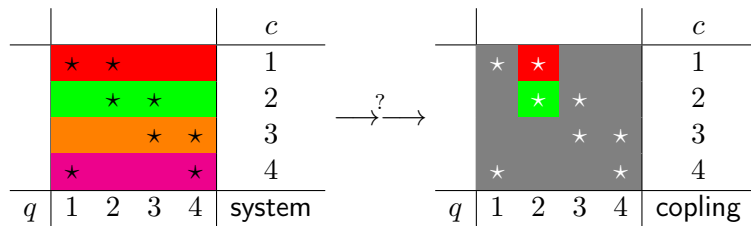
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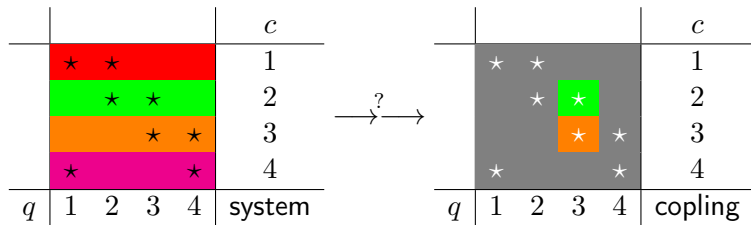
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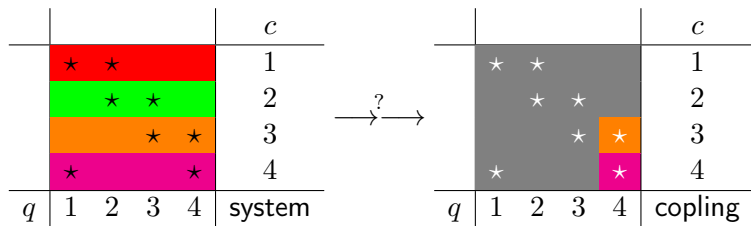
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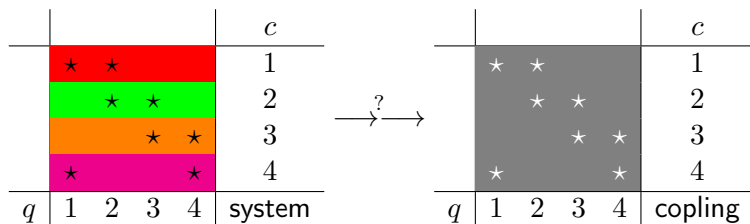
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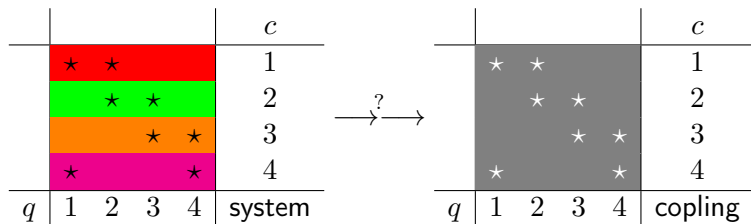
## Contextuality-by-Default: Logic of contextuality analysis



- Given: A system of dichotomous random variables (each random variable is uniquely identified by its content and its context).
- Question: Does it have a coupling which contains the maximal coupling of every pair of content-sharing random variables?
- If it does (does not), the system is noncontextual (resp., contextual).

Intuition behind the notion of a contextual system

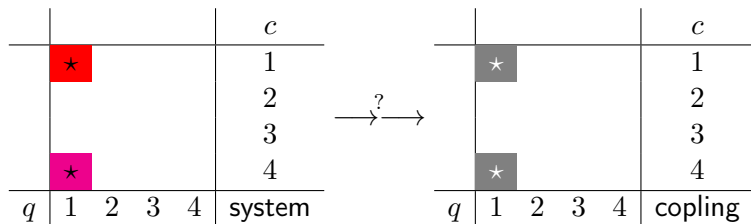
## Intuition behind the notion of a contextual system



- The maximal coupling of two content-sharing random variables shows how similar they would be in isolation from other random variables in the system.

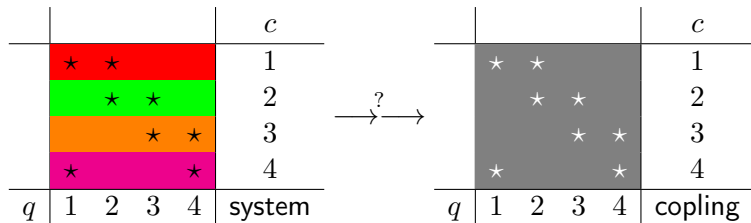


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- The maximal coupling of two content-sharing random variables shows how similar they would be in isolation from other random variables in the system.
- In contextual systems, the within-context joint distributions make the content-sharing random variables more dissimilar than they would be in isolation.

## Special case: Consistently connected systems and identity couplings

- In QM, traditional interest is in *consistently connected* systems.

	+	-
<i>A</i>	<i>p</i>	$1 - p$
<i>B</i>	<i>q</i>	$1 - p$

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	+	-
A	$p$	$1 - p$
B	$q$	$1 - p$

	$B' = +$	$B' = -$	
$A' = +$	$p$	0	$p$
$A' = -$	0	$1 - p$	$1 - p$
	$p$	$1 - p$	

equivalently,  $A' = B'$  with probability 1

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★	★			$c_{oo}$
	★	★		$c_{xo}$
		★	★	$c_{xx}$
★			★	$c_{ox}$
$q_{o\cdot}$	$q_{\cdot o}$	$q_{x\cdot}$	$q_{\cdot x}$	

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						$R_{o\cdot}^{o\cdot} = +1$	$r$	$p$
						$R_{o\cdot}^{o\cdot} = -1$	$q$	$1 - p - q - r$
	★	★		$c_{x\cdot}$		$R_{x\cdot}^{x\cdot} = +1$	$R_{x\cdot}^{x\cdot} = +1$	$R_{x\cdot}^{x\cdot} = -1$
						$R_{x\cdot}^{x\cdot} = +1$	0	0
						$R_{x\cdot}^{x\cdot} = -1$	$q'$	$1 - q'$
		★	★	$c_{xx}$		$R_{x\cdot}^{xx} = +1$	$R_{x\cdot}^{xx} = +1$	$R_{x\cdot}^{xx} = -1$
						$R_{x\cdot}^{xx} = +1$	0	0
						$R_{x\cdot}^{xx} = -1$	0	1
★			★	$c_{o\times}$		$R_{o\cdot}^{o\times} = +1$	$R_{o\cdot}^{o\times} = +1$	$R_{o\cdot}^{o\times} = -1$
						$R_{o\cdot}^{o\times} = +1$	0	$p'$
						$R_{o\cdot}^{o\times} = -1$	0	$1 - p'$
$q_{o\cdot}$	$q_{o\cdot}$	$q_{x\cdot}$	$q_{x\cdot}$					

# Why to deal with inconsistently connected systems?

			*	*	*	$C_{xxx}$
*			*	*		$C_{oxx}$
	*		*		*	$C_{xox}$
		*		*	*	$C_{xxo}$
*		*		*		$C_{oxo}$
*	*		*			$C_{oox}$
	*	*			*	$C_{xoo}$
*	*	*				$C_{ooo}$
$q_{o..}$	$q_{.o.}$	$q_{..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

Counterfactual values?

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*	*		*			$c_{oox}$
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*		*		*		$c_{oxo}$
*	*		*			$c_{oox}$
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*	*	*				$c_{ooo}$
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- Can we assign values to the same-content variables in other contexts?

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$q_{o..}$	$q_{o\cdot}$	$q_{\cdot o}$	$q_{\cdot\cdot x}$	$q_{\cdot x\cdot}$	$q_{x\cdot\cdot}$	

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*	*	*				$c_{ooo}$
$q_{o..}$	$q_{.o.}$	$q_{..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

- Equivalently: Can we assign values to all variables in all contexts?



## Counterfactual values?

			*	*	*	$c_{xxx}$
*			*	*		$c_{oxx}$
	*		*		*	$c_{xox}$
		*		*	*	$c_{xxo}$
*		*		*		$c_{oxo}$
*	*		*			$c_{oox}$
	*	*			*	$c_{xoo}$
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$q_{o..}$	$q_{.o.}$	$q_{..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

- Equivalently: Can we construct a coupling of the system?

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*	*		*			$c_{oox}$
	*	*			*	$c_{xoo}$
*	*	*				$c_{ooo}$
$q_{o..}$	$q_{.o.}$	$q_{..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

- Equivalently: Can we construct a coupling of the system? Yes, of course (in the absence of constraints).

## Counterfactual values?

			*	*	*	$c_{xxx}$
*			*	*		$c_{oxx}$
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*	*	*				$c_{ooo}$
$q_{o..}$	$q_{o.}$	$q_{..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

- A meaningful version: Can we  $\left[ \begin{array}{c} \text{assign values to} \\ \text{construct a coupling of} \end{array} \right]$  the system *subject to certain constraints* imposed on the same-content variables?

## Counterfactual values?

			★	★	★	$c_{xxx}$
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		★		★	★	$c_{x xo}$
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*		*		*		$C_{o xo}$
*	*		*			$C_{oox}$
	*	*			*	$C_{x oo}$
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$q_{o..}$	$q_{.o.}$	$q_{..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

- Can we assign values so that any two same-content variables always coincide? Only if the system is noncontextual and consistently connected (non-signaling).

# Distributional differences and the language of direct (causal) influences



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- That's why they can be considered manifestations of what we call *direct influences*.
- Direct influences are *information-transferring*: by observing that a distribution is different, one can guess that the variable is in another context.
- Difference in the identity of content-sharing random variables due to direct influences (= differences in their distributions) is measured by maximal couplings.

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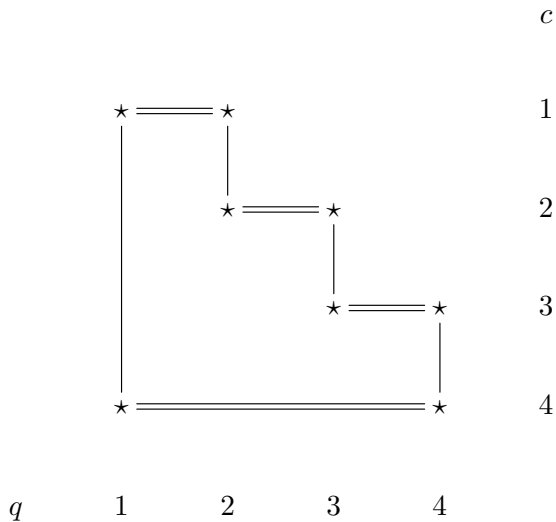


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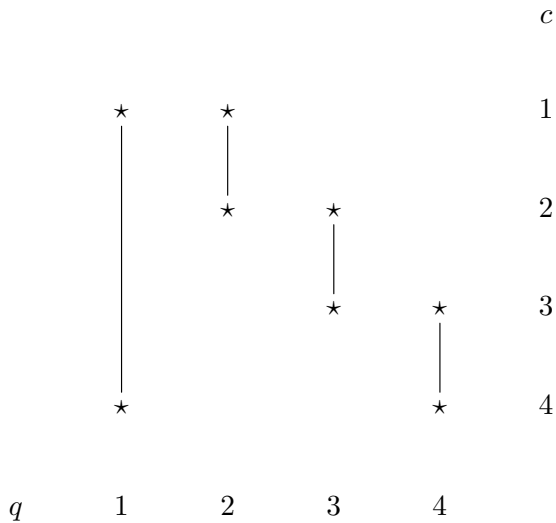
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- The identity of a random variable can be defined by the elements of contexts that occur simultaneously with its recording, or even in the future.
  - This may be the main reason why contextuality is not a physical concept.

## Contextuality-by-Default: Measures of contextuality

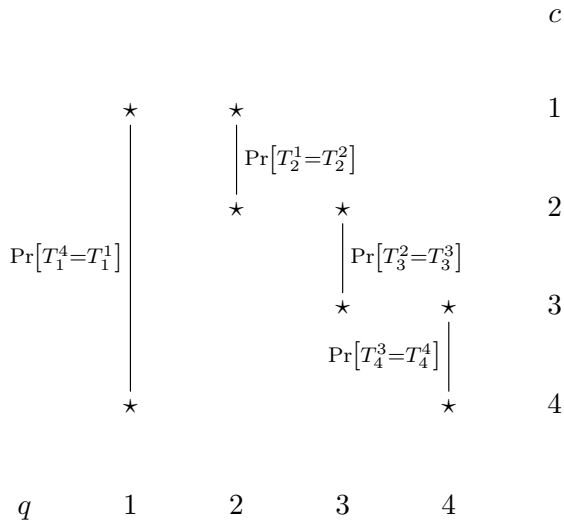
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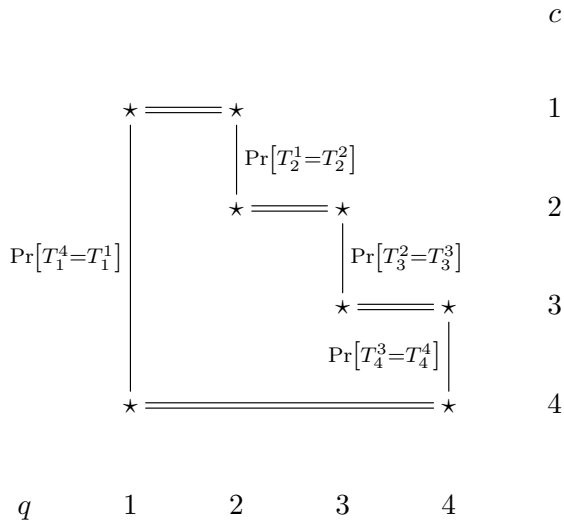
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$$\Delta_{\max} = \sum_{\text{couplings } (T_q^c, T_{q'}^{c'}) \text{ of } R_q^c, R_{q'}^{c'}} \max \Pr [T_q^c = T_{q'}^{c'}]$$

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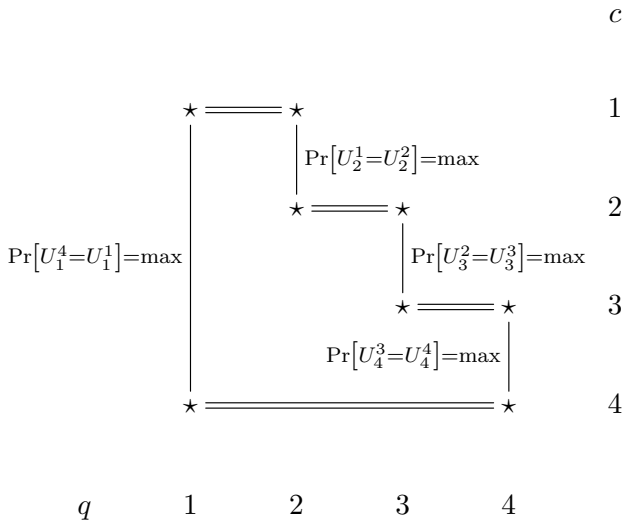
$$\text{CNTX}_1 = \Delta_{\max} - \Delta_{\text{achievable}}$$

Theorem (*Kujala-Dzhafarov, 2016*)

*In a cyclic system of dichotomous random variables,  $\text{CNTX}_1$  equals*

$$\frac{1}{2} \max_{(\iota_1, \dots, \iota_k) \in \{-1, 1\}^n: \prod_{i=1}^n \iota_i = -1} \sum_{i=1}^n \iota_i \langle R_i^i R_{i \oplus 1}^i \rangle - n + 2 - \sum_{i=1}^n |\langle R_i^i \rangle - \langle R_i^{i \oplus 1} \rangle|.$$

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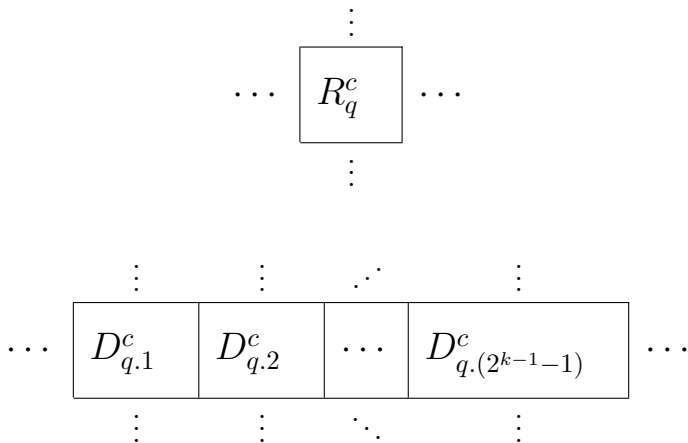
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- Relationship between  $CNTX_1$  and  $CNTX_2$  is not well-understood:
  - they are conjectured to be proportional for all cyclic systems:  
 $CNTX_2 = CNTX_1 / (\text{rank} - 1)$ .
    - (conjectured by Cervantes and Kujala based on numerical calculations)

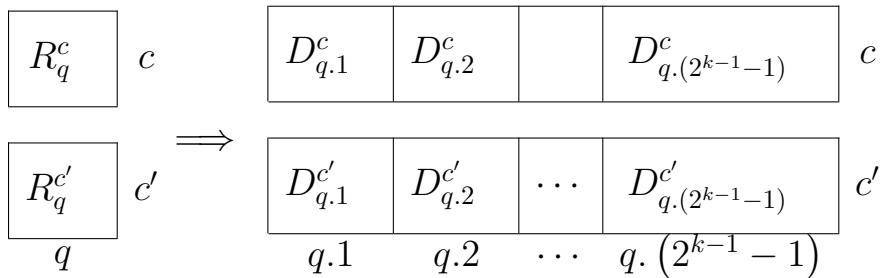
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## Contextuality-by-Default: Canonical systems

Theorem (*Dzhafarov-Cervantes-Kujala, 2017*)

*A canonical system is noncontextual only if for every pair of (original) content-sharing  $R_q^c$  and  $R_q^{c'}$  one of them nominally dominates the other.*

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Definition

$R_q^c$  nominally dominates  $R_q^{c'}$  if  $\Pr [R_q^c = i] < \Pr [R_q^{c'} = i]$  for no more than one value of  $i = 1, \dots, k$ .

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Example

$\updownarrow$	1	2	3	4	5		$\updownarrow$	1	2	3	4	5	
$R_q^c$	0.1	0.2	0.2	0.5	0	,	$R_q^c$	0.1	0.2	0.2	0.5	0	,
$R_q^{c'}$	0.1	0.2	0.2	0.5	0		$R_q^{c'}$	0.2	0.1	0.2	0.5	0	
$\downarrow$	1	2	3	4	5		$\times$	1	2	3	4	5	
$R_q^c$	0.1	0.2	0.2	0.5	0	,	$R_q^c$	0.1	0.2	0.2	0.5	0	,
$R_q^{c'}$	0.5	0	0.1	0.4	0		$R_q^{c'}$	0.3	0.3	0.1	0.2	0.1	

