

Contextuality Analysis

Ehtibar N. Dzhafarov

Acknowledging AFOSR grant FA9550-14-1-0318

QCQMB 2018, Prague, Czech Republic

Acknowledging collaboration with

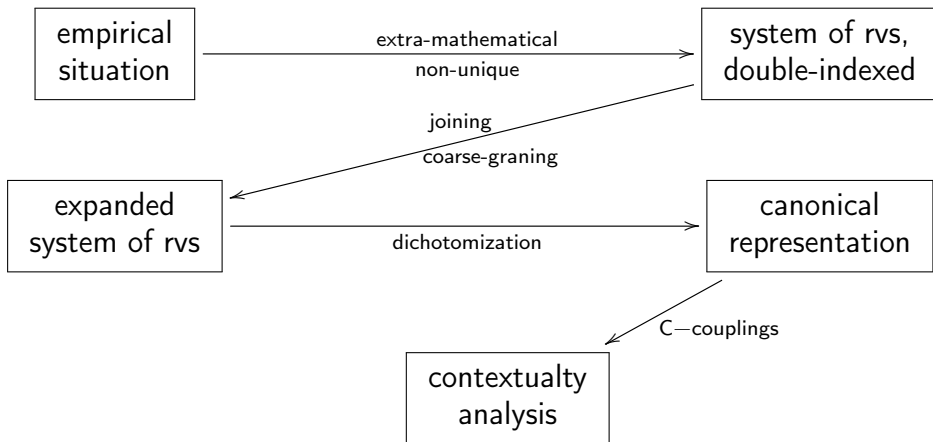
Janne V. Kujala,

Victor Cervantes, Maria Kon

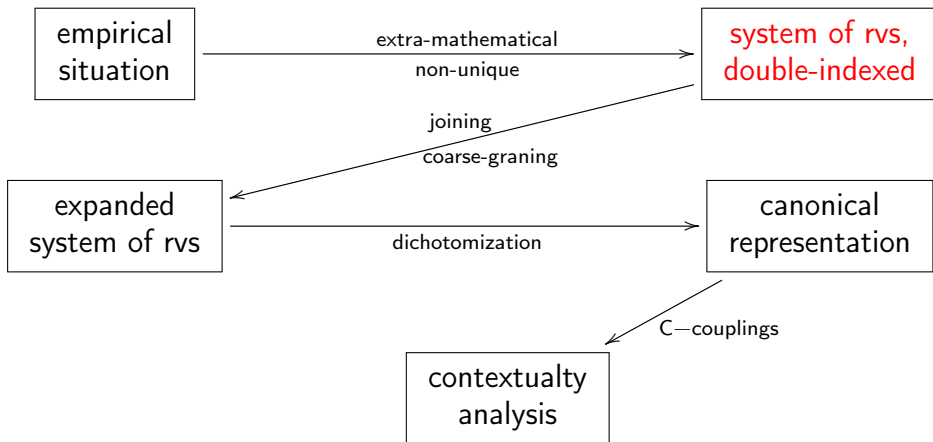
Benefited from recent discussions with

Matt Jones, Pawel Kurzynski, Andrei Khrennikov,
Federico Holic, Acacio de Barros

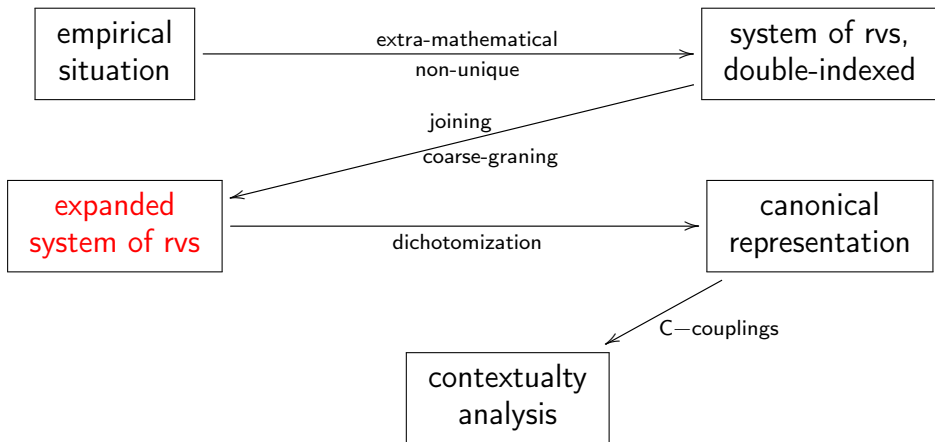
Preamble



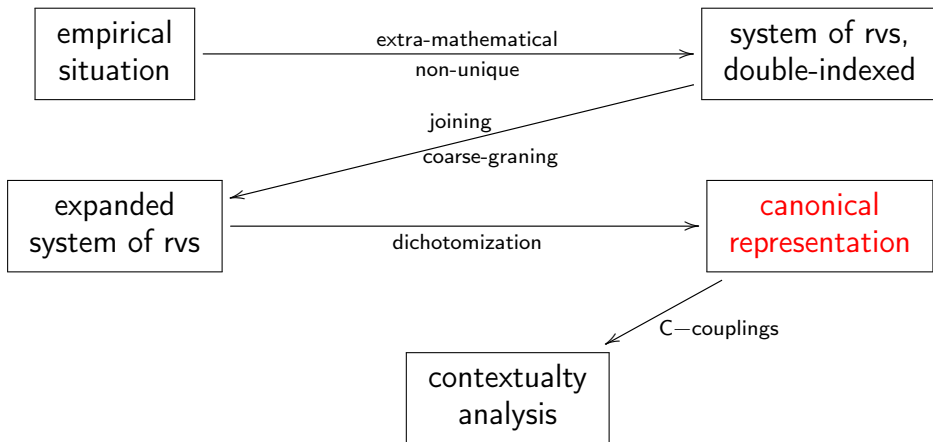
Preamble



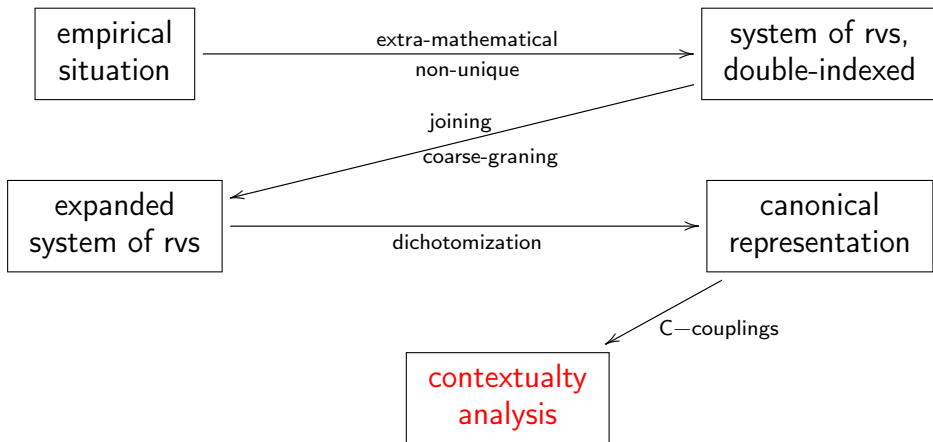
Preamble



Preamble



Preamble



Plan

Plan

- 1 Preliminaries: Couplings, C-couplings.

Plan

- 1 Preliminaries: Couplings, C-couplings.
- 2 Basics: Systems of random variables; Definition of C-(non)contextuality; Measure of C-contextuality.

Plan

- 1 Preliminaries: Couplings, C-couplings.
- 2 Basics: Systems of random variables; Definition of C-(non)contextuality; Measure of C-contextuality.
- 3 Double-Slit Paradigm.

Plan

- 1 Preliminaries: Couplings, C-couplings.
- 2 Basics: Systems of random variables; Definition of C-(non)contextuality; Measure of C-contextuality.
- 3 Double-Slit Paradigm.
- 4 More-than-binary random variables: Dichotomizations.

Plan

- 1 Preliminaries: Couplings, C-couplings.
- 2 Basics: Systems of random variables; Definition of C-(non)contextuality; Measure of C-contextuality.
- 3 Double-Slit Paradigm.
- 4 More-than-binary random variables: Dichotomizations.
- 5 Directions of development, summary.

Preliminaries: Couplings

$$X_1, X_2, \dots, X_n$$
$$(Y_1, Y_2, \dots, Y_n)$$

Preliminaries: Couplings

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

Preliminaries: Couplings

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

 \implies

	B' = +	B' = -	
A' = +	?	?	p
A' = -	?	?	1 - p
	q	1 - q	

Preliminaries: Couplings

IND

	$B' = +$	$B' = -$	
$A' = +$	pq		p
$A' = -$			$1 - p$
	q	$1 - q$	

Preliminaries: Couplings

IND

	$B' = +$	$B' = -$	
$A' = +$	pq	$p(1 - q)$	p
$A' = -$	$(1 - p)q$	$(1 - p)(1 - q)$	$1 - p$
	q	$1 - q$	

Preliminaries: Couplings

MAX

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

Preliminaries: Couplings

MAX

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

Preliminaries: Couplings

$$X_1, X_2, \dots, X_n$$

$$(Y_1, Y_2, \dots, Y_n)$$

Definition

A statement on (property of) a joint distribution of random variables is called a C-statement (C-property) if, for any n and any (allowable) X_1, X_2, \dots, X_n , a coupling (Y_1, Y_2, \dots, Y_n) that satisfies C exists and is unique. This coupling is called the C-coupling of X_1, X_2, \dots, X_n .

IND is a C-property
(for any class of random variables)

MultiMax is a C-property
for dichotomic X_1, X_2, \dots, X_n

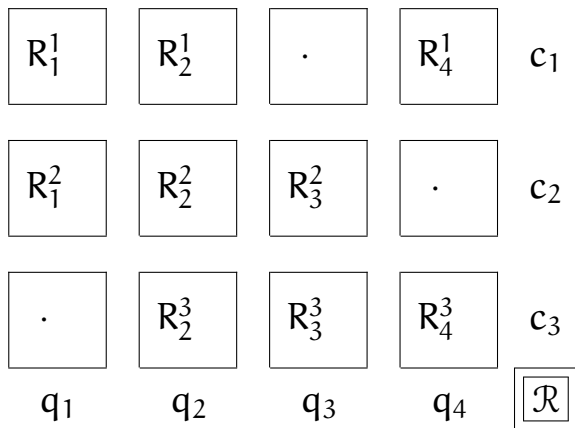
Preliminaries: Couplings

MultiMax is a C-property
for dichotomic X_1, X_2, \dots, X_n

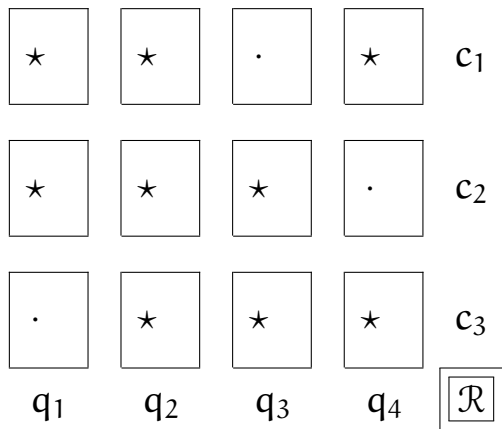
$$(Y_1, Y_2, \dots, Y_n)$$

$$C \equiv \forall i, j : (Y_i, Y_j) \text{ is MAX coupling of } X_i, X_j$$

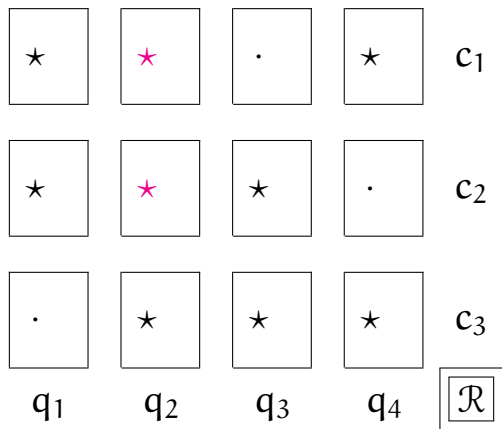
Contextuality-by-Default (CbD): Generalities



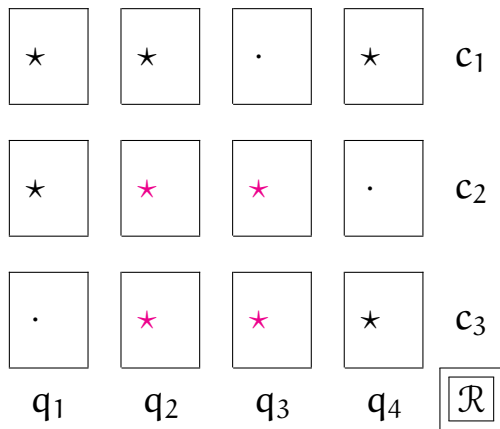
CbD: Generalities



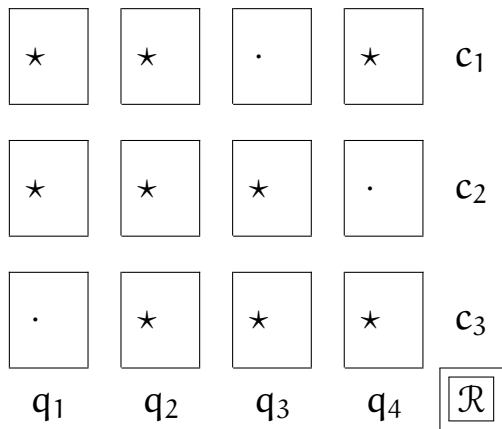
CbD: Generalities



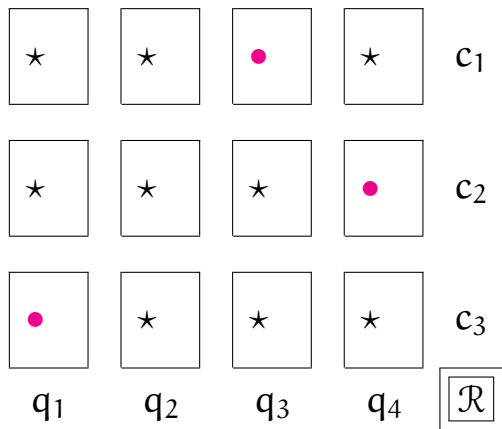
CbD: Generalities



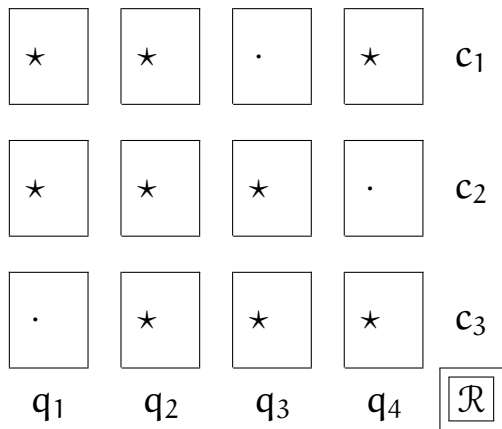
CbD: Generalities



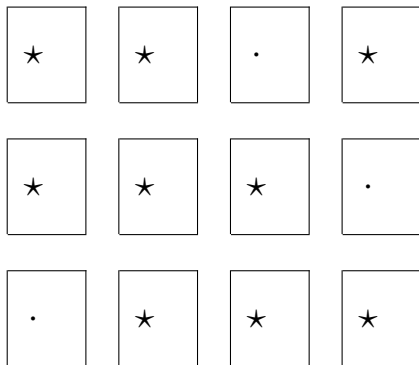
CbD: Generalities



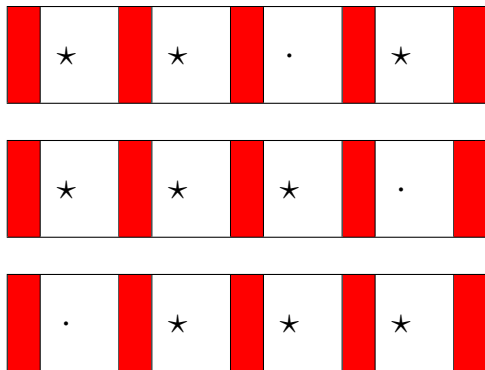
CbD: Generalities



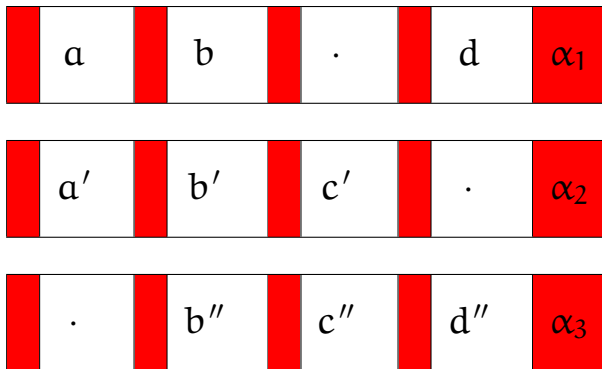
CbD: Generalities



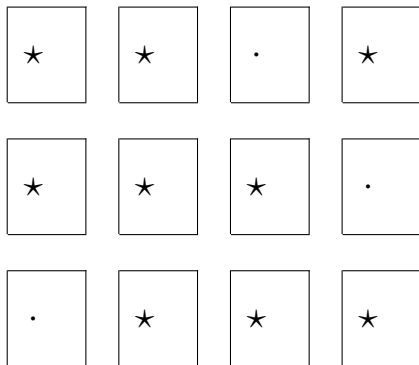
CbD: Generalities



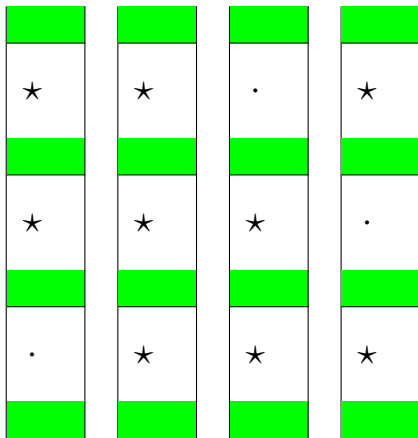
CbD: Generalities



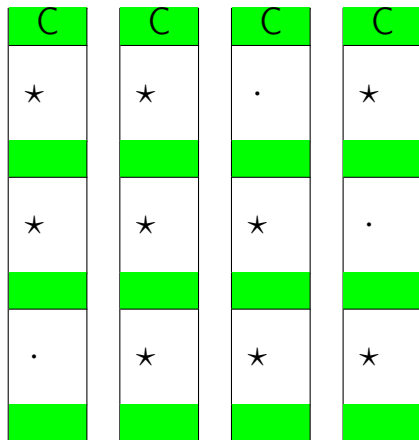
CbD: Generalities



CbD: Generalities



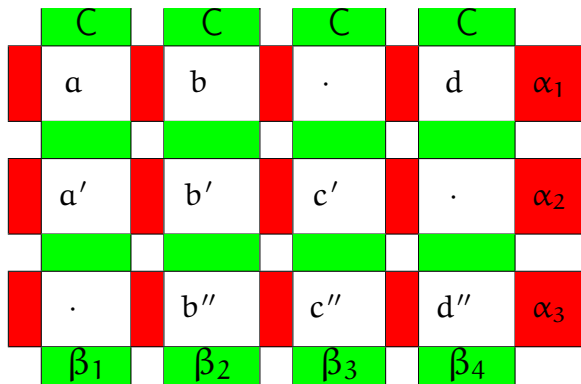
CbD: Generalities



CbD: Generalities

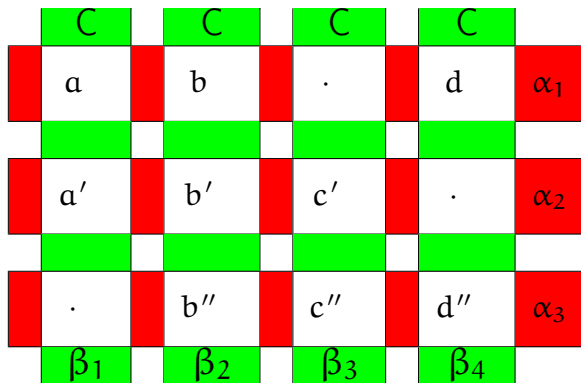
C	C	C	C
a	b	.	d
a'	b'	c'	.
.	b''	c''	d''
β_1	β_2	β_3	β_4

C-(non)contextuality: Definition



$\Pr [a, b, d, a', b', c', b'', c'', d'']$ exists?

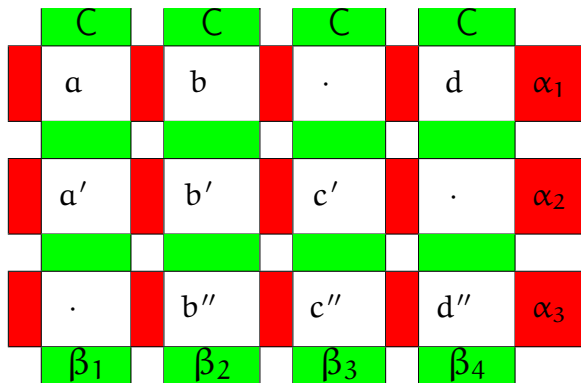
C-(non)contextuality: Definition



$\Pr [a, b, d, a', b', c', b'', c'', d'']$ exists?

Yes \rightarrow system is noncontextual (w.r.t.C)

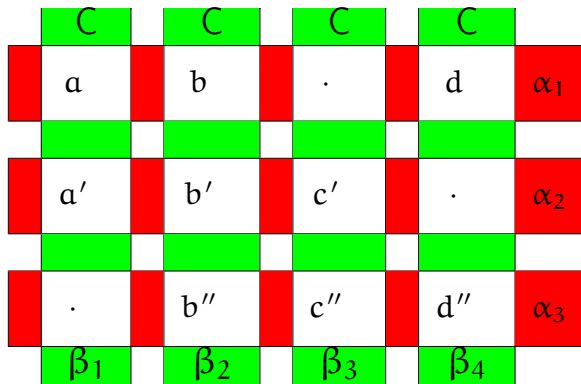
C-(non)contextuality: Definition



$\Pr [a, b, d, a', b', c', b'', c'', d'']$ exists?

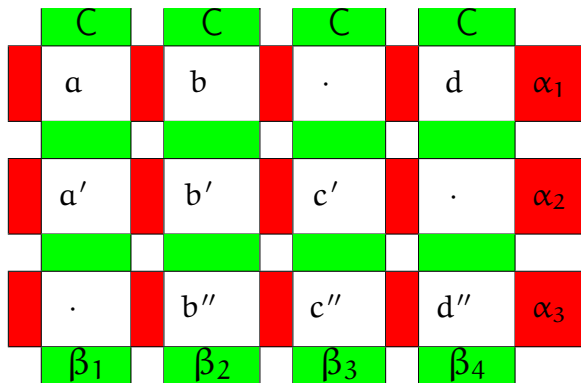
No \rightarrow system is contextual (w.r.t.C)

C-(non)contextuality: Definition



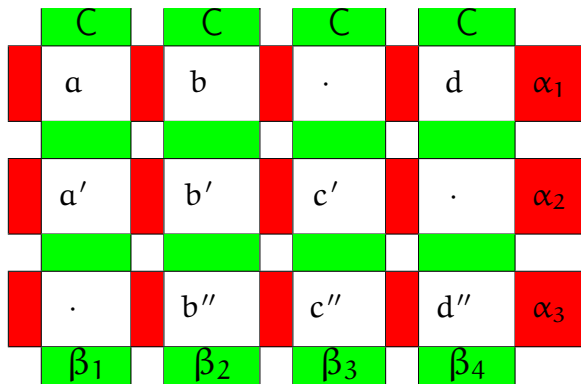
Any system is noncontextual w.r.t. $C \equiv \text{IND}$

C-(non)contextuality: Definition



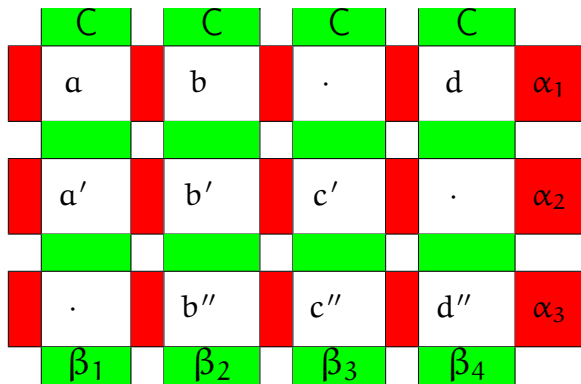
$\Pr [a, b, d, a', b', c', b'', c'', d'']$ for $C \equiv \text{MultiMax?}$

C-(non)contextuality: Definition



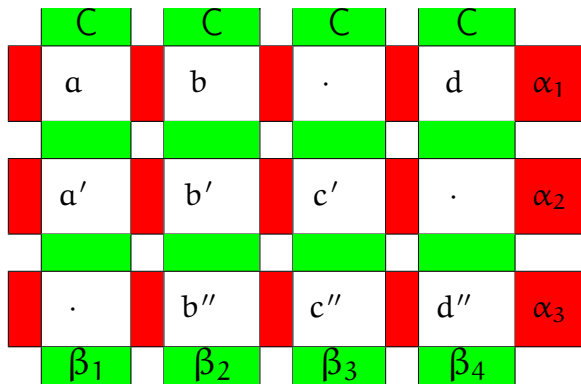
$\Pr [a, b, d, a', b', c', b'', c'', d'']$ for $C \equiv \text{MultiMax?}$
effectively determinable by LP

C-contextuality: Measure (through quasicouplings)



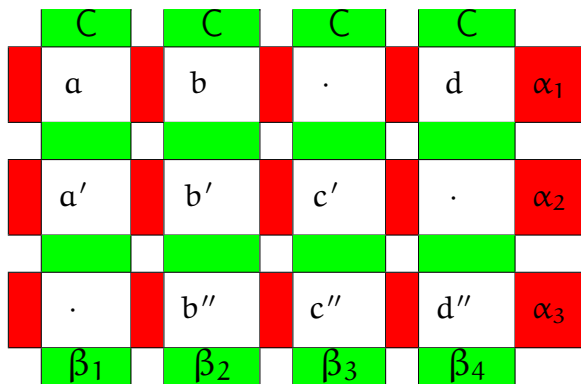
$\text{QPr} [a, b, d, a', b', c', b'', c'', d'']$

C-contextuality: Measure (through quasicouplings)



$\text{QPr} [a, b, d, a', b', c', b'', c'', d'']$ always exists

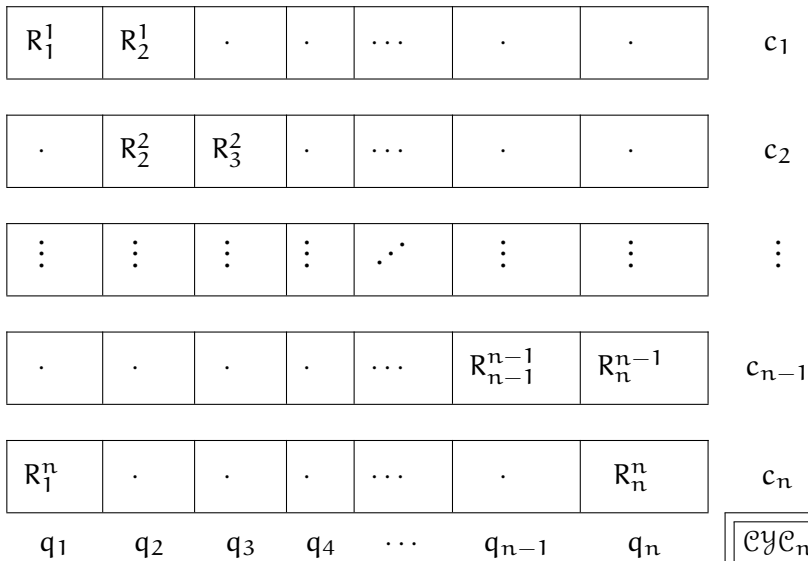
C-contextuality: Measure (through quasicouplings)



$$\min \sum |\text{QPr}[a, b, d, a', b', c', b'', c'', d'']| - 1$$

$= \min \text{TV} - 1$, effectively determinable by LP

Example: Cyclic systems of binary random variables



\mathcal{C}_n

Example: Cyclic systems of binary random variables

- $n > 5$ investigated in psychophysics (with Ru Zhang, Cervantes)
- $n = 5$ — KCBS-type system
- $n = 4$ — EPR/Bohm-Bell-type system (“Snow Queen” experiment, with Cervantes)
- $n = 3$ — Suppes-Zanotti (original Bell) or Leggett-Garg-type system (with Basieva, Cervantes, Khrennikov)
- $n = 2$ — question order (Moore-Wang-Busemeyer) type system

Example: Cyclic systems of binary random variables

Theorem

The degree of contextuality in a cyclic system of binary random variables is $1/2$ of

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

Example: Cyclic systems of binary random variables

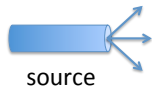
Theorem

A cyclic system of binary random variables is contextual iff

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

2-Slit, setup

Context $c_{o\times}$: only left slit open



q_o : open left slit

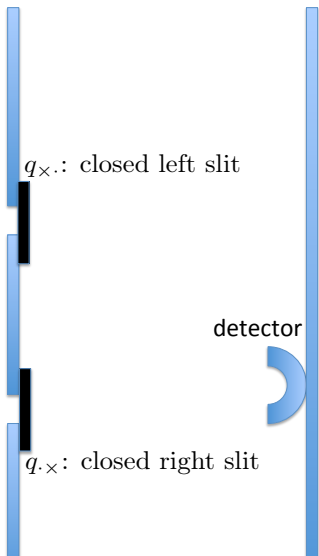
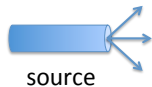
q_{\times} : closed right slit

detector



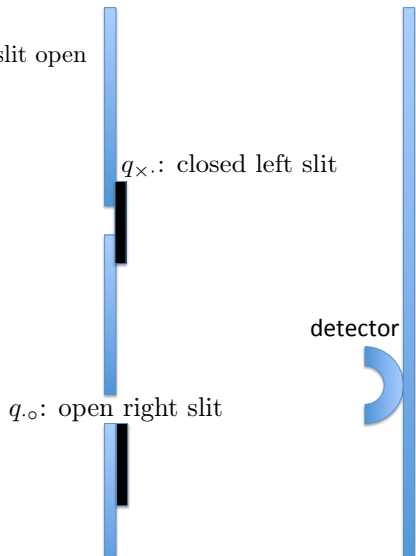
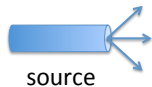
2-Slit, setup

Context $c_{\times\times}$: both slits closed



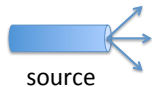
2-Slit, setup

Context $c_{\times o}$: only right slit open



2-Slit, setup

Context c_{oo} : both slits open



$q_{o.}$: open left slit

$q_{.o}$: open right slit

detector



2-Slit, random variables chosen

$$R_{\circ} = \begin{cases} +1 & \text{if } \begin{array}{l} \text{passes through } q_{\circ} \\ \text{and} \\ \text{hits detector} \end{array} \\ -1 & \text{otherwise} \end{cases}$$

2-Slit, random variables chosen

$$R_{\circ\circ} = \begin{cases} +1 & \text{if } \begin{array}{l} \text{passes through } q_{\circ} \\ \text{and} \\ \text{hits detector} \end{array} \\ -1 & \text{otherwise} \end{cases}$$

2-Slit, random variables chosen

$$R_{o.}^{\circ\times} = \begin{cases} +1 & \text{if } \begin{array}{l} \text{passes through } q_{o.} \\ \text{and} \\ \text{hits detector} \end{array} \\ -1 & \text{otherwise} \end{cases}$$

2-Slit, random variables chosen

$$R_{x.}^{x_0} = \begin{cases} +1 & \text{if } \begin{array}{l} \text{passes through } q_x \\ \text{and} \\ \text{hits detector} \end{array} \\ -1 & \text{otherwise} \end{cases}$$

2-Slit, random variables chosen

$$R_{x.}^{x.} = \begin{cases} +1 & \text{if } \begin{array}{l} \text{passes through } q.x \\ \text{and} \\ \text{hits detector} \end{array} \\ -1 & \text{otherwise} \end{cases}$$

2-Slit, observed probabilities and scenarios

context c_{oo}	$R_{\circ}^{\circ\circ} = +1$	$R_{\circ}^{\circ\circ} = -1$	
$R_{\circ}^{\circ\circ} = +1$	r	p	$r + p$
$R_{\circ}^{\circ\circ} = -1$	q	$1 - p - q - r$	$1 - r - p$
	$r + q$	$1 - r - q$	

2-Slit, observed probabilities and scenarios

context c_{x_o}	$R_{\circ}^{x_o} = +1$	$R_{\circ}^{x_o} = -1$	
$R_{x\cdot}^{x_o} = +1$	0	0	0
$R_{x\cdot}^{x_o} = -1$	q'	$1 - q'$	1
	q'	$1 - q'$	

2-Slit, observed probabilities and scenarios

context $c_{o \times}$	$R_{\cdot \times}^{o \times} = +1$	$R_{\cdot \times}^{o \times} = -1$	
$R_{o \cdot}^{o \times} = +1$	0	p'	p'
$R_{o \cdot}^{o \times} = -1$	0	$1 - p'$	$1 - p'$
	0	1	

2-Slit, observed probabilities and scenarios

context c_{xx}	$R_{.x}^{xx} = +1$	$R_{.x}^{xx} = -1$	
$R_{x.}^{xx} = +1$	0	0	0
$R_{x.}^{xx} = -1$	0	1	1
	0	1	

2-Slit, content-context matrix

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

2-Slit, content-context matrix

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

2-Slit, content-context matrix

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

2-Slit, content-context matrix

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

2-Slit, content-context matrix

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

2-Slit, content-context matrix

★	★	c_{oo}
$q_{o\cdot}$	$q_{\cdot o}$	

3-Slit, content-context matrix

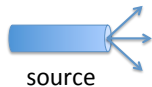
			★	★	★	C_{xxx}
★			★	★		C_{oxx}
	★		★		★	C_{xox}
		★		★	★	C_{xxo}
★		★		★		C_{oxo}
★	★		★			C_{oox}
	★	★			★	C_{xoo}
★	★	★				C_{ooo}
$q_{o..}$	$q_{o.o}$	$q_{o..o}$	$q_{..x}$	$q_{.x.}$	$q_{x..}$	

3-Slit, content-context matrix

★		★	$\mathbf{C}_{\circ\times\circ}$
★	★		$\mathbf{C}_{\circ\circ\times}$
	★	★	$\mathbf{C}_{\times\circ\circ}$
★	★	★	$\mathbf{C}_{\circ\circ\circ}$
$\mathbf{q}_{\circ..}$	$\mathbf{q}_{\cdot\circ\cdot}$	$\mathbf{q}_{\dots\circ}$	

2-Slit, many detectors

Context c_{oo} : both slits open



q_o : open left slit

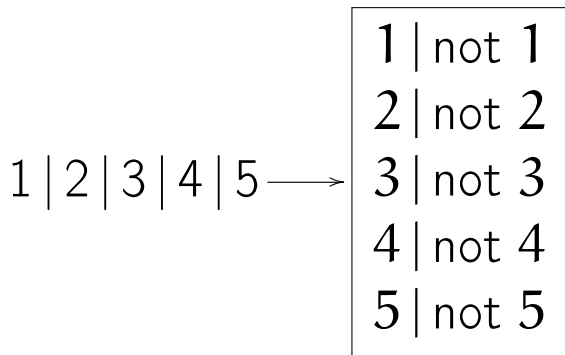
q_o : open right slit

detectors



5

Canonical Representation



Canonical Representation

1 | 2 | 3 | 4 | 5 →

12 | not 12

13 | not 13

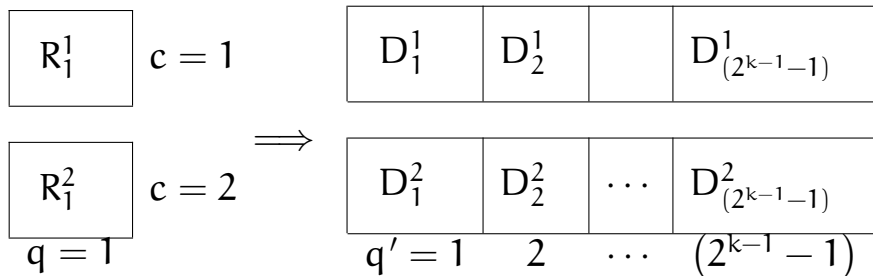
...

34 | not 34

35 | not 35

45 | not 45

Canonical Representation



Canonical Representation

Theorem

The system (consisting of R_1^1 and R_1^2) is noncontextual if and only if one of the R_1^1 and R_1^2 nominally dominates the other.

Canonical Representation

Theorem

The system (consisting of R_1^1 and R_1^2) is noncontextual if and only if one of the R_1^1 and R_1^2 nominally dominates the other.

Definition

R_1^1 nominally dominates R_1^2 if $\Pr [R_1^1 = i] < \Pr [R_1^2 = i]$ for no more than one value of $i = 1, \dots, k$.

Canonical Representation

Theorem

The system (consisting of R_1^1 and R_1^2) is noncontextual if and only if one of the R_1^1 and R_1^2 nominally dominates the other.

Example

$x:$	1	2	3	4	5
$\Pr [R_1^1 = x]$.2	.3	.1	.3	.1
$\Pr [R_1^2 = x]$.1	.2	.4	.3	0

Canonical Representation

Theorem

The system (consisting of R_1^1 and R_1^2) is noncontextual if and only if one of the R_1^1 and R_1^2 nominally dominates the other.

Example

$x:$	1	2	3	4	5
$\Pr [R_1^1 = x]$.2	.3	.1	.3	.1
$\Pr [R_1^2 = x]$.1	.2	.4	.3	0

Directions of development

- 1 Applications for C-couplings other than multimaximal ones.

Directions of development

- 1 Applications for C-couplings other than multimaximal ones.
- 2 Theory with incomplete sets of dichotomizations (e.g., for continuous random variables on \mathbb{R}).

Directions of development

- 1 Applications for C-couplings other than multimaximal ones.
- 2 Theory with incomplete sets of dichotomizations (e.g., for continuous random variables on \mathbb{R}).
- 3 Contextually labeled observables in Hilbert spaces.

Summary

- 1 Contextuality analysis applies to systems of content-context-indexed random variables, after they have been transformed into a canonical (split representation) form, in which all random variables are dichotomous. Each of these dichotomous random variables answers the question: does the outcome of a given initial random variable fall within a given subset of its possible values?

Summary

- 1 Contextuality analysis applies to systems of content-context-indexed random variables, after they have been transformed into a canonical (split representation) form, in which all random variables are dichotomous. Each of these dichotomous random variables answers the question: does the outcome of a given initial random variable fall within a given subset of its possible values?
- 2 A system of random variables is considered C-noncontextual iff the canonical representation thereof has a probabilistic coupling in which any set of content-sharing dichotomous random variables satisfies C.

Summary

- 1 Contextuality analysis applies to systems of content-context-indexed random variables, after they have been transformed into a canonical (split representation) form, in which all random variables are dichotomous. Each of these dichotomous random variables answers the question: does the outcome of a given initial random variable fall within a given subset of its possible values?
- 2 A system of random variables is considered C -noncontextual iff the canonical representation thereof has a probabilistic coupling in which any set of content-sharing dichotomous random variables satisfies C .
- 3 The minimal total variation (less one) across all possible quasi-couplings of the canonical representation of the system is a measure of contextuality of the system.

Summary

CbD approach to contextuality satisfies the following:

Summary

CbD approach to contextuality satisfies the following:

- 1 any subsystem of a noncontextual system of random variables is noncontextual;

Summary

CbD approach to contextuality satisfies the following:

- 1 any subsystem of a noncontextual system of random variables is noncontextual;
- 2 componentwise transformations of a noncontextual system of random variables are noncontextual;

Summary

CbD approach to contextuality satisfies the following:

- 1 any subsystem of a noncontextual system of random variables is noncontextual;
- 2 componentwise transformations of a noncontextual system of random variables are noncontextual;
- 3 a consistently connected (“non-signaling”) system is (non)contextual iff it is (non)contextual in the traditional sense.

