#### Context-Dependence and Contextuality

Ehtibar N. Dzhafarov

#### Supported by AFOSR grant FA9550-14-1-0318

QCQMB 2017, Prague, Czech Republic

#### Acknowledging collaboration with

Janne V. Kujala, Matt Jones,

Victor Cervantes, Ru Zhang,

Acacio de Barros, Gary Oas

### Benefited from discussions with

#### Samson Abramsky, Rui Soares Barbosa,

Pawel Kurzynski, Sam Staton,

Andrei Khrennikov, Federico Holic

## Contextuality: My approach



## Contextuality: Other Approaches



• Contextual labeling (identity). Fundamental and universal.

- **O** Contextual labeling (identity). Fundamental and universal.
- Oirect (cross-) influences (inconsistent connectedness). Can be present or absent.

- **O** Contextual labeling (identity). Fundamental and universal.
- Oirect (cross-) influences (inconsistent connectedness). Can be present or absent.
- Sontextuality proper. Can be present or absent.

#### Contextuality-by-Default Theory: Contextual Labeling

1. Random variables are contextually labeled (i.e., their context is part of their identity).



### Contextuality-by-Default Theory: Contextual Labeling

1. Random variables are contextually labeled (i.e., their context is part of their identity).



This implies that the sets of random variables in different contexts are disjoint.

2. The distributions of two connected random variables (measuring the same content in different contexts) may be different.

2. The distributions of two connected random variables (measuring the same content in different contexts) may be different.



2. The distributions of two connected random variables (measuring the same content in different contexts) may be different.



2. The distributions of two connected random variables (measuring the same content in different contexts) may be different.



The interpretation is: "Direct" influences (cross-influences), Signaling, Disturbance, Context-dependent biases, etc.

3. Contextuality is present if the joint distributions within contexts are incompatible with certain joint distributions imposed on the content-sharing random variables.

3. Contextuality is present if the joint distributions within contexts are incompatible with certain joint distributions imposed on the content-sharing random variables.



3. Contextuality is incompatibility of joint distributions of the bunched random variables with certain joint distributions imposed on the content-sharing random variables.

R <sub>1</sub> <sup>1</sup>	R <sub>2</sub> <sup>1</sup>	•	$R_4^1$	c <sub>1</sub>
R <sub>1</sub> <sup>2</sup>	R <sub>2</sub> <sup>2</sup>	R <sub>3</sub> <sup>2</sup>	•	c <sub>2</sub>
•	R <sub>2</sub> <sup>3</sup>	R <sub>3</sub> <sup>3</sup>	R <sub>4</sub> <sup>3</sup>	c <sub>3</sub>
q <sub>1</sub>	q <sub>2</sub>	<b>q</b> <sub>3</sub>	<b>q</b> <sub>4</sub>	R















3. Contextuality is incompatibility of joint distributions of the bunched random variables with certain couplings imposed on the connected random variables.



There is a universal measure of contextuality involving signed measures.





Couplings are imposed so that any two connected random variables are equal to each other with maximal possible probability. This is called a multi-maximal coupling. (*Can be generalized.*)

# Necessity of Contextual Labeling: Traditional View is Contradictory

Noncontextual labeling is not an option, even in the absence of direct influences:



(Non)Contextuality may exist with or without direct influences.

- (Non)Contextuality may exist with or without direct influences.
- **2** Direct influences are causal, contextuality is correlational:

- (Non)Contextuality may exist with or without direct influences.
- 2 Direct influences are causal, contextuality is correlational:
  - direct influences are manifested in changes of marginal distributions, contextuality is revealed on the level of joint distributions;

- (Non)Contextuality may exist with or without direct influences.
- 2 Direct influences are causal, contextuality is correlational:
  - direct influences are manifested in changes of marginal distributions, contextuality is revealed on the level of joint distributions;
  - contextuality can relate spacelike-separated measurements;

- (Non)Contextuality may exist with or without direct influences.
- ② Direct influences are causal, contextuality is correlational:
  - direct influences are manifested in changes of marginal distributions, contextuality is revealed on the level of joint distributions;
  - contextuality can relate spacelike-separated measurements;
  - with timelike separation, a future measurement can create an effective context for a past one.



Couplings are imposed so that any two connected random variables are equal to each other with maximal possible probability. This is called a multi-maximal coupling. (*Can be generalized.*)





In a well-designed theory a multi-maximal coupling always exists and is unique.



In a well-designed theory a multi-maximal coupling always exists and is unique.

• This is true if all random variables are binary (dichotomous).



In a well-designed theory a multi-maximal coupling always exists and is unique.

- This is true if all random variables are binary (dichotomous).
- 2 This is not generally true.

## Example: Cyclic systems of binary random variables

R <sub>1</sub> <sup>1</sup>	R <sub>2</sub> <sup>1</sup>	•	•	•••	•	•	c1
-----------------------------	-----------------------------	---	---	-----	---	---	----

•	R <sub>2</sub> <sup>2</sup>	R <sub>3</sub> <sup>2</sup>	•		•		c <sub>2</sub>
---	-----------------------------	-----------------------------	---	--	---	--	----------------

:	:	:	:		:	:	:
•	•	•	•	•	•	•	•

•	•	•	•		$R_{n-1}^{n-1}$	$R_n^{n-1}$	c <sub>n-1</sub>
---	---	---	---	--	-----------------	-------------	------------------

$R_1^n$	•	•	•	 •	R <sub>n</sub> <sup>n</sup>	c <sub>n</sub>
qı	q <sub>2</sub>	q <sub>3</sub>	$q_4$	 $q_{n-1}$	qn	CyCn

### Example: Cyclic systems of binary random variables



#### Example: Cyclic systems of binary random variables

#### Theorem

A cyclic system of binary random variables is contextual if and only if

$$\max_{(\iota_1,\ldots,\iota_k)\in\{-1,1\}^n:\prod_{i=1}^n\iota_i=-1}\sum_{i=1}^n\iota_i\left\langle R^i_iR^i_{i\oplus1}\right\rangle>n-2+\sum_{i=1}^n\left|\left\langle R^i_i\right\rangle-\left\langle R^{i\ominus1}_i\right\rangle\right|.$$







Ontic-epistemic interplay: the representation is non-unique (in many ways).





Joining:





Joining:



Ontic-epistemic interplay: joining is non-unique and selective. There are cases (e.g., question-order effect) when it is unwarranted.





#### Coarsening (coarse-graining):





#### Coarsening (coarse-graining):



Ontic-epistemic interplay: coarsening is non-unique and selective. Not all possible coarsenings may be of interest (depends on the internal structure, e.g., linear ordering or metric, of the space of possible values).



Coarsening (coarse-graining):



Ontic-epistemic interplay: coarsening is non-unique and selective. Not all possible coarsenings may be of interest (depends on the internal structure, e.g., linear ordering or metric, of the space of possible values).





Each random variable in the expanded system is replaced with its split representation, a binary random variable isolating one of the values from the rest (detecting this value).



Each random variable in the expanded system is replaced with its split representation, a binary random variable isolating one of the values from the rest (detecting this value).



The support of the system does not change.

#### Theorem

The system  $\mathfrak{D}$  is noncontextual if and only if one of the  $R_1^1$  and  $R_1^2$  nominally dominates the other.

#### Theorem

The system  ${\mathbb D}$  is noncontextual if and only if one of the  $R^1_1$  and  $R^2_1$  nominally dominates the other.

#### Definition

 $R_1^1$  nominally dominates  $R_1^2$  if  $\mathsf{Pr}\left[R_1^1=i\right]<\mathsf{Pr}\left[R_1^2=i\right]$  for no more than one value of  $i=1,\ldots,k.$ 

#### Theorem

The system  ${\mathbb D}$  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
$\Pr\left[R_{1}^{1}=x\right]$	.2	.3	.1	.3	.1
$\Pr\left[R_1^2 = x\right]$	.1	.2	.4	.3	0

#### Theorem

The system  ${\mathbb D}$  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
$\Pr\left[R_1^1 = x\right]$	.2	.3	.1	.3	.1
$\Pr\left[R_1^2 = x\right]$	.1	.2	.4'	.3	0

### Measure of contextuality: Quasi-couplings



TOTAL VARIATION:
$$\sum | p(s_1^1, s_2^1, s_4^1, s_1^2, s_2^2, s_2^2, s_3^2, s_3^3, s_4^3) | = V_T \ge 1$$
 $V_T = 1$  if and only if the quasi-coupling is a proper coupling

### Measure of contextuality

#### Theorem

Any system  $\Re$  of random variables has a quasi-coupling S such that (i) any pair  $(S_q^c, S_q^{c'})$  is properly distributed, and the probability of  $S_q^c = S_q^{c'}$  has the maximal possible value;

(ii) the value of total variation  $V_T$  in S has the smallest possible value among all quasi-couplings satisfying (i).

### Measure of contextuality

#### Theorem

Any system  $\Re$  of random variables has a quasi-coupling S such that (i) any pair  $(S_q^c, S_q^{c'})$  is properly distributed, and the probability of  $S_q^c = S_q^{c'}$  has the maximal possible value;

(ii) the value of total variation  $V_T$  in S has the smallest possible value among all quasi-couplings satisfying (i).

The minimum value of  $V_{\rm T}-1$  can be taken as a universal measure of contextuality in any system of random variables.

Its value is 0 if and only if the system is not contextual.

### Dummy measurements



#### Dummy measurements



#### Theorem

Contextuality of a system does not change if every context-content pair associated with no measurement is assigned a deterministic random variable ("dummy measurement").



