

Title: Tutorial: Causal Inference Meets Quantum Physics

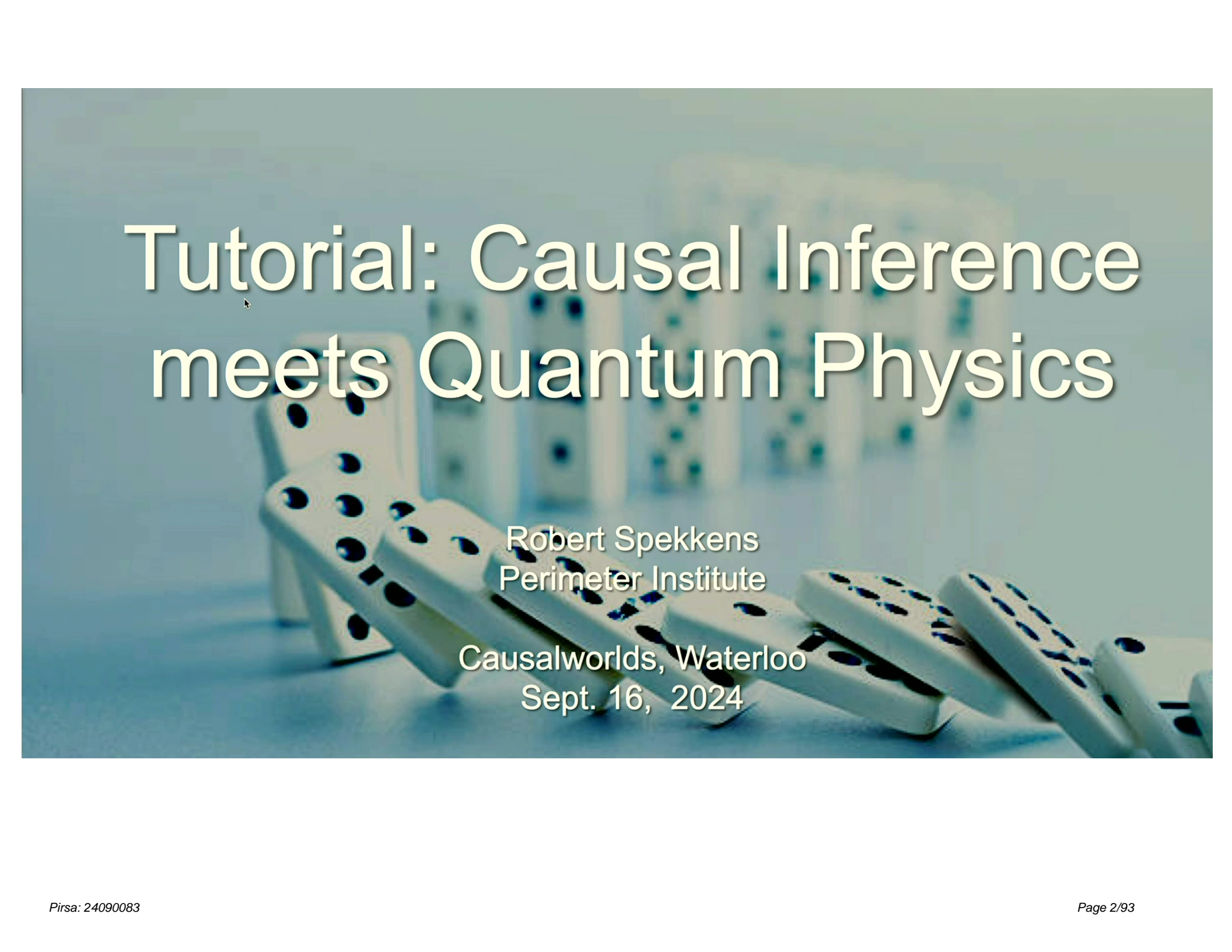
Speakers: Robert Spekkens

Series: Quantum Foundations, Quantum Information

Date: September 16, 2024 - 9:20 AM

URL: <https://pirsa.org/24090083>

Abstract: Can the effectiveness of a medical treatment be determined without the expense of a randomized controlled trial? Can the impact of a new policy be disentangled from other factors that happen to vary at the same time? Questions such as these are the purview of the field of causal inference, a general-purpose science of cause and effect, applicable in domains ranging from epidemiology to economics. Researchers in this field seek in particular to find techniques for extracting causal conclusions from statistical data. Meanwhile, one of the most significant results in the foundations of quantum theory—Bell’s theorem—can also be understood as an attempt to disentangle correlation and causation. Recently, it has been recognized that Bell’s result is an early foray into the field of causal inference and that the insights derived from almost 60 years of research on his theorem can supplement and improve upon state-of-the-art causal inference techniques. In the other direction, the conceptual framework developed by causal inference researchers provides a fruitful new perspective on what could possibly count as a satisfactory causal explanation of the quantum correlations observed in Bell experiments. Efforts to elaborate upon these connections have led to an exciting flow of techniques and insights across the disciplinary divide. This tutorial will highlight some of what is happening at the intersection of these two fields.



Tutorial: Causal Inference meets Quantum Physics

Robert Spekkens
Perimeter Institute

Causalworlds, Waterloo
Sept. 16, 2024

15 hours of lectures
Available online at <https://pirsa.org/c23016>

Causal
Inference:
Classical and
Quantum

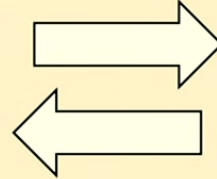
PHYS 777-007
Lecturer: Robert Spekkens
TA: Marina Ansanelli

March 6, 2023



Causarum Investigatio
"Investigate the causes"

The field of
Causal
Inference



The field of
Quantum
Foundations

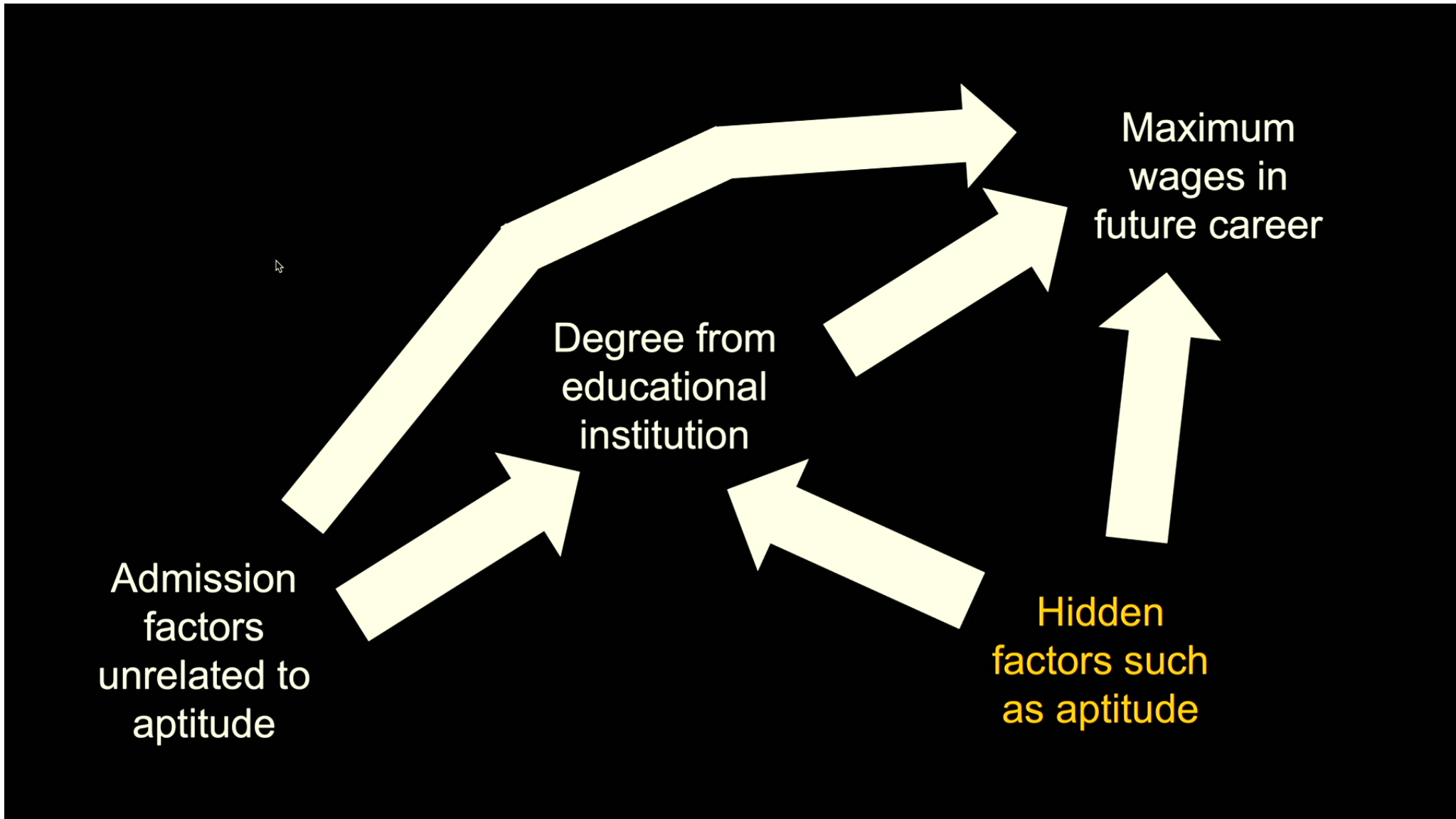
Instrumental inequalities

Causal Inference in the presence of hidden variables

Maximum wages
in future career
above some
threshold?

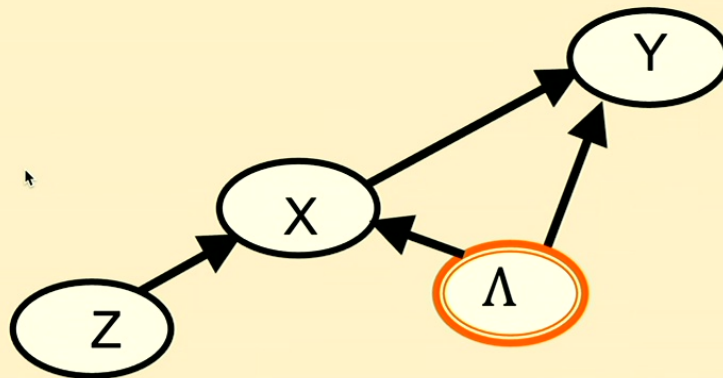
Degree from
educational
institution?

	Yes	No
Yes	79%	21%
No	43%	57%



Causal structure

Parameters



$$P_{X|\Lambda Z}$$

$$P_{Y|\Lambda X}$$

$$P_{\Lambda}$$

$$P_{XY|Z} = \sum_{\Lambda} P_{Y|X\Lambda} P_{X|Z\Lambda} P_{\Lambda}$$

Example of causal compatibility constraint:

$$P_{XY|Z}(00|0) + P_{XY|Z}(01|1) \leq 1$$

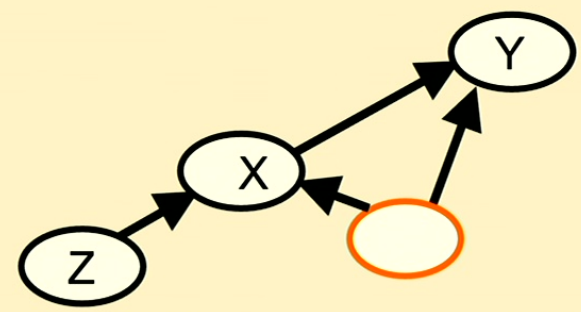
Pearl, 1993

The evidence

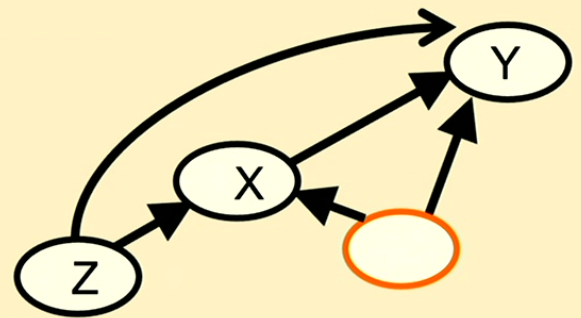
		Z=0	
		Y=0	Y=1
Z=0	X=0	0.79	0.21
	X=1	0.43	0.57
		Z=1	
		Y=0	Y=1
Z=1	X=0	0.59	0.41
	X=1	0.39	0.61

Violates Pearl's instrumental inequality!

The hypotheses



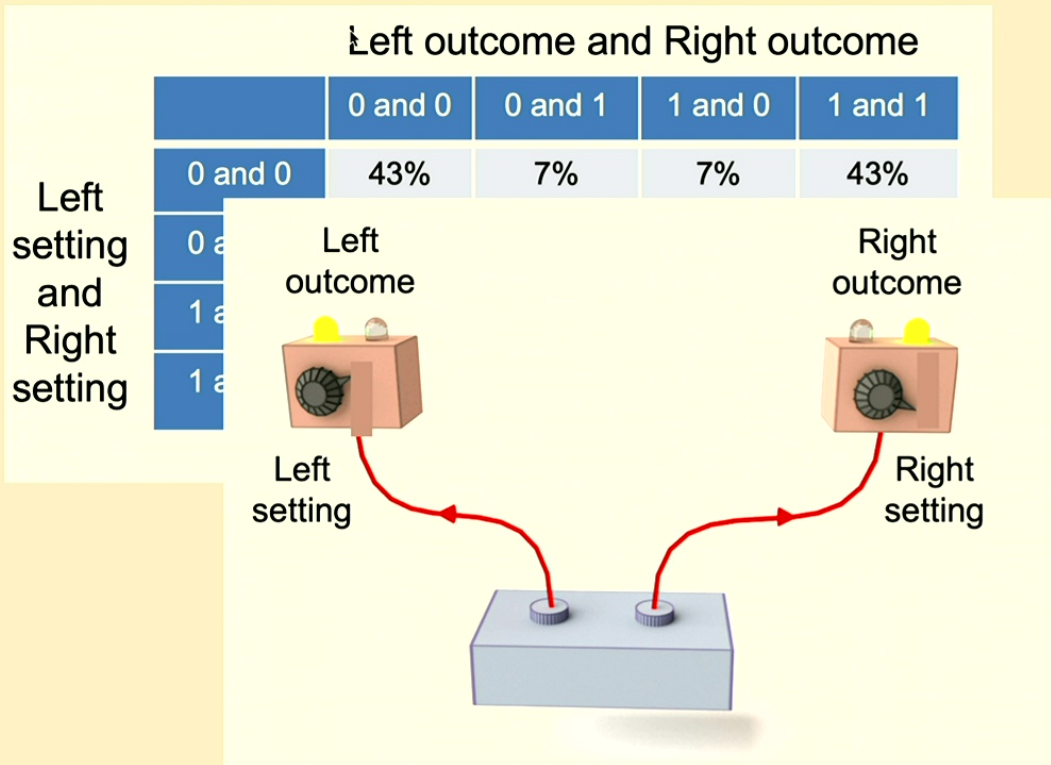
Implies a constraint:
Pearl's instrumental inequality



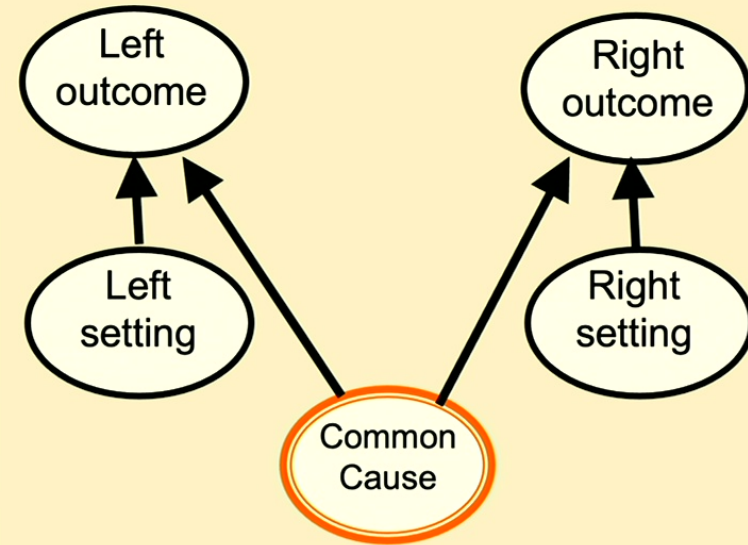
Bell inequalities

		Left outcome and Right outcome			
		0 and 0	0 and 1	1 and 0	1 and 1
Left setting and Right setting	0 and 0	43%	7%	7%	43%
	0 and 1	43%	7%	7%	43%
	1 and 0	43%	7%	7%	43%
	1 and 1	7%	43%	43%	7%

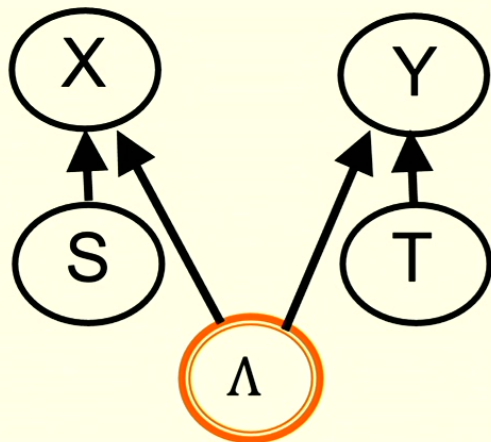
The evidence



The natural hypothesis



Causal structure



Parameters

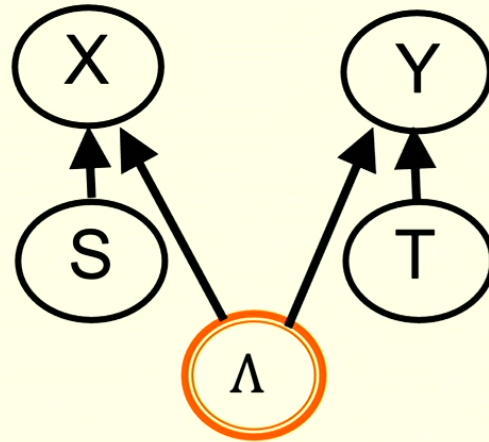
$$P_{X|S\Lambda}$$
$$P_{Y|T\Lambda}$$
$$P_{\Lambda}$$

$$P_{XY|ST} = \sum_{\Lambda} P_{Y|T\Lambda} P_{X|S\Lambda} P_{\Lambda}$$

Examples of causal compatibility constraints:

$$P_{X|ST} = P_{X|S}$$
$$P_{Y|ST} = P_{Y|T}$$

Causal structure



Parameters

$$P_{X|S\Lambda}$$
$$P_{Y|T\Lambda}$$
$$P_{\Lambda}$$

$$P_{XY|ST} = \sum_{\Lambda} P_{Y|T\Lambda} P_{X|S\Lambda} P_{\Lambda}$$

Examples of causal compatibility constraints:

$$P_{X|ST} = P_{X|S}$$
$$P_{Y|ST} = P_{Y|T}$$

$$\frac{1}{4} \sum_{x=y} P_{XY|ST}(xy|00) + \frac{1}{4} \sum_{x=y} P_{XY|ST}(xy|01)$$
$$\frac{1}{4} \sum_{x=y} P_{XY|ST}(xy|10) + \frac{1}{4} \sum_{x \neq y} P_{XY|ST}(xy|11) \leq \frac{3}{4}$$

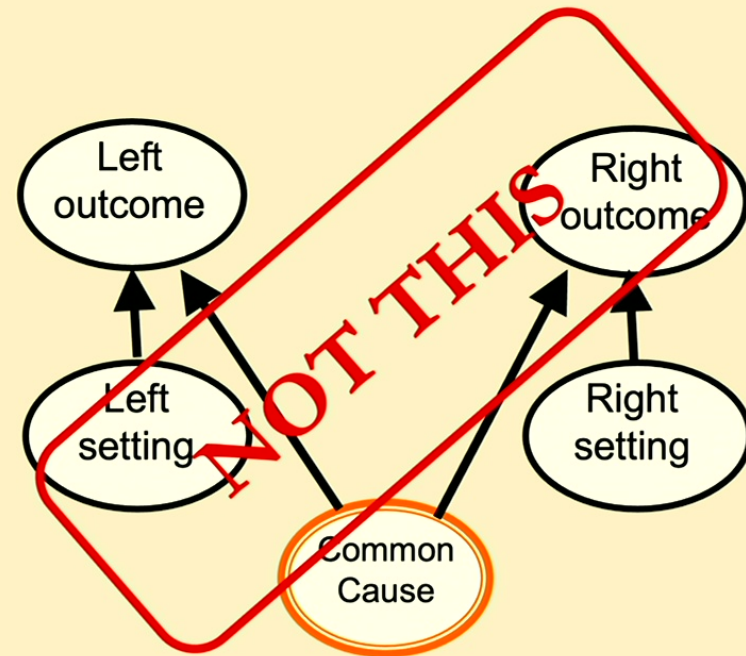
Clauser, Horne, Shimony and Holte, Phys. Rev. Lett.23, 880 (1967)

The evidence

		Left outcome and Right outcome			
		0 and 0	0 and 1	1 and 0	1 and 1
Left setting and Right setting	0 and 0	43%	7%	7%	43%
	0 and 1	43%	7%	7%	43%
	1 and 0	43%	7%	7%	43%
	1 and 1	7%	43%	43%	7%

Violates the
Bell Inequalities!

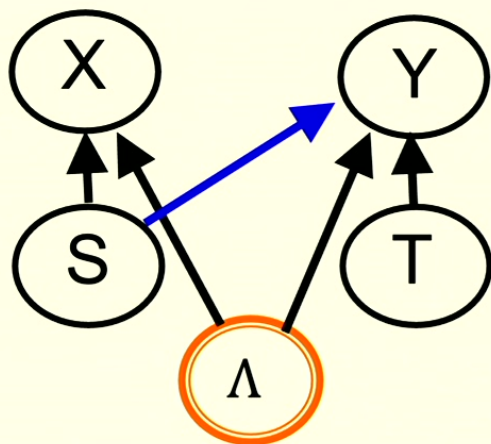
The natural hypothesis



Implies a constraint:
Bell Inequalities

Incompatible

Causal structure



Parameters

$$P_{X|S\Lambda}$$
$$P_{Y|ST\Lambda}$$
$$P_{\Lambda}$$

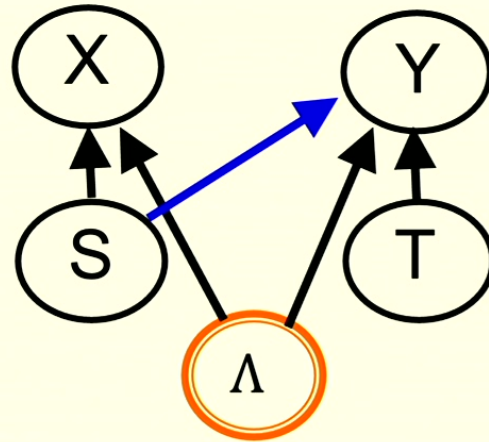
$$P_{XY|ST} = \sum_{\Lambda} P_{Y|ST\Lambda} P_{X|S\Lambda} P_{\Lambda}$$

Causal compatibility constraints:

$$P_{X|ST} = P_{X|S}$$

But the data *also* satisfies $P_{Y|ST} = P_{Y|T}$

Causal structure



Parameters

$$P_{X|S\Lambda}$$
$$P_{Y|ST\Lambda}$$
$$P_{\Lambda}$$

$$P_{XY|ST} = \sum_{\Lambda} P_{Y|ST\Lambda} P_{X|S\Lambda} P_{\Lambda}$$

Causal compatibility constraints:

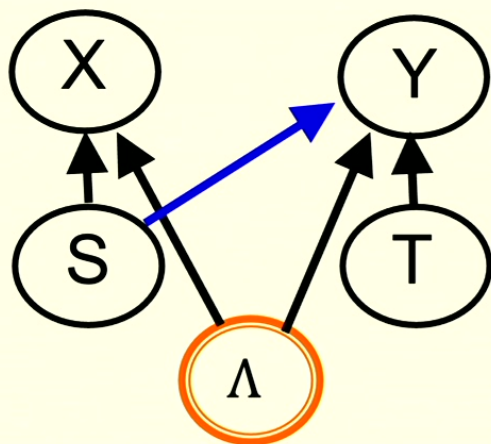
$$P_{X|ST} = P_{X|S}$$

But the data *also* satisfies $P_{Y|ST} = P_{Y|T}$

Reproducing this requires **fine-tuning**

Wood and RWS, New J. Phys. 17, 033002 (2015)

Causal structure



Parameters

$$P_{X|S\Lambda}$$
$$P_{Y|ST\Lambda}$$
$$P_{\Lambda}$$

$$P_{XY|ST} = \sum_{\Lambda} P_{Y|ST\Lambda} P_{X|S\Lambda} P_{\Lambda}$$

Causal compatibility constraints:

$$P_{X|ST} = P_{X|S}$$

But the data *also* satisfies $P_{Y|ST} = P_{Y|T}$

Reproducing this requires **violation of Leibniz's methodological principle**

Schmid, Selby, Spekkens, to appear

QUANTUM THEORY



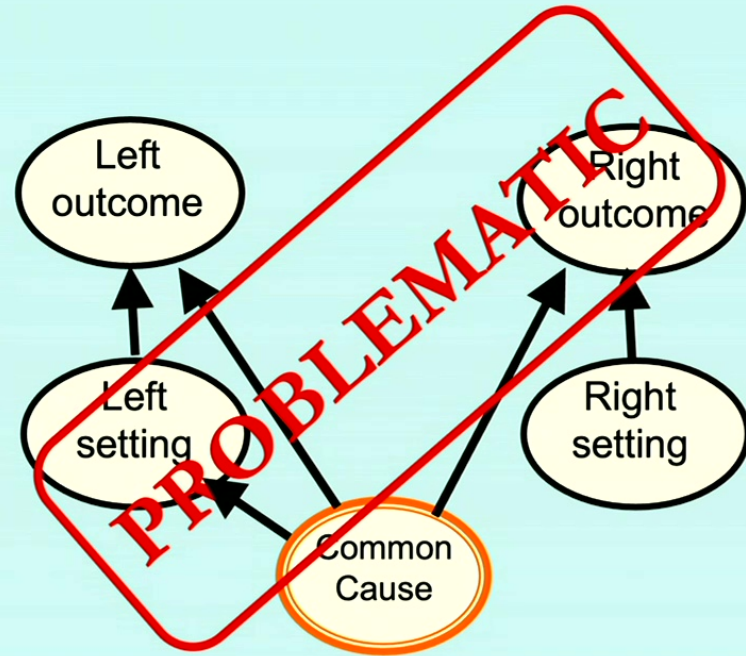
Diego Delso / CC BY-SA

RELATIVITY

The evidence

		Left outcome and Right outcome			
		0 and 0	0 and 1	1 and 0	1 and 1
Left setting and Right setting	0 and 0	43%	7%	7%	43%
	0 and 1	43%	7%	7%	43%
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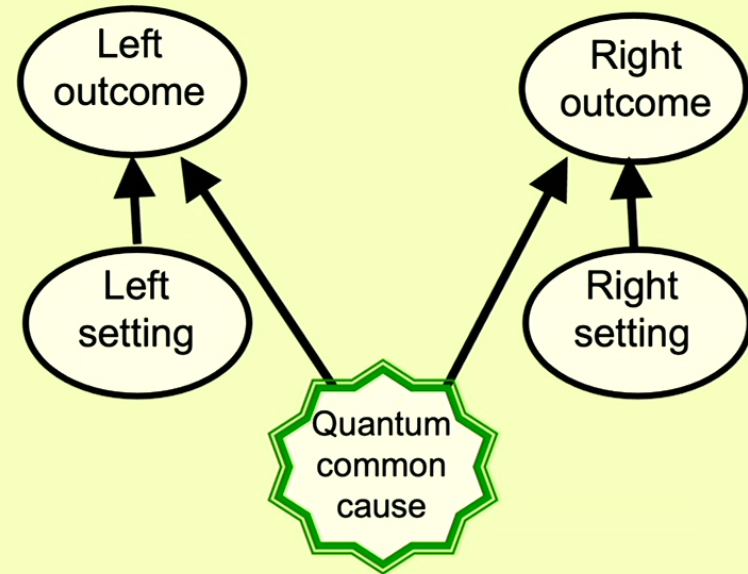
Superdeterminism



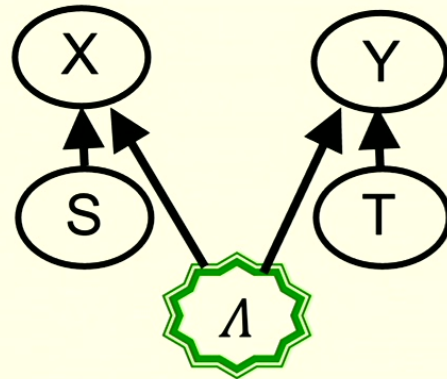
The evidence

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	1 and 0	43%	7%	7%	43%
	1 and 1	7%	43%	43%	7%

A new possibility



Quantum causal model



$$\rho_{X|S\Lambda}$$

$$\rho_{Y|T\Lambda}$$

$$\rho_{\Lambda}$$

$$[\rho_{X|S\Lambda}, \rho_{Y|T\Lambda}] = 0$$

$$P_{XY|ST} = \text{Tr}_{\Lambda}(\rho_{X|S\Lambda}\rho_{Y|T\Lambda}\rho_{\Lambda})$$

Leifer and RWS, PRA 88, 052130 (2013)

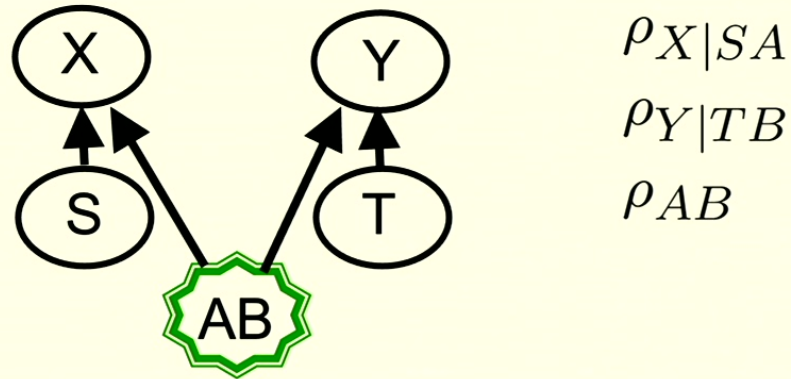
Henson, Lal & Pusey NJP 16, 113043 (2014)

Costa, Shrapnel NJP 18(6) (2016)

Allen, Barrett, Horsman, Lee & RWS, PRX 7, 031021 (2017)

Barrett, Lorenz, Oreshkov, arXiv:1906.10726

Quantum causal model



$$\rho_{X|SA}$$

$$\rho_{Y|TB}$$

$$\rho_{AB}$$

$$P_{XY|ST} = \text{Tr}_{AB}(\rho_{X|SA}\rho_{Y|TB}\rho_{AB})$$

Leifer and RWS, PRA 88, 052130 (2013)

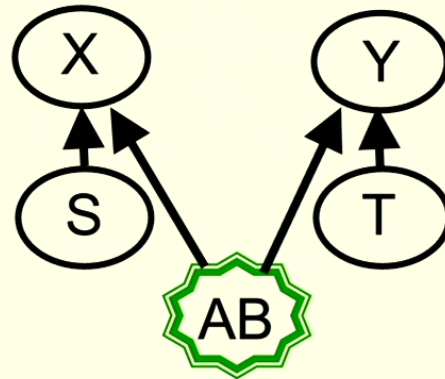
Henson, Lal & Pusey NJP 16, 113043 (2014)

Costa, Shrapnel NJP 18(6) (2016)

Allen, Barrett, Horsman, Lee & RWS, PRX 7, 031021 (2017)

Barrett, Lorenz, Oreshkov, arXiv:1906.10726

Quantum causal model



$\{E_{x|s}^A\}_x$ for each s

$\{E_{y|t}^B\}_y$ for each t

ρ_{AB}

$$P_{XY|ST}(xy|st) = \text{Tr}_{AB}((E_{x|s}^A \otimes E_{y|t}^B)\rho_{AB})$$

Leifer and RWS, PRA 88, 052130 (2013)

Henson, Lal & Pusey NJP 16, 113043 (2014)

Costa, Shrapnel NJP 18(6) (2016)

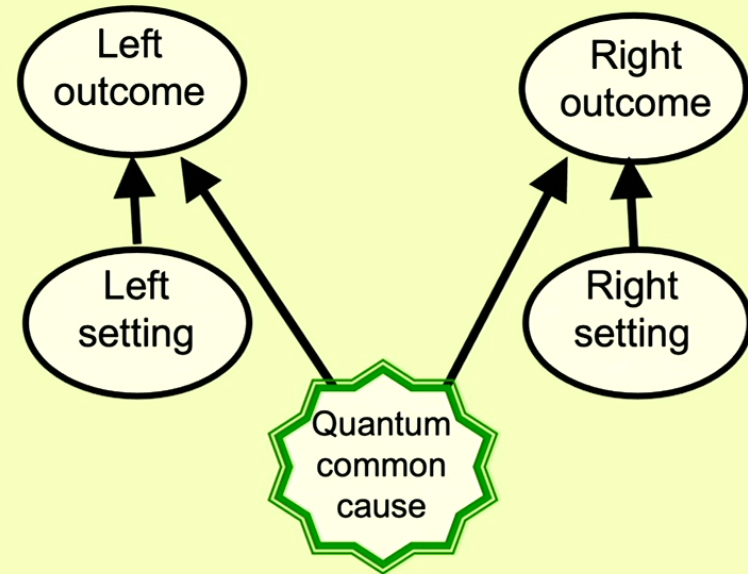
Allen, Barrett, Horsman, Lee & RWS, PRX 7, 031021 (2017)

Barrett, Lorenz, Oreshkov, arXiv:1906.10726

The evidence

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	1 and 1	7%	43%	43%	7%

A new possibility

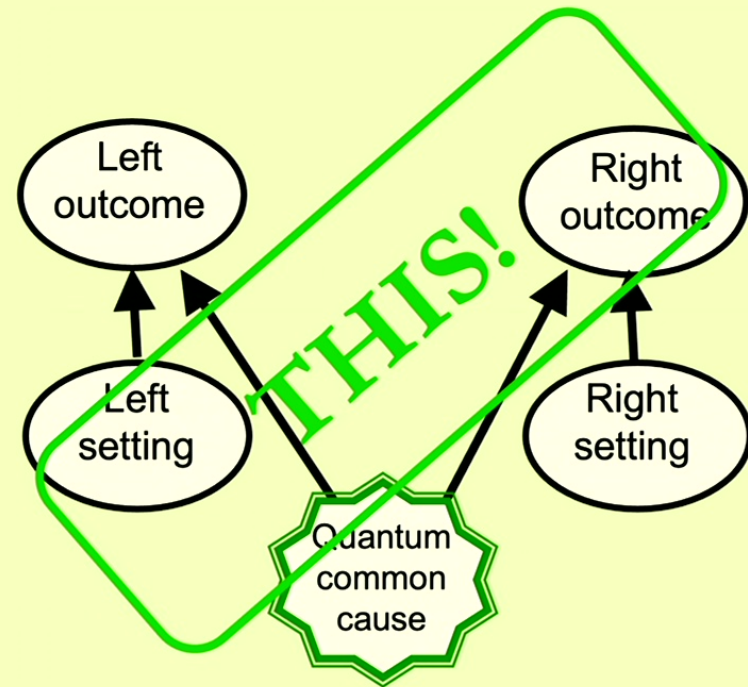


Compatible

The evidence

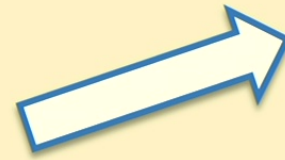
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		0 and 0	0 and 1	1 and 0	1 and 1
Left setting and Right setting	0 and 0	43%	7%	7%	43%
	0 and 1	43%	7%	7%	43%
	1 and 0	43%	7%	7%	43%
	1 and 1	7%	43%	43%	7%

A new possibility

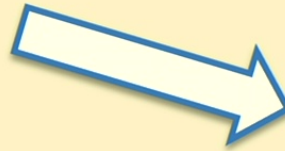


Compatible

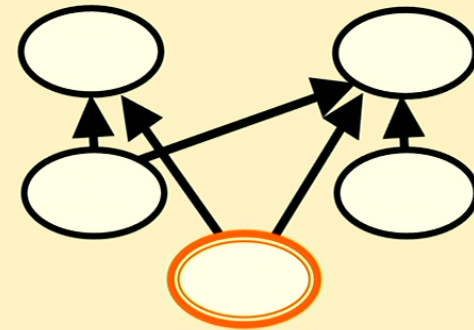
Violation of Bell inequalities



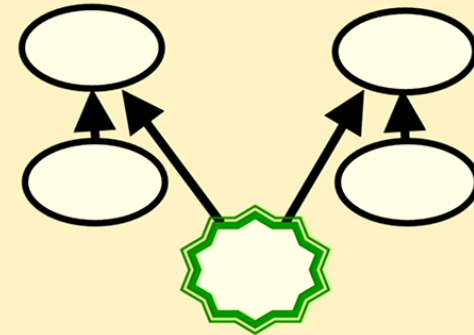
or



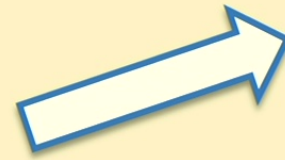
Witnessing need for different structure



Witnessing quantumness



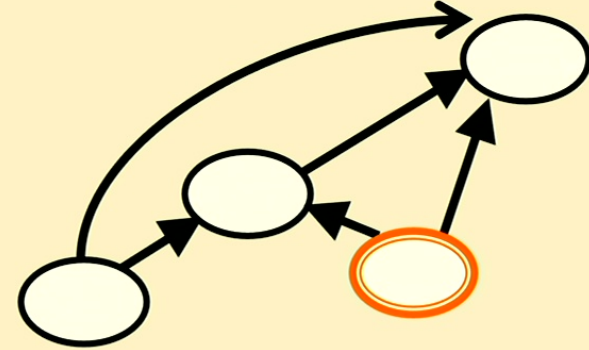
Violation of
Instrumental
Inequalities



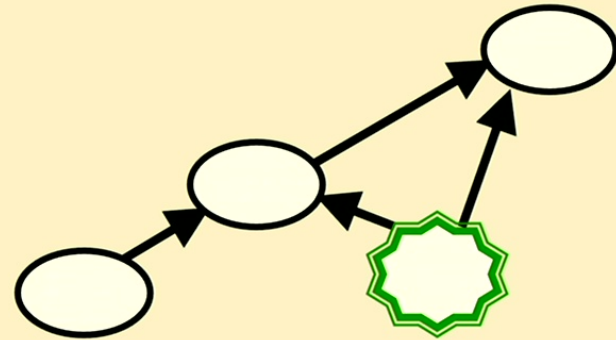
or



Witnessing need for
different structure



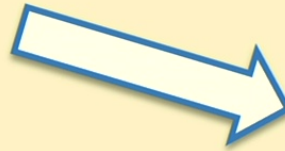
Witnessing
quantumness



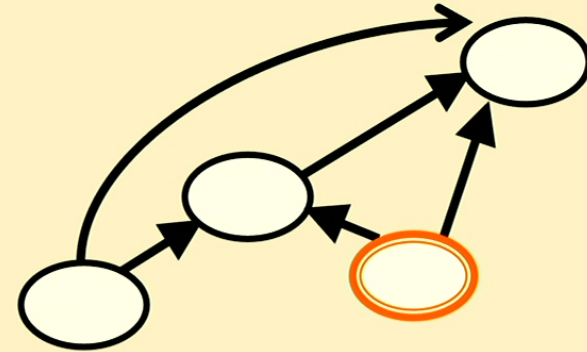
Violation of Instrumental Inequalities



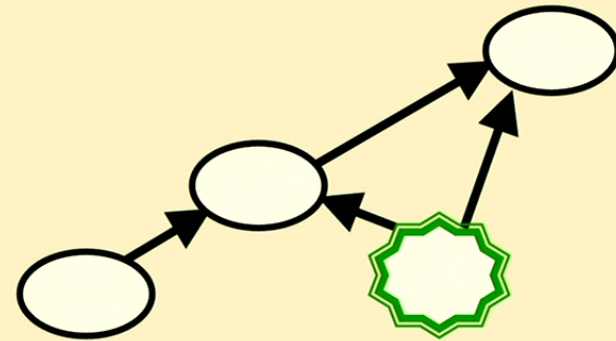
or



Witnessing need for different structure



Witnessing quantumness

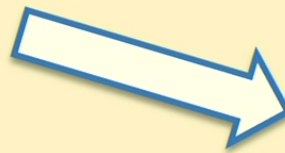


Van Himbeeck et al., Quantum 3 (2019): 186
Chaves et al., Nat. Phys. 47, 291296 (2018)

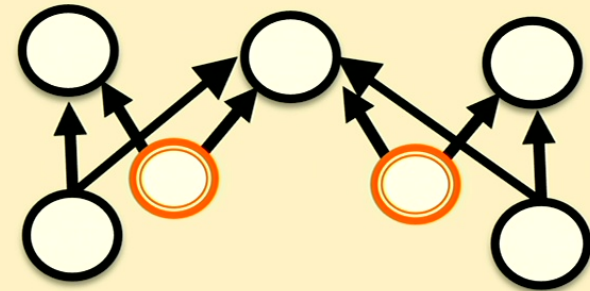
Violation of certain
causal compatibility
inequalities



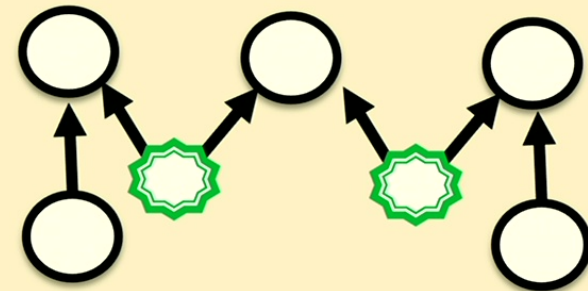
or



Witnessing need for
different structure

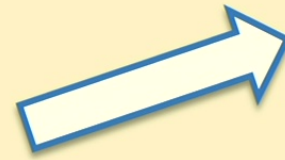


Witnessing
quantumness

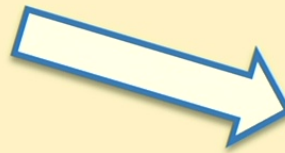


Branciard et al., Phys. Rev. A 85, 032119 (2012)

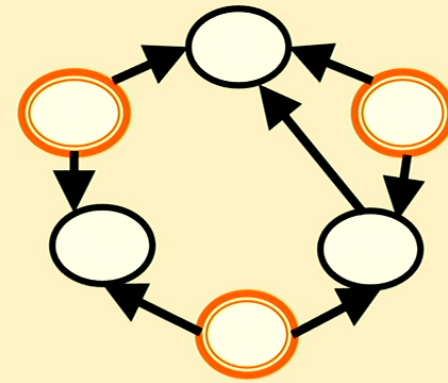
Violation of certain
causal compatibility
inequalities



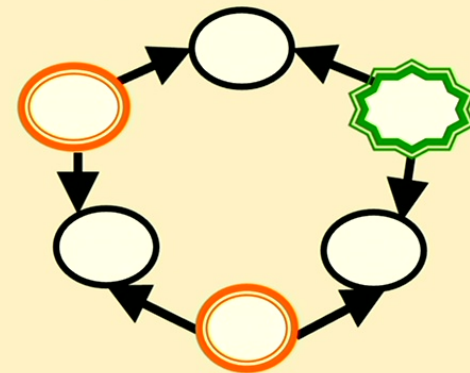
or



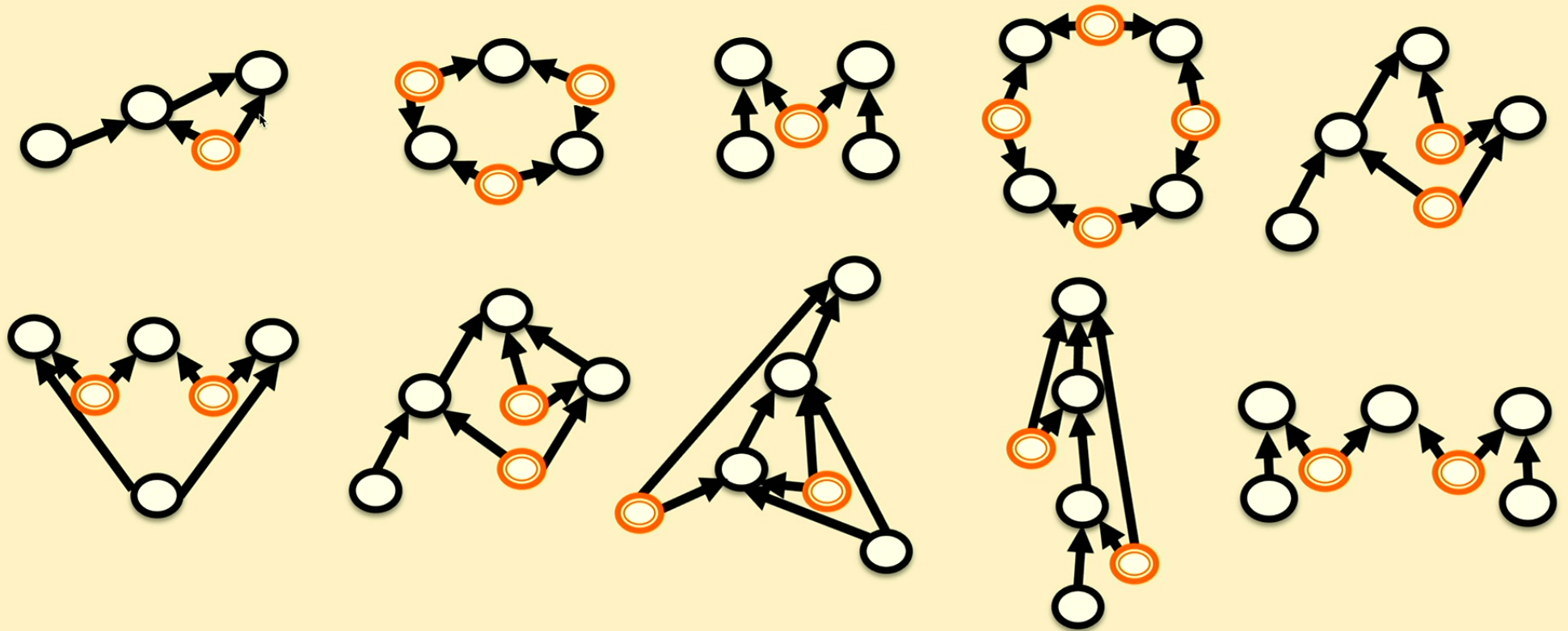
Witnessing need for
different structure

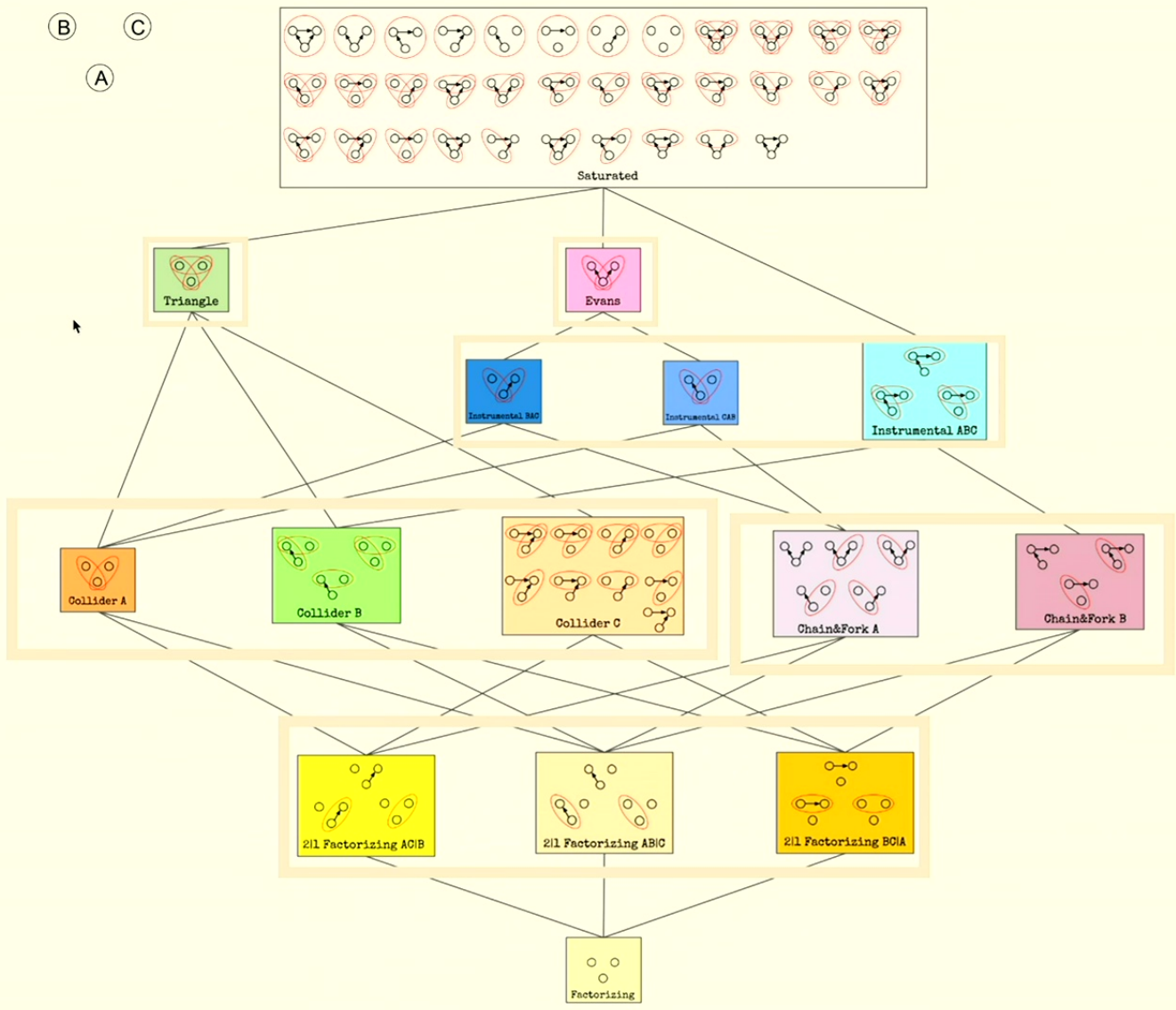


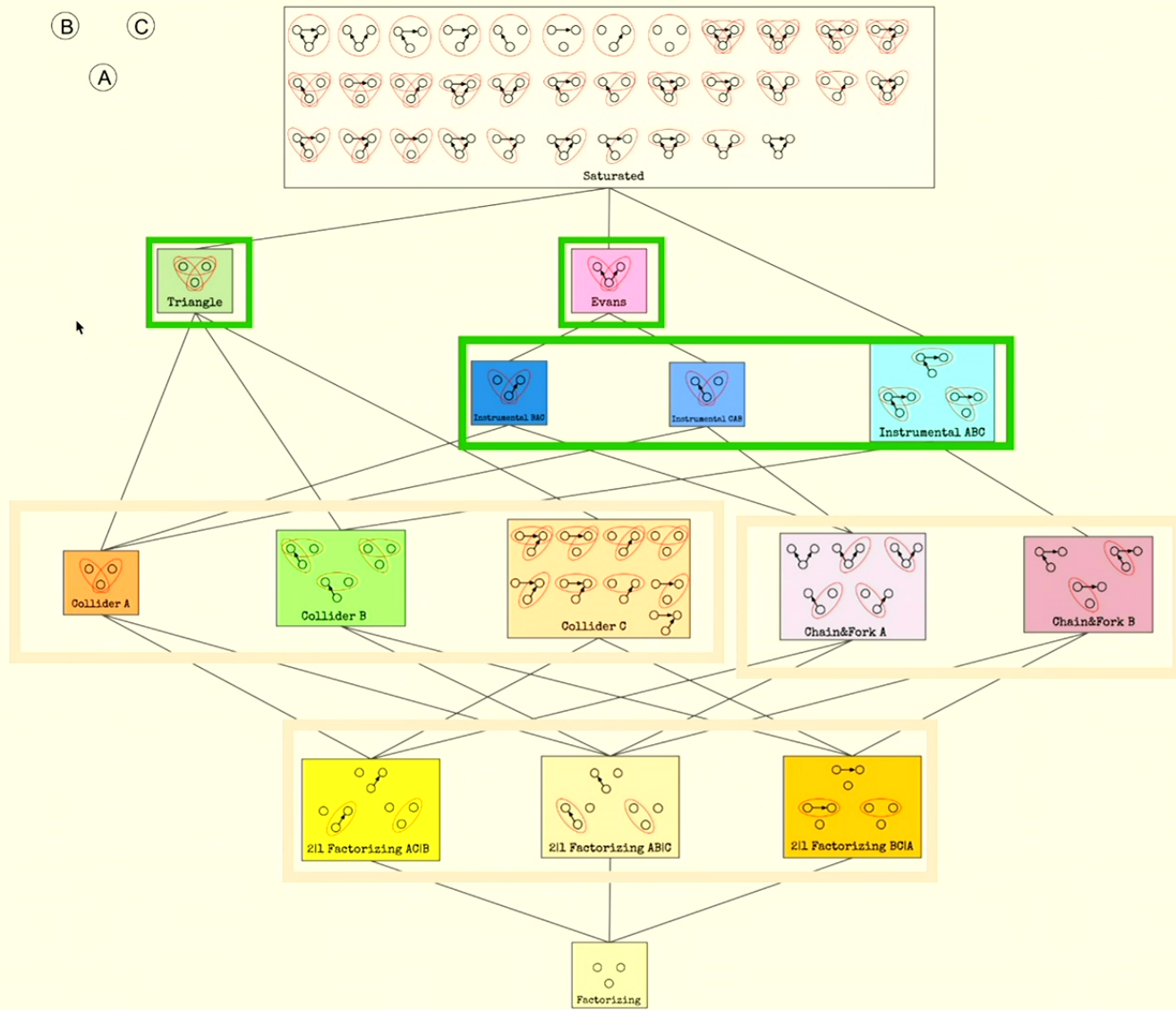
Witnessing
quantumness



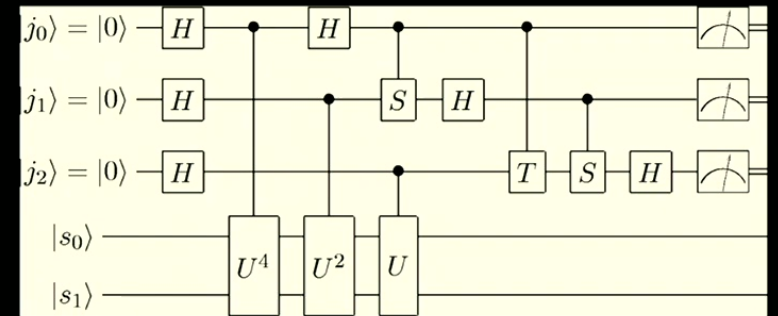
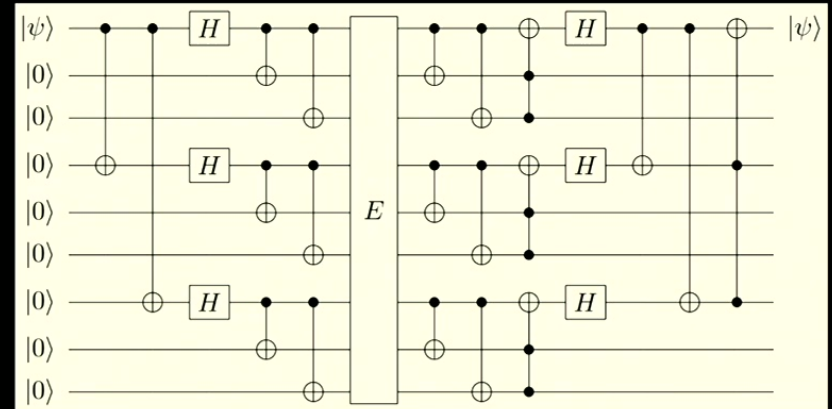
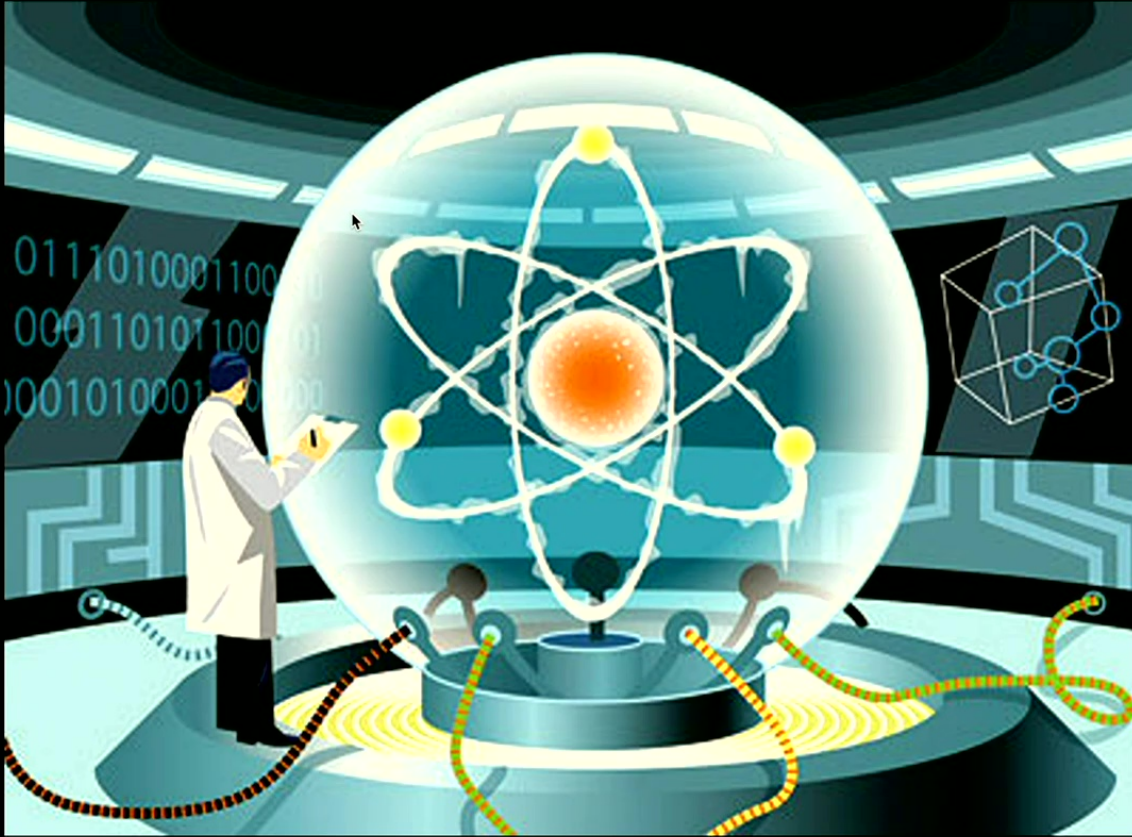
Some causal structures that admit of quantum-classical gaps:







Applications for Quantum Technology



The formalism and conceptual scheme of causal inference resolved various puzzles of statistics (e.g., Simpson's paradox, Berkson's paradox)

The lesson:

We must unscramble the omelette of inference and causation both conceptually and in the formalism

↳
But what hope do we have of succeeding
in the quantum context if we do not
understand how to do so in the classical
context?

Quantum
Causation and Inference

Classical
Causation and Inference



Relativistic
Notions of Space
and Time

PreRelativistic
Notions of Space
and Time



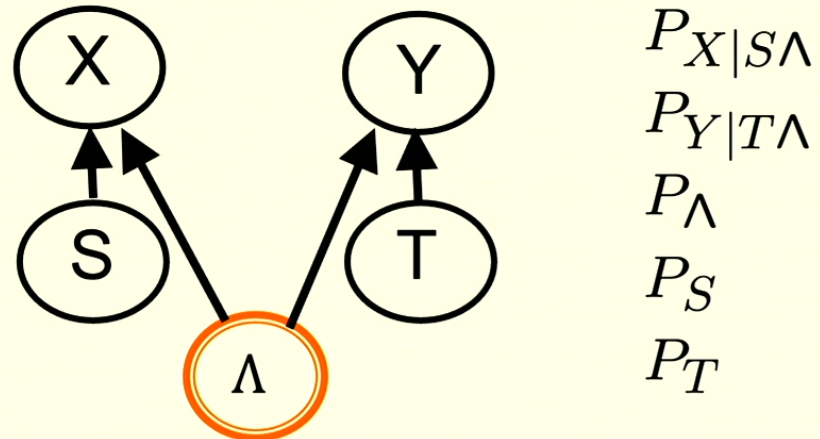
QUANTUM THEORY



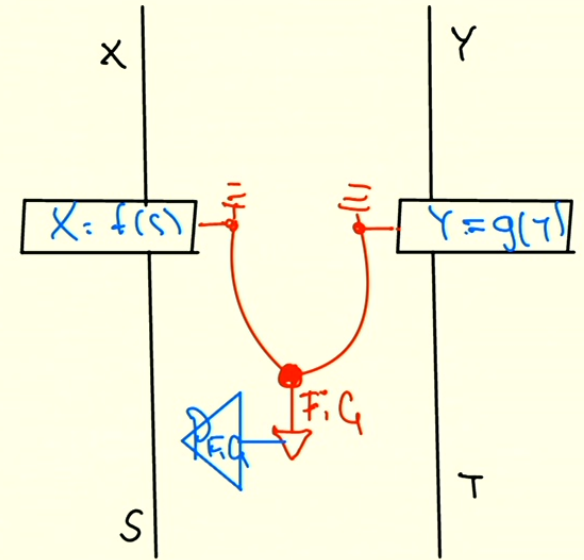
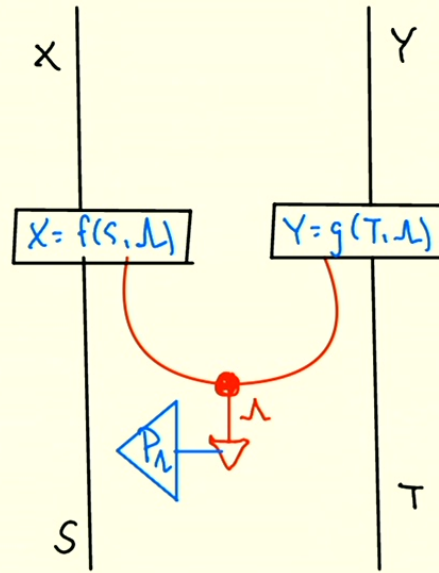
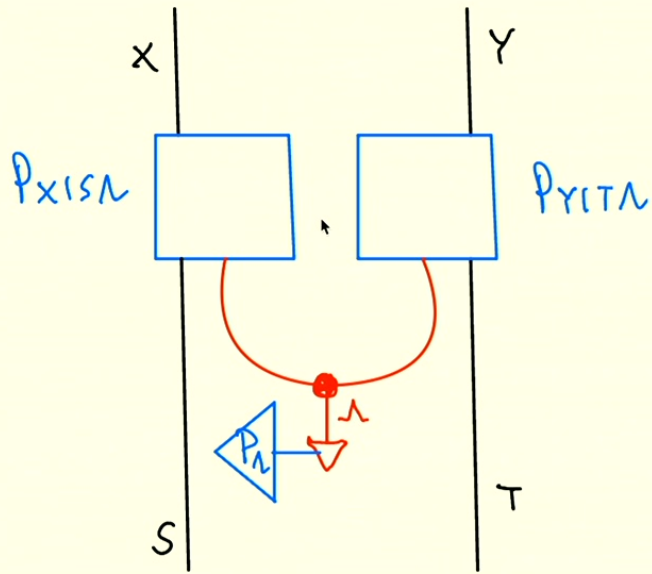
RELATIVITY

Deriving causal compatibility inequalities

Bell scenario



$$P_{XY|ST} = \sum_{\Lambda} P_{Y|T\Lambda} P_{X|S\Lambda} P_{\Lambda}$$



$$P_{XY|ST} = \sum_{f,g} \delta_{X,f(S)} \delta_{Y,g(T)} P_{F,G}(f, g)$$

If X,Y,S,T are binary, Λ can have cardinality 16

$$p_{xy|st} := P_{XY|ST}(xy|st)$$

$$x, y, s, t \in \{0, 1\}$$

$$q_{fg} := P_{F,G}(f, g)$$

$$f, g \in \{\text{id}, \text{fp}, r_0, r_1\}$$

$$p_{00|00} = q_{r_0, r_0} + q_{r_0, \text{id}} + q_{\text{id}, r_0} + q_{\text{id}, \text{id}}$$

$$p_{00|01} = q_{r_0, r_1} + q_{r_0, \text{fp}} + q_{\text{id}, r_1} + q_{\text{id}, \text{fp}}$$

$$0 \leq q_{fg} \leq 1 \quad \forall f, g$$

-
-
-

16 linear equalities + inequalities

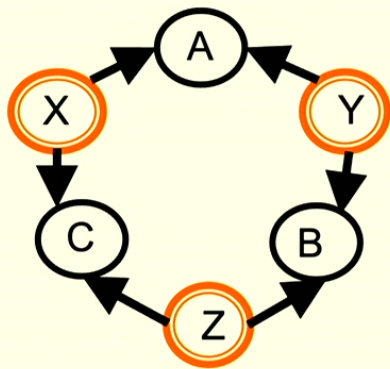
One must implement linear quantifier

elimination to implicitize the 16 q's.

Techniques for determining upper bounds on cardinalities of the latent variables in more general causal structures

R. Evans, *Annals of Statistics*, **46**, 2623 (2018)

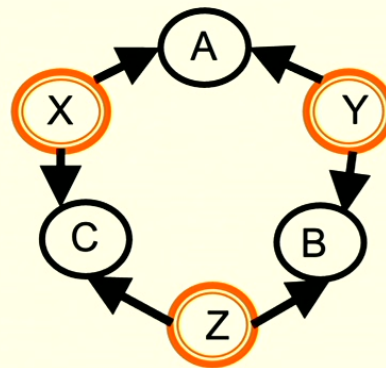
D. Rosset, N. Gisin, and E. Wolfe. *Quantum Inf. & Comp.* **18**, 0910 (2018)



A, B, C binary \rightarrow

Sufficient for
X, Y, Z to be 6-valued

Triangle scenario



$$P_{A|XY}$$

$$P_{B|YZ}$$

$$P_{C|XZ}$$

$$P_X$$

$$P_Y$$

$$P_Z$$

$$P_{ABC} = \sum_{XYZ} P_{A|XY} P_{B|YZ} P_{C|ZX} P_X P_Y P_Z$$

With more than one latent variable,
we require **nonlinear** quantifier elimination
which scales badly

Other techniques implicitizing parameters referring to hidden variables

Entropy cone techniques:

R. Chaves and T. Fritz, Phys. Rev. A 85 (2012)

T. Fritz, New J. Phys. 14 103001 (2012)

R. Chaves, L. Luft, D. Gross, New J. Phys. 16, 043001 (2014)

R. Chaves, L. Luft, T. O. Maciel, D. Gross, D. Janzing, B. Schölkopf, Proceedings of UAI 2014

M. Weilenmann and R. Colbeck, Proc. Roy. Soc. A 473.2207 (2017): 20170483.

Covariance matrix techniques:

A. Kela, K. von Prillwitz, J. Aberg, R. Chaves, and D. Gross, arXiv:1701.00652 (2017).

4

Strategy 2: Implicitization of parameters referring to counterfactual possibilities

Example 1:

$$\begin{aligned}P_{XY}^{\text{target}} &= \frac{1}{2}[00] + \frac{1}{2}[11] \\P_{YZ}^{\text{target}} &= \frac{1}{2}[01] + \frac{1}{2}[10] \\P_{XZ}^{\text{target}} &= \frac{1}{2}[01] + \frac{1}{2}[10]\end{aligned}$$

Question: $\exists P_{XYZ}$

such that

$$\begin{aligned}P_{XY} &= P_{XY}^{\text{target}} \\P_{YZ} &= P_{YZ}^{\text{target}} \\P_{XZ} &= P_{XZ}^{\text{target}}\end{aligned}$$

where

$$\begin{aligned}P_{XY} &:= \sum_Z P_{XYZ} \\P_{YZ} &:= \sum_X P_{XYZ} \\P_{XZ} &:= \sum_Y P_{XYZ}\end{aligned}$$

?

Example 2:

$$\begin{aligned} P_{XY}^{\text{target}} &= \frac{1}{2}[00] + \frac{1}{2}[11] = \frac{1}{2}[01] + \frac{1}{2}[10] \\ P_{YZ}^{\text{target}} &= \frac{1}{2}[01] + \frac{1}{2}[10] \\ P_{XZ}^{\text{target}} &= \frac{1}{2}[01] + \frac{1}{2}[10] \end{aligned}$$

Question: $\exists P_{XYZ}$

such that

$$\begin{aligned} P_{XY} &= P_{XY}^{\text{target}} \\ P_{YZ} &= P_{YZ}^{\text{target}} \\ P_{XZ} &= P_{XZ}^{\text{target}} \end{aligned}$$

where

$$\begin{aligned} P_{XY} &:= \sum_Z P_{XYZ} \\ P_{YZ} &:= \sum_X P_{XYZ} \\ P_{XZ} &:= \sum_Y P_{XYZ} \end{aligned}$$

?

Example 2:

$$\begin{aligned} P_{XY}^{\text{target}} &= \frac{1}{2}[00] + \frac{1}{2}[11] = \frac{1}{2}[01] + \frac{1}{2}[10] \\ P_{YZ}^{\text{target}} &= \frac{1}{2}[01] + \frac{1}{2}[10] \\ P_{XZ}^{\text{target}} &= \frac{1}{2}[01] + \frac{1}{2}[10] \end{aligned}$$

Question: $\exists P_{XYZ}$

such that

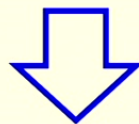
$$\begin{aligned} P_{XY} &= P_{XY}^{\text{target}} \\ P_{YZ} &= P_{YZ}^{\text{target}} \\ P_{XZ} &= P_{XZ}^{\text{target}} \end{aligned} \quad \text{where}$$
$$\begin{aligned} P_{XY} &:= \sum_Z P_{XYZ} \\ P_{YZ} &:= \sum_X P_{XYZ} \\ P_{XZ} &:= \sum_Y P_{XYZ} \end{aligned} \quad ?$$

Answer: no!

consider $[000], [001], [010], [011], [100], [101], [110], [111]$

Consider binary X, Y and Z

$\exists P_{XYZ}$ with $P_X, P_Y, P_Z, P_{XY}, P_{YZ}, P_{XZ}$ as marginals



Marginal compatibility constraints

$$P_X = \sum_Y P_{XY} = \sum_Z P_{XZ}$$

$$P_Y = \sum_X P_{XY} = \sum_Z P_{YZ}$$

$$P_Z = \sum_X P_{XZ} = \sum_Y P_{YZ}$$

$$P_X + P_Y + P_Z - P_{XY} - P_{YZ} - P_{XZ} \leq 1$$

How this is proven:

$\exists Q_{XYZ}$ with $P_X, P_Y, P_Z, P_{XY}, P_{YZ}, P_{XZ}$ as marginals

$$\forall x, y : P_{XY}(xy) = \sum_z Q_{XYZ}(xyz)$$

$$\forall y, z : P_{YZ}(yz) = \sum_x Q_{XYZ}(xyz)$$

$$\forall x, z : P_{XZ}(xz) = \sum_y Q_{XYZ}(xyz)$$

$$\forall x : P_X(x) = \sum_{y,z} Q_{XYZ}(xyz)$$

$$\forall y : P_Y(y) = \sum_{x,z} Q_{XYZ}(xyz)$$

$$\forall z : P_Z(z) = \sum_{x,y} Q_{XYZ}(xyz)$$

$$\forall x, y, z : 0 \leq Q_{XYZ}(xyz) \leq 1$$

Linear quantifier
elimination



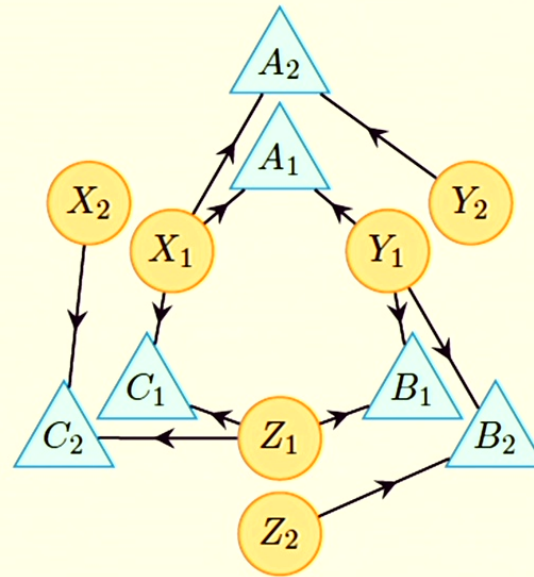
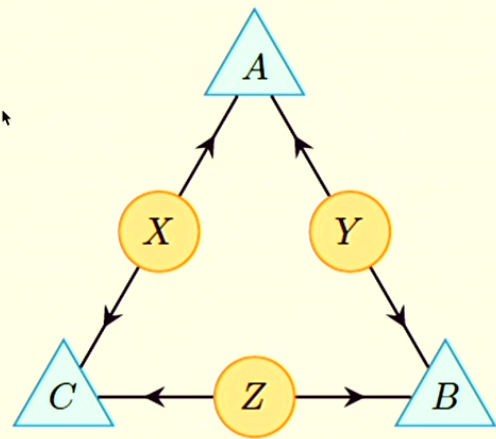
$$\forall x : P_X(x) = \sum_y P_{XY}(xy) = \sum_z P_{XZ}(xz)$$

$$\forall y : P_Y(y) = \sum_x P_{XY}(xy) = \sum_z P_{YZ}(yz)$$

$$\forall z : P_Z(z) = \sum_x P_{XZ}(xz) = \sum_y P_{YZ}(yz)$$

$$\forall x, y, z : P_X(x) + P_Y(y) + P_Z(z) - P_{XY}(xy) - P_{YZ}(yz) - P_{XZ}(xz) \leq 1$$

Inflation DAGs

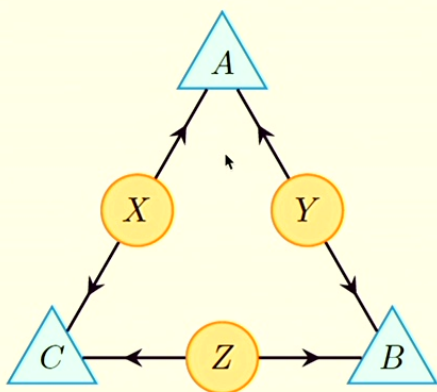


$G' \in \text{inflations}(G)$

if and only if

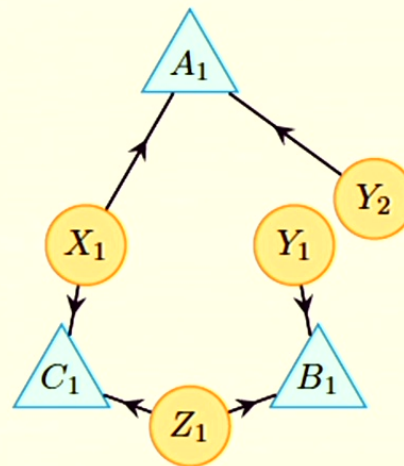
$\forall A_i \in \text{nodes}(G') : \text{ansubgraph}_{G'}(A_i) \sim \text{ansubgraph}_G(A).$

model M on DAG G



$P_{A|XY}$
 $P_{B|YZ}$
 $P_{C|XZ}$
 P_X
 P_Y
 P_Z

$M' = G \rightarrow G'$ Inflation of M

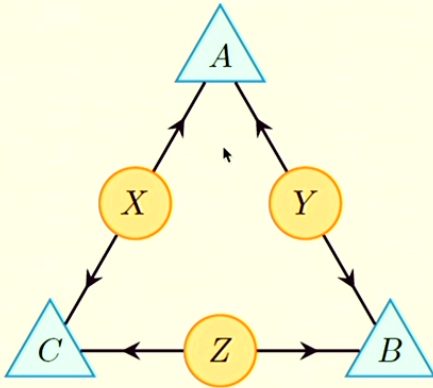


$P_{A_1|X_1 Y_2}$
 $P_{B_1|Y_1 Z_1}$
 $P_{C_1|X_1 Z_1}$
 P_{X_1}
 P_{Y_1}
 P_{Y_2}
 P_{Z_1}

with symmetry constraint:

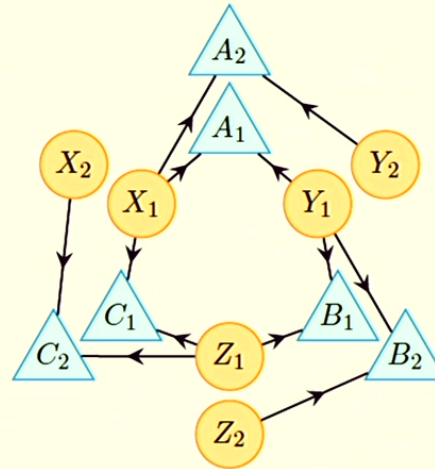
$$P_{Y_1} = P_{Y_2}$$

model M on DAG G



$P_{A|XY}$
 $P_{B|YZ}$
 $P_{C|XZ}$
 P_X
 P_Y
 P_Z

$M' = G \rightarrow G'$ Inflation of M

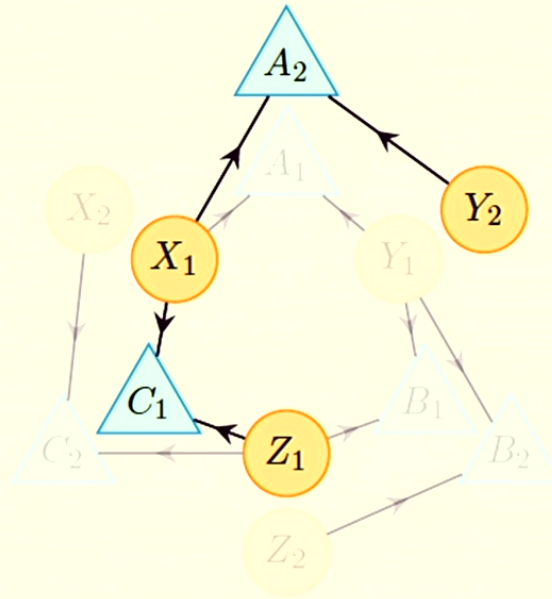
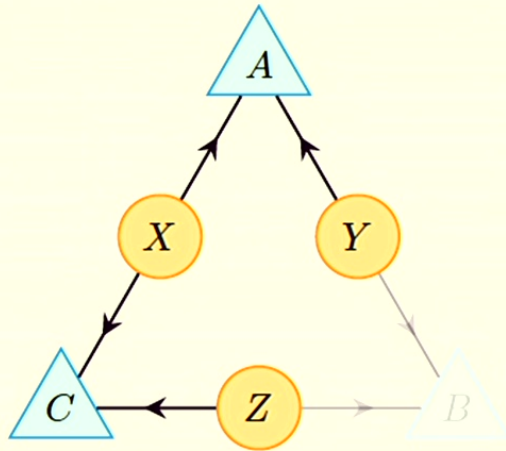


$P_{A_1|X_1Y_1}$ $P_{A_2|X_1Y_2}$
 $P_{B_1|Y_1Z_1}$ $P_{B_2|Y_1Z_2}$
 $P_{C_1|X_1Z_1}$ $P_{C_2|X_2Z_1}$
 P_{X_1} P_{X_2}
 P_{Y_1} P_{Y_2}
 P_{Z_1} P_{Z_2}

with symmetry constraints:

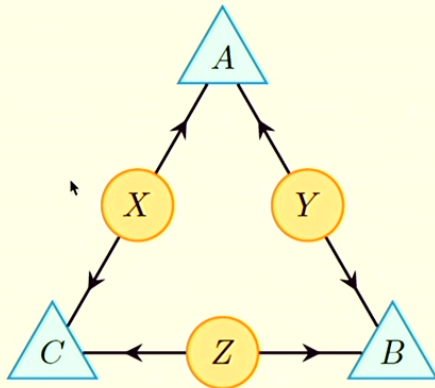
$$\begin{aligned}
 P_{A_1|X_1Y_1} &= P_{A_2|X_1Y_2} \\
 P_{B_1|Y_1Z_1} &= P_{B_2|Y_1Z_2} \\
 P_{C_1|X_1Z_1} &= P_{C_2|X_2Z_1} \\
 P_{X_1} &= P_{X_2} \\
 P_{Y_1} &= P_{Y_2} \\
 P_{Z_1} &= P_{Z_2}
 \end{aligned}$$

Injectable sets of observed variables
in the inflation DAG



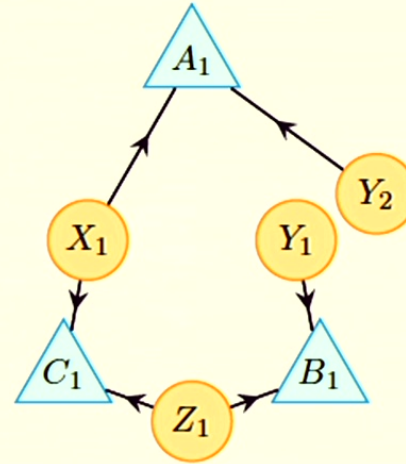
$\{A_2C_1\}$ is an injectable set

model M on DAG G



$P_{A|XY}$
 $P_{B|YZ}$
 $P_{C|XZ}$
 P_X
 P_Y
 P_Z

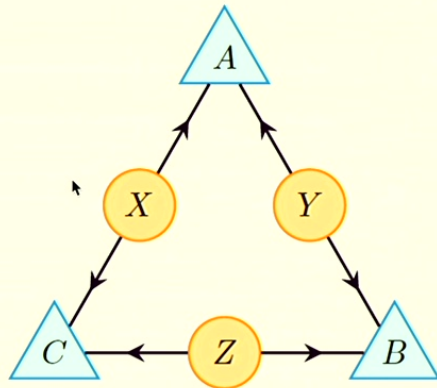
$M' = G \rightarrow G'$ Inflation of M



$P_{A_1|X_1Y_2}$
 $P_{B_1|Y_1Z_1}$
 $P_{C_1|X_1Z_1}$
 P_{X_1}
 P_{Y_1}
 P_{Y_2}
 P_{Z_1}

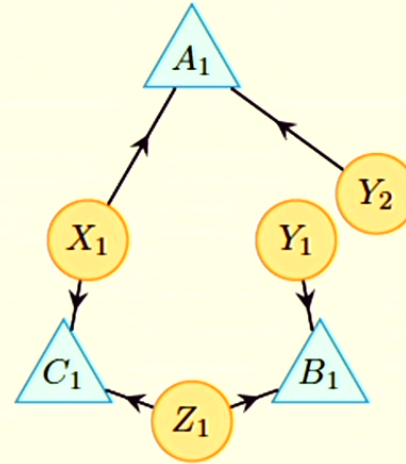
$\{A_1C_1\}$ is an injectable set

model M on DAG G



- $P_{A|XY}$
- $P_{B|YZ}$
- $P_{C|XZ}$
- P_X
- P_Y
- P_Z

$M' = G \rightarrow G'$ Inflation of M



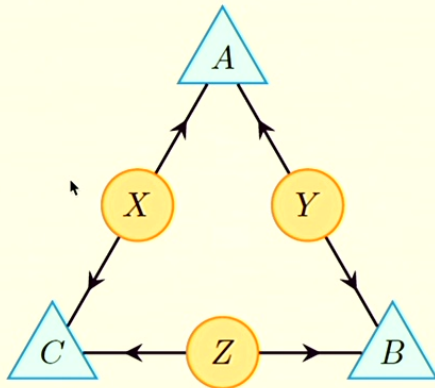
- $P_{A_1|X_1Y_2}$
- $P_{B_1|Y_1Z_1}$
- $P_{C_1|X_1Z_1}$
- P_{X_1}
- P_{Y_1}
- P_{Y_2}
- P_{Z_1}

$\{A_1C_1\}$ is an injectable set

$$P_{A_1C_1} = \sum_{X_1Y_2Z_1} P_{A_1|X_1Y_2} P_{C_1|X_1Z_1} P_{X_1} P_{Y_2} P_{Z_1}$$

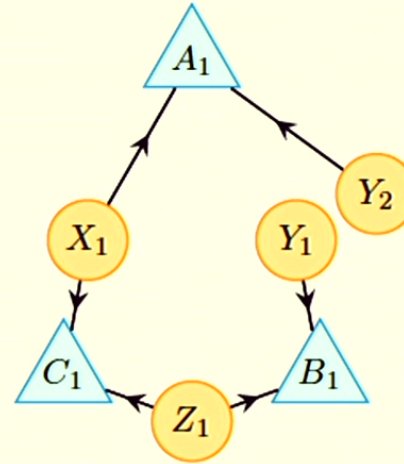
$$P_{AC} = \sum_{XYZ} P_{A|XY} P_{C|XZ} P_X P_Y P_Z$$

model M on DAG G



$$\begin{aligned}
 &P_{A|XY} \\
 &P_{B|YZ} \\
 &P_{C|XZ} \\
 &P_X \\
 &P_Y \\
 &P_Z
 \end{aligned}$$

$M' = G \rightarrow G'$ Inflation of M



$$\begin{aligned}
 &P_{A_1|X_1Y_2} \\
 &P_{B_1|Y_1Z_1} \\
 &P_{C_1|X_1Z_1} \\
 &P_{X_1} \\
 &P_{Y_1} \\
 &P_{Y_2} \\
 &P_{Z_1}
 \end{aligned}$$

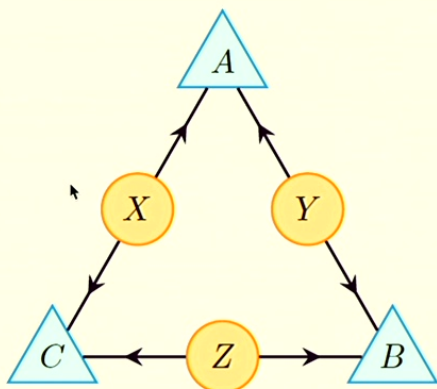
$\{A_1C_1\}$ is an injectable set

$$P_{A_1C_1} = \sum_{X_1Y_2Z_1} P_{A_1|X_1Y_2} P_{C_1|X_1Z_1} P_{X_1} P_{Y_2} P_{Z_1}$$

$$P_{AC} = \sum_{XYZ} P_{A|XY} P_{C|XZ} P_X P_Y P_Z$$

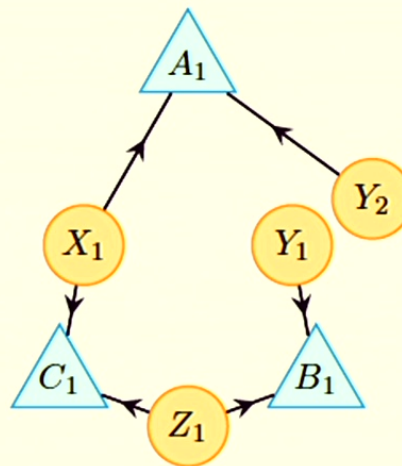
$$P_{AC} \text{ compatible with } M \implies P_{A_1C_1} = P_{AC} \text{ compatible with } M'$$

model M on DAG G



- $P_{A|XY}$
- $P_{B|YZ}$
- $P_{C|XZ}$
- P_X
- P_Y
- P_Z

$M' = G \rightarrow G'$ Inflation of M



- $P_{A_1|X_1Y_2}$
- $P_{B_1|Y_1Z_1}$
- $P_{C_1|X_1Z_1}$
- P_{X_1}
- P_{Y_1}
- P_{Y_2}
- P_{Z_1}

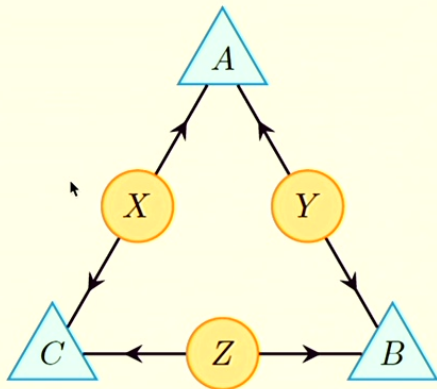
$\{A_1B_1\}$ is *not* an injectable set

$$P_{A_1B_1} = \left(\sum_{X_1Y_2} P_{A_1|X_1Y_2} P_{Y_2} P_{X_1} \right) \left(\sum_{Z_1Y_1} P_{B_1|Y_1Z_1} P_{Y_1} P_{Z_1} \right)$$

$$P_{AB} = \sum_{XYZ} P_{A|XY} P_{B|YZ} P_X P_Y P_Z$$

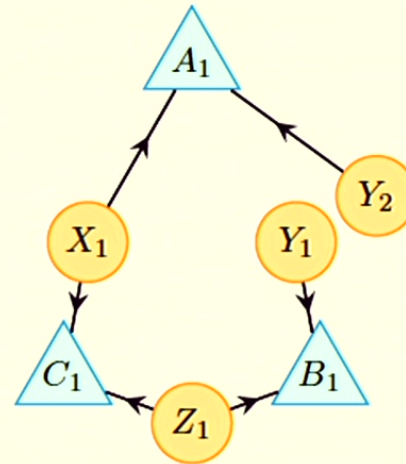
P_{AB} compatible with M $\not\Rightarrow$ $P_{A_1B_1} = P_{AB}$ compatible with M'

model M on DAG G



$P_{A|XY}$
 $P_{B|YZ}$
 $P_{C|XZ}$
 P_X
 P_Y
 P_Z

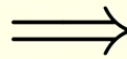
$M' = G \rightarrow G'$ Inflation of M



$P_{A_1|X_1Y_2}$
 $P_{B_1|Y_1Z_1}$
 $P_{C_1|X_1Z_1}$
 P_{X_1}
 P_{Y_1}
 P_{Y_2}
 P_{Z_1}

Injectable sets: $\{A_1\}, \{B_1\}, \{C_1\}, \{A_1C_1\}, \{B_1C_1\}$

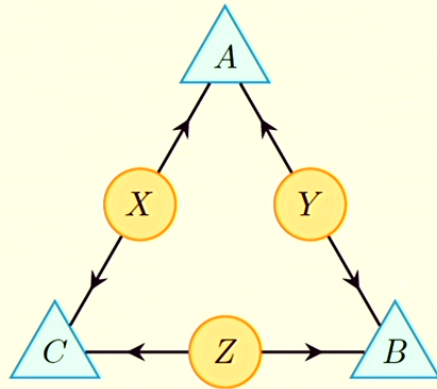
$(P_A, P_B, P_C, P_{AC}, P_{BC})$
 compatible with M



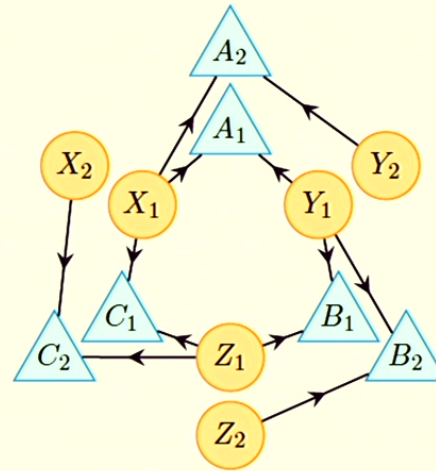
$(P_{A_1}, P_{B_1}, P_{C_1}, P_{A_1C_1}, P_{B_1C_1})$
 compatible with M'

where $P_{A_1} = P_A$ $P_{A_1C_1} = P_{AC}$
 $P_{B_1} = P_B$ $P_{B_1C_1} = P_{BC}$
 $P_{C_1} = P_C$

model M on DAG G



$M' = G \rightarrow G'$ Inflation of M



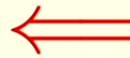
Injectable sets: $\{A_1\}, \{B_1\}, \{C_1\}, \{A_2\}, \{B_2\}, \{C_2\},$
 $\{A_1B_1\}, \{A_1, B_2\}, \{B_1C_1\}, \{B_1, C_2\}, \{C_1, A_1\}, \{C_1, A_2\},$
 $\{A_1B_1C_1\}$

$M' = G \rightarrow G'$ Inflation of M

$\mathcal{S} \subseteq \text{ImagesInjectableSets}(G)$

$\mathcal{S}' \subseteq \text{InjectableSets}(G')$

$\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$
is **not** compatible with M



$\{P_{\mathbf{V}'} : \mathbf{V}' \in \mathcal{S}'\}$
where $P_{\mathbf{V}'} = P_{\mathbf{V}}$ for $\mathbf{V}' \sim \mathbf{V}$
is **not** compatible with M'

$M' = G \rightarrow G'$ Inflation of M

$\mathcal{S} \subseteq \text{ImagesInjectableSets}(G)$

$\mathcal{S}' \subseteq \text{InjectableSets}(G')$

$\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$
is compatible with M

\implies

$\{P_{\mathbf{V}'} : \mathbf{V}' \in \mathcal{S}'\}$
where $P_{\mathbf{V}'} = P_{\mathbf{V}}$ for $\mathbf{V}' \sim \mathbf{V}$
is compatible with M'

Let $I_{\mathcal{S}}$ be an inequality that acts on
the family of distributions $\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$

Deriving
causal compatibility inequalities
by the inflation technique

$M' = G \rightarrow G'$ Inflation of M

$\mathcal{S} \subseteq \text{ImagesInjectableSets}(G)$

$\mathcal{S}' \subseteq \text{InjectableSets}(G')$

$\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$
is compatible with M

\implies

$\{P_{\mathbf{V}'} : \mathbf{V}' \in \mathcal{S}'\}$
where $P_{\mathbf{V}'} = P_{\mathbf{V}}$ for $\mathbf{V}' \sim \mathbf{V}$
is compatible with M'

$M' = G \rightarrow G'$ Inflation of M

$\mathcal{S} \subseteq \text{ImagesInjectableSets}(G)$

$\mathcal{S}' \subseteq \text{InjectableSets}(G')$

$I_{\mathcal{S}'}$ is a **causal compatibility inequality** for model M'

$\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$
is compatible with M

\implies

$\{P_{\mathbf{V}'} : \mathbf{V}' \in \mathcal{S}'\}$
where $P_{\mathbf{V}'} = P_{\mathbf{V}}$ for $\mathbf{V}' \sim \mathbf{V}$
is compatible with M'

\Downarrow

$I_{\mathcal{S}}$ is **satisfied** for
 $\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$

\Longleftarrow

$I_{\mathcal{S}'}$ is **satisfied** for
 $\{P_{\mathbf{V}'} : \mathbf{V}' \in \mathcal{S}'\}$
where $P_{\mathbf{V}'} = P_{\mathbf{V}}$ for $\mathbf{V}' \sim \mathbf{V}$

$M' = G \rightarrow G'$ Inflation of M

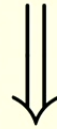
$\mathcal{S} \subseteq \text{ImagesInjectableSets}(G)$

$\mathcal{S}' \subseteq \text{InjectableSets}(G')$

$I_{\mathcal{S}'}$ is a **causal compatibility inequality** for model M'

$\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$

is compatible with M



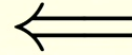
$I_{\mathcal{S}}$ is **satisfied** for
 $\{P_{\mathbf{V}} : \mathbf{V} \in \mathcal{S}\}$

$M' = G \rightarrow G'$ Inflation of M

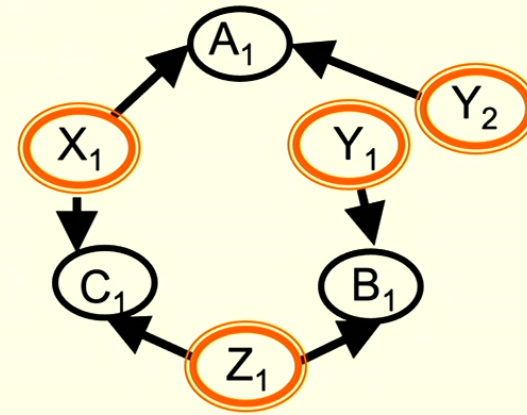
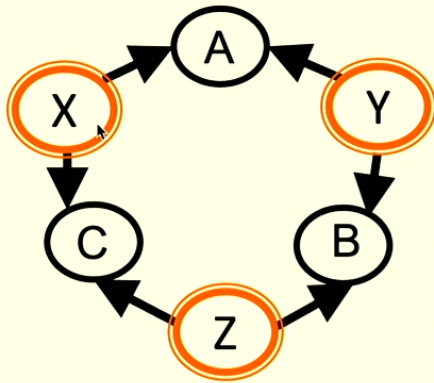
$\mathcal{S} \subseteq \text{ImagesInjectableSets}(G)$

$\mathcal{S}' \subseteq \text{InjectableSets}(G')$

$I_{\mathcal{S}}$ is a **causal compatibility inequality** for model M



$I_{\mathcal{S}'}$ is a **causal compatibility inequality** for model M'

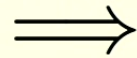


$P_A + P_B + P_C - P_A P_B - P_{BC} - P_{AC} \leq 1$
 is a causal compatibility
 inequality for M



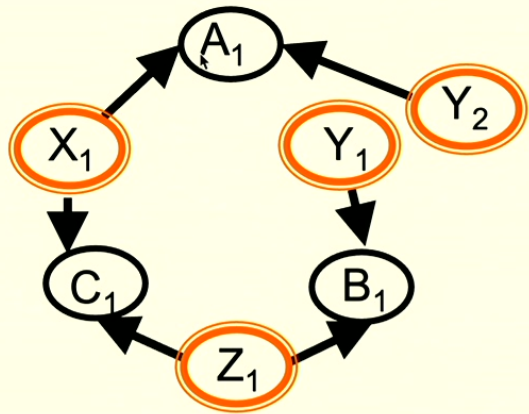
$P_{A_1} + P_{B_1} + P_{C_1} - P_{A_1} P_{B_1} - P_{B_1 C_1} - P_{A_1 C_1} \leq 1$
 is a causal compatibility
 inequality for M'

$(P_{A_1C_1}, P_{B_1C_1}, P_{A_1B_1})$
is a valid set of marginals



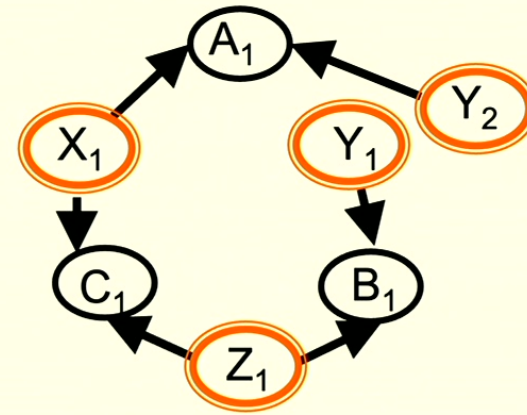
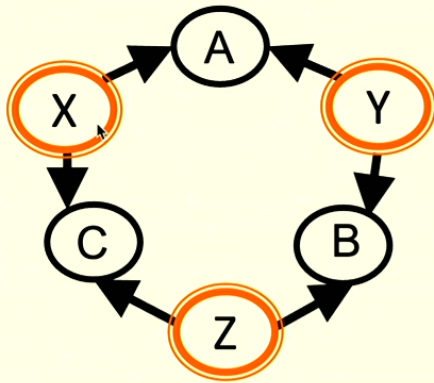
$(P_{A_1C_1}, P_{B_1C_1}, P_{A_1B_1})$ satisfy
 $P_{A_1} + P_{B_1} + P_{C_1} - P_{A_1B_1} - P_{B_1C_1} - P_{A_1C_1} \leq 1$

Linear quantifier elimination



$(P_{A_1C_1}, P_{B_1C_1}, P_{A_1B_1})$
is compatible with M'

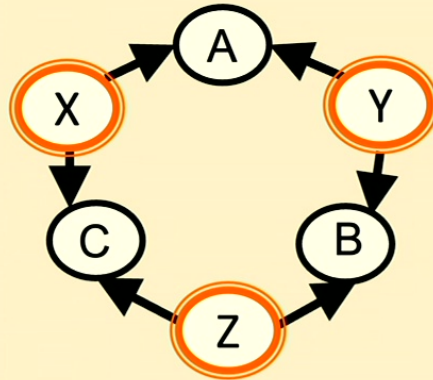
$$A_1 \perp_d B_1 \implies P_{A_1B_1} = P_{A_1}P_{B_1}$$



$P_A + P_B + P_C - P_A P_B - P_{BC} - P_{AC} \leq 1$
 is a causal compatibility
 inequality for M



$P_{A_1} + P_{B_1} + P_{C_1} - P_{A_1} P_{B_1} - P_{B_1 C_1} - P_{A_1 C_1} \leq 1$
 is a causal compatibility
 inequality for M'

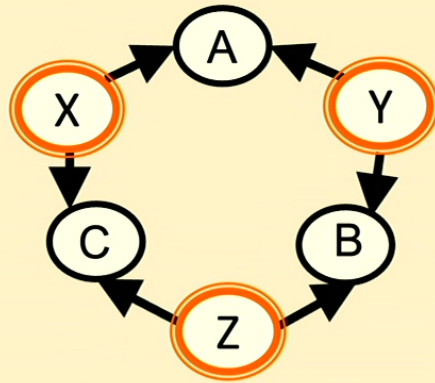


$$P_A + P_B + P_C - P_A P_B - P_{BC} - P_{AC} \leq 1$$

causal compatibility inequality

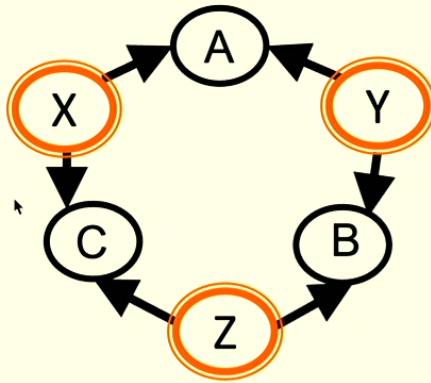
rules out

$$P_{ABC} = \frac{1}{2}[000] + \frac{1}{2}[111]$$



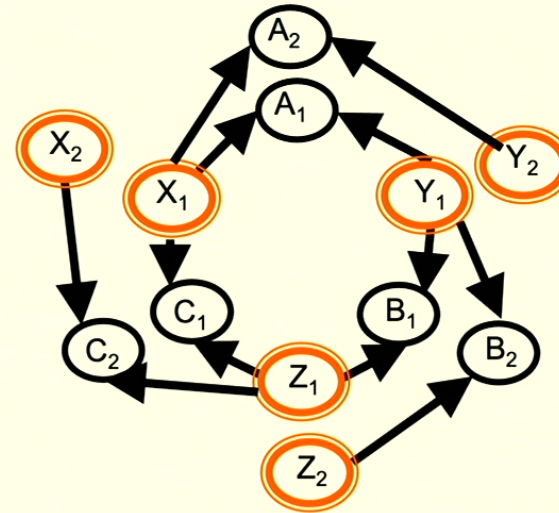
$$\begin{aligned} &P_A(1)P_B(1)P_C(1) \\ &\leq P_{AB}(11)P_C(1) + P_{BC}(11)P_A(1) \\ &\quad + P_{AC}(11)P_B(1) + P_{ABC}(000) \end{aligned}$$

causal compatibility inequality



$$P_A(1)P_B(1)P_C(1) \leq P_{AB}(11)P_C(1) + P_{BC}(11)P_A(1) + P_{AC}(11)P_B(1) + P_{ABC}(000)$$

is a causal compatibility inequality for M

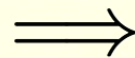


$$P_{A_2}(1)P_{B_2}(1)P_{C_2}(1) \leq P_{A_1B_2}(11)P_{C_2}(1) + P_{B_1C_2}(11)P_{A_2}(1) + P_{A_2C_1}(11)P_{B_2}(1) + P_{A_1B_1C_1}(000)$$

is a causal compatibility inequality for M'

$$(P_{A_2B_2C_2}, P_{A_1B_2C_2}, P_{A_2B_1C_2}, P_{A_2B_2C_1}, P_{A_1B_1C_1})$$

is a valid set of marginals

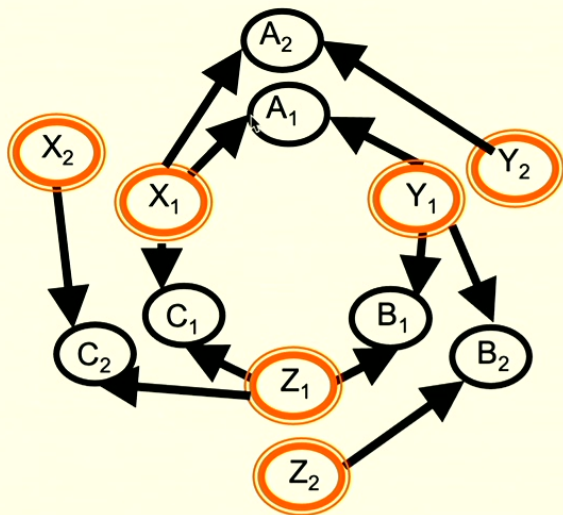


$$P_{A_2B_2C_2}(111)$$

$$\leq P_{A_1B_2C_2}(111) + P_{B_1C_2A_2}(111)$$

$$+ P_{A_2C_1B_2}(111) + P_{A_1B_1C_1}(000)$$

Linear quantifier elimination



$$(P_{A_2B_2C_2}, P_{A_1B_2C_2}, P_{A_2B_1C_2}, P_{A_2B_2C_1}, P_{A_1B_1C_1})$$

is compatible with M'

$$A_1B_2 \perp_d C_2 \implies P_{A_1B_2C_2} = P_{A_1B_2}P_{C_2},$$

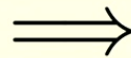
$$B_1C_2 \perp_d A_2 \implies P_{B_1C_2A_2} = P_{B_1C_2}P_{A_2},$$

$$A_2C_1 \perp_d B_2 \implies P_{A_2C_1B_2} = P_{A_2C_1}P_{B_2},$$

$$A_2 \perp_d B_2 \perp_d C_2 \implies P_{A_2B_2C_2} = P_{A_2}P_{B_2}P_{C_2}$$

$$(P_{A_2B_2C_2}, P_{A_1B_2C_2}, P_{A_2B_1C_2}, P_{A_2B_2C_1}, P_{A_1B_1C_1})$$

is compatible with M'



$$P_{A_2}(1)P_{B_2}(1)P_{C_2}(1)$$

$$\leq P_{A_1B_2}(11)P_{C_2}(1) + P_{B_1C_2}(11)P_{A_2}(1)$$

$$+ P_{A_2C_1}(11)P_{B_2}(1) + P_{A_1B_1C_1}(000)$$

Polynomial inequality constraints for causal compatibility with the original DAG



Linear inequality constraints from marginal compatibility
(from linear quantifier elimination)

+

Polynomial equality constraints from causal compatibility with the inflated DAG
(e.g., from d-separation relations)

- ↖ The technique defines an algorithm for deriving causal compatibility inequalities and for testing compatibility

Proof that this provides a convergent hierarchy of tests:
Navascués & Wolfe, J. Causal Inf. 8(1) 70 (2020)

Approaches to Bell arguments that follow essentially
the logic of the inflation technique:

Fine's proof of CHSH inequalities

Hardy's proof of Bell's theorem

The Greenberger-Horne-Zeilinger proof of Bell's theorem

Bell's theorem by way of Kochen-Specker noncontextuality

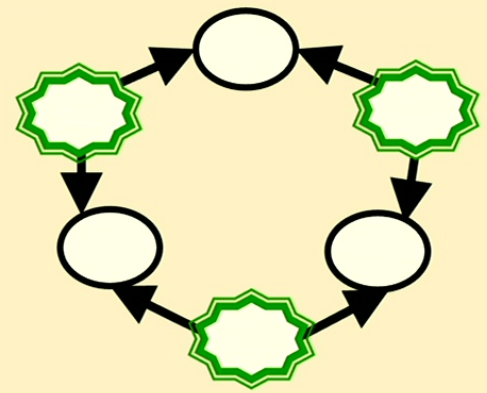
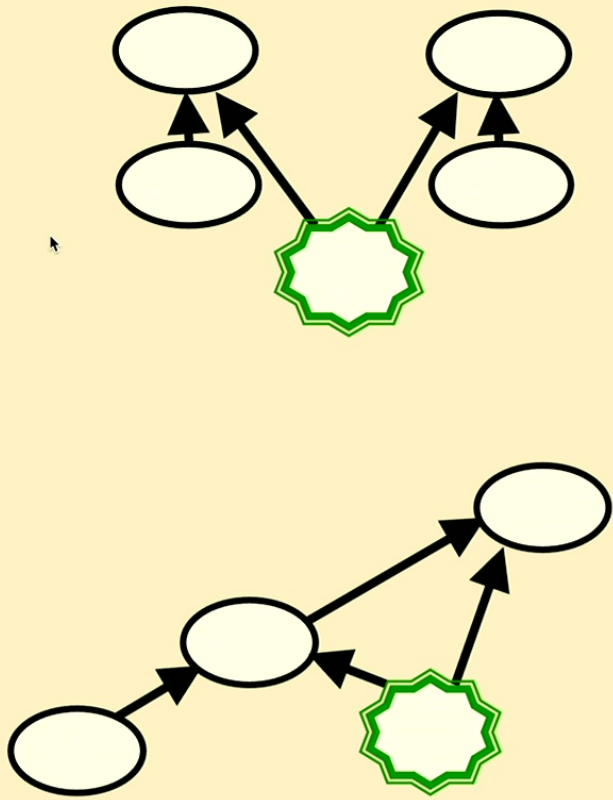
Braunstein-Caves entropic inequalities

Symmetric extensions / shareability of local correlations

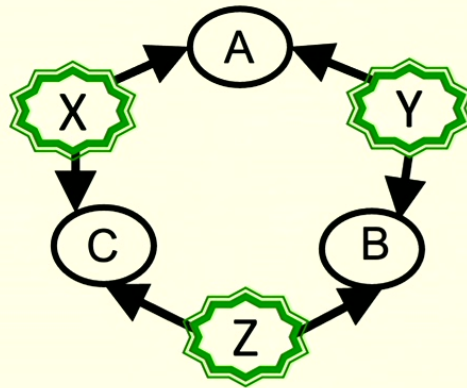
...

What probability distributions over classical variables are compatible with a given causal structure when the latent systems can be quantum?

The analogue of finding the Tsirelson bound for the Bell scenario



Triangle scenario



$$\rho_{A|XY}$$

$$\rho_{B|YZ}$$

$$\rho_{C|XZ}$$

$$\rho_X$$

$$\rho_Y$$

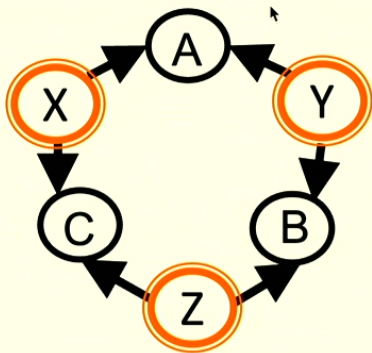
$$\rho_Z$$

$$[\rho_{A|XY}, \rho_{B|YZ}] = 0$$

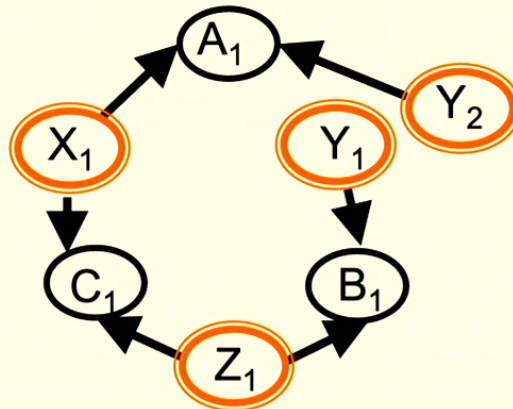
$$[\rho_{B|YZ}, \rho_{C|XZ}] = 0$$

$$[\rho_{A|XY}, \rho_{C|XZ}] = 0$$

$$P_{ABC} = \text{Tr}_{XYZ} (\rho_{A|XY} \rho_{B|YZ} \rho_{C|XZ} \rho_X \rho_Y \rho_Z)$$

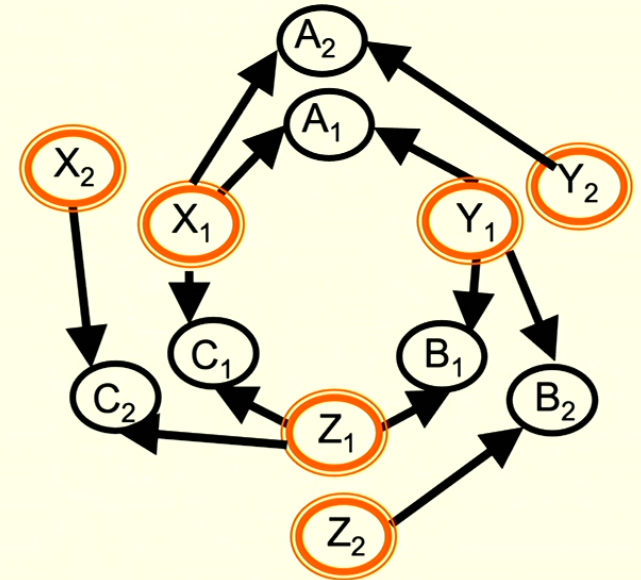


Triangle



Cut inflation of Triangle

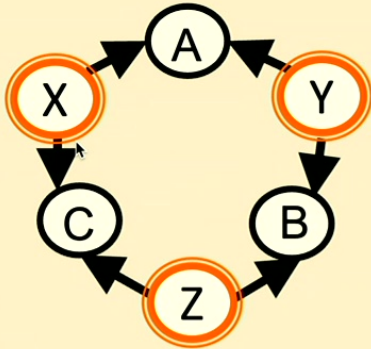
Non-fan-out



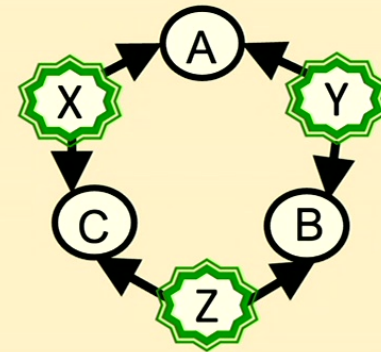
Spiral inflation of Triangle

Fan-out

Classical latents



Quantum latents

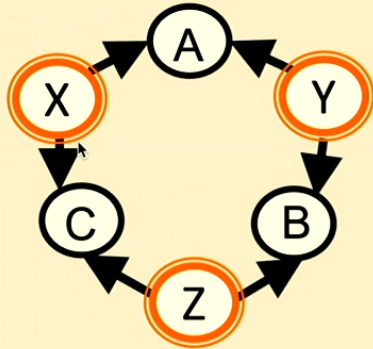


$$P_A + P_B + P_C - P_A P_B - P_{BC} - P_{AC} \leq 1$$

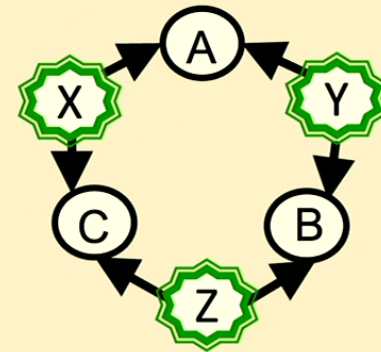
$$P_A + P_B + P_C - P_A P_B - P_{BC} - P_{AC} \leq 1$$

This inequality was obtained from a non-fanout inflation and therefore holds for both quantum and classical latents

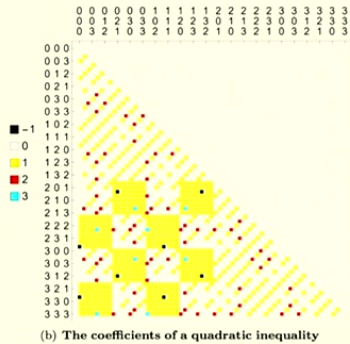
Classical latents



Quantum latents

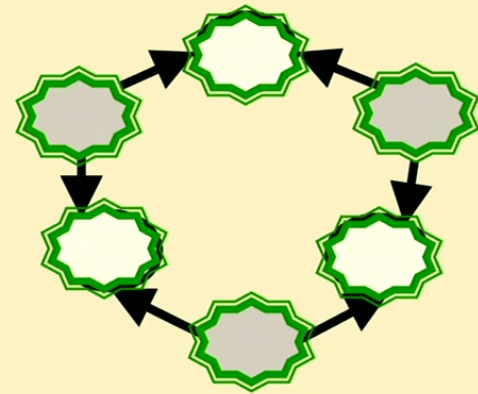
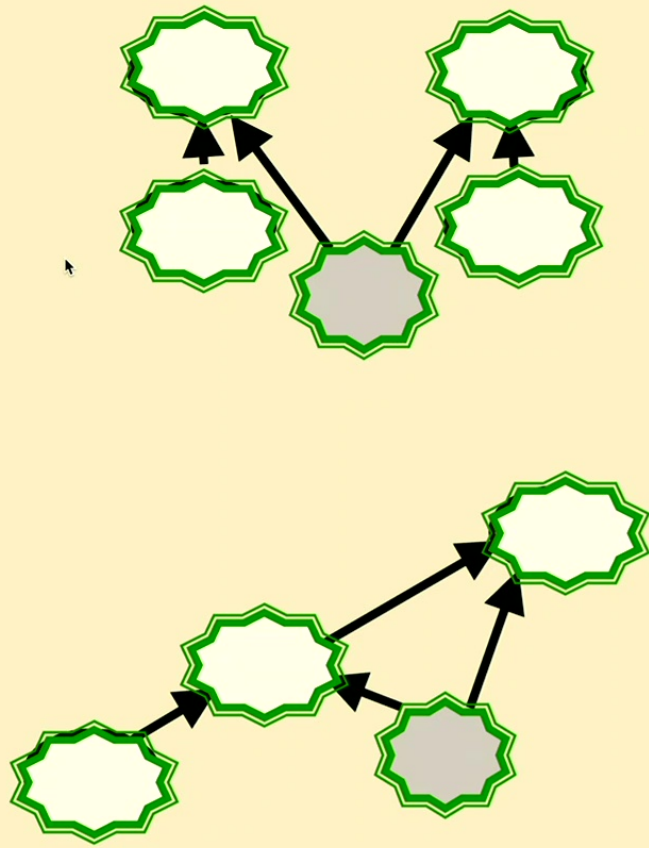


$$\sum_{a,b,c,a',b',c'} y_{abca'b'c'} P_{ABC}(abc) P_{ABC}(a'b'c') \geq 0$$



This inequality can be quantumly violated

Polino et al., Nat. Comm. 14, 909 (2023)



15 hours of lectures
Available online at <https://pirsa.org/c23016>

Causal
Inference:
Classical and
Quantum

PHYS 777-007
Lecturer: Robert Spekkens
TA: Marina Ansanelli

March 6, 2023



Causarum Investigatio
"Investigate the causes"