

MOTIVATION

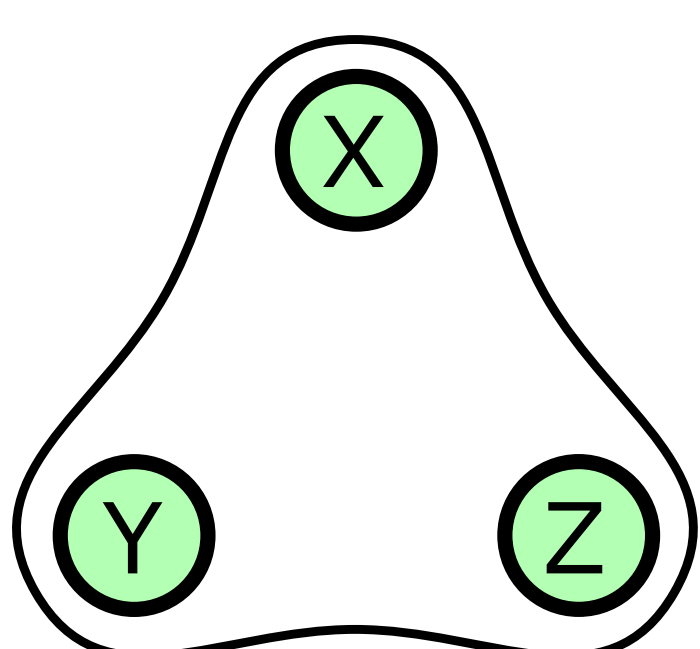
1. Pairwise interactions can be insufficient
2. Relational data is not widely available
3. High-order $\neq \sum$ pairwise

Observations

X_1 Y_1 Z_1
 X_2 Y_2 Z_2
 X_3 Y_3 Z_3
.
.
.
.
.

?

Higher order
interactions



Can we directly find high-order interactions from observational (non-relational) data?

INFORMATION MEASURES

Dependence (2 variables) = interactions.

$$MI(X_1; X_2) = H(X_1) + H(X_2) - H(X_1, X_2)$$

Dependence measure generalised to $d \geq 3$:

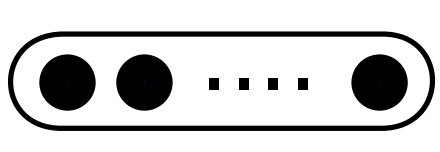
$$TC(d) = \sum_{i=1}^d H(X_i) - H(X_1, \dots, X_d)$$

joint
distribution



$P_{12\dots d}$

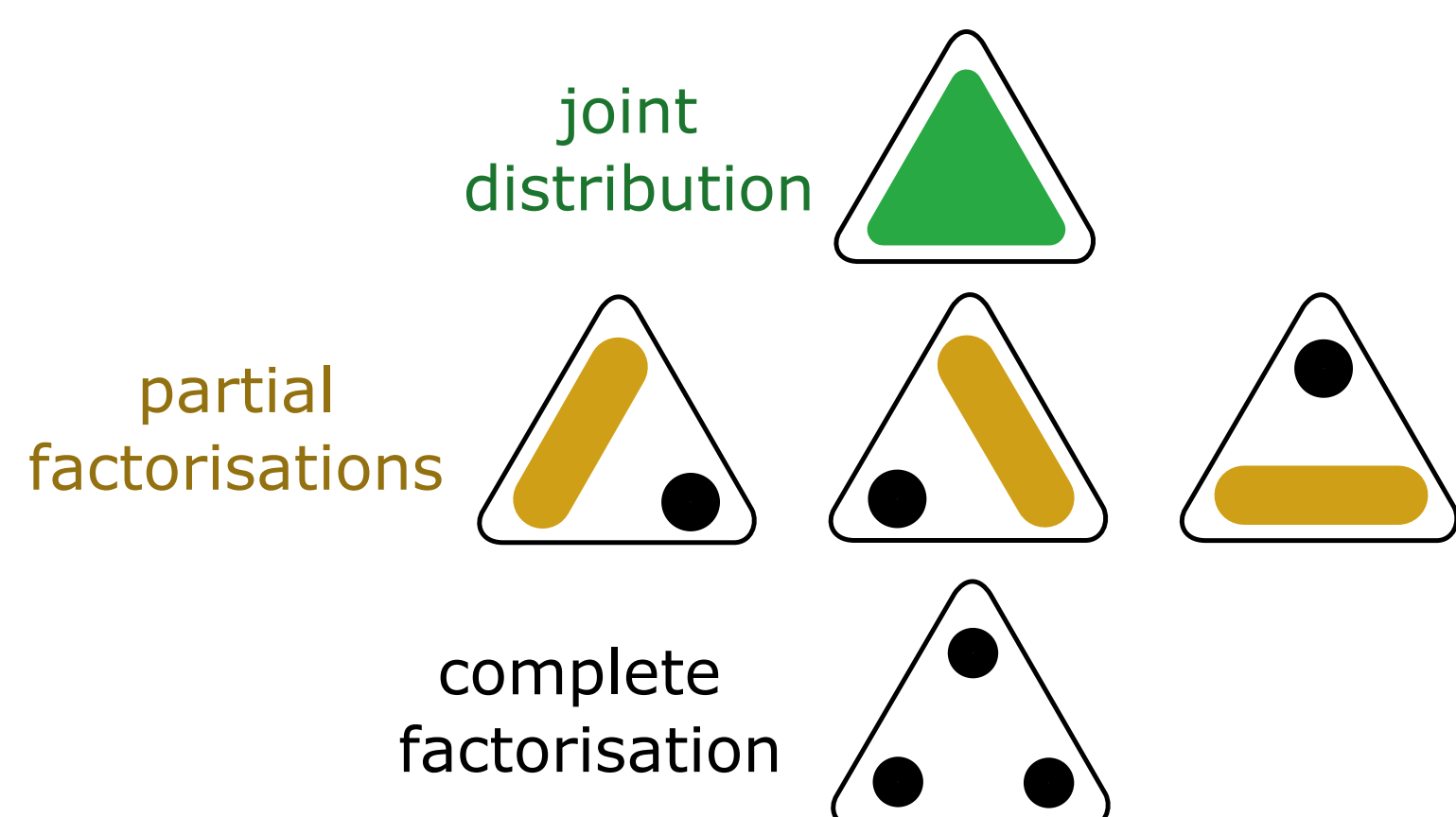
complete
factorisation



$P_1 P_2 \dots P_d$

Non-zero Total Correlation is less informative as d becomes large. Eg, 5 possibilities ($d = 3$):

P_{123} $P_1 P_{23}$ $P_2 P_{13}$ $P_3 P_{12}$ $P_1 P_2 P_3$



Interaction information for 3 variables:

$$II(3) = H(X) + H(Y) + H(Z) + H(X, Y, Z) - H(X, Y) - H(X, Z) - H(Y, Z)$$

LATTICE

Definition 1 Lattice \approx set + order

E.g., Partition lattice = partitions + refinement and its **Möbius inversion** when $d = 3$:

$$P_{123} = \Delta_{123} + \Delta_{12|3} + \Delta_{1|23} + \Delta_{2|13} + \Delta_{1|2|3}$$

$$\Delta_{123} = P_{123} - P_1 P_{23} - P_2 P_{13} - P_3 P_{12} + 2P_1 P_2 P_3$$

Lattice formalisation is beneficial:

1. compare interaction measures
2. generate *new* interaction measures

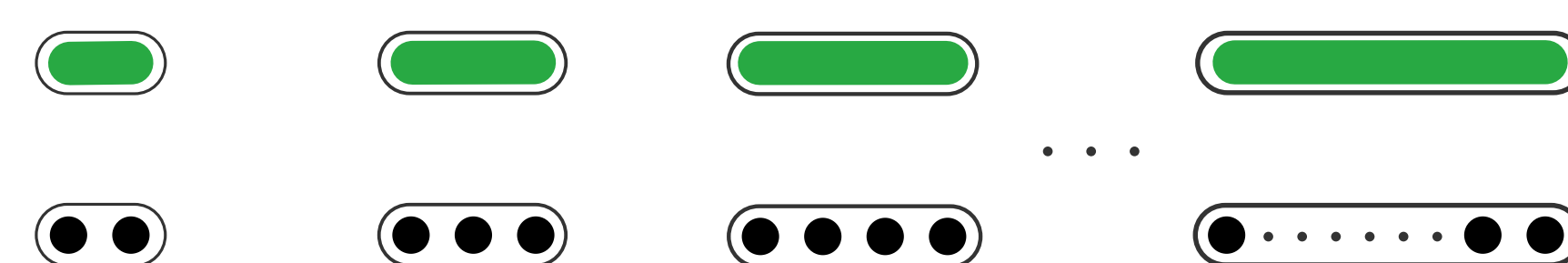
LATTICE EMBEDDINGS

Which lattices are associated with the existing information-theoretic interaction measures?

Two-Element Chain - Simple

MI and TC are generated by the Two-Element Chain C_2 , which are isomorphic for all d .

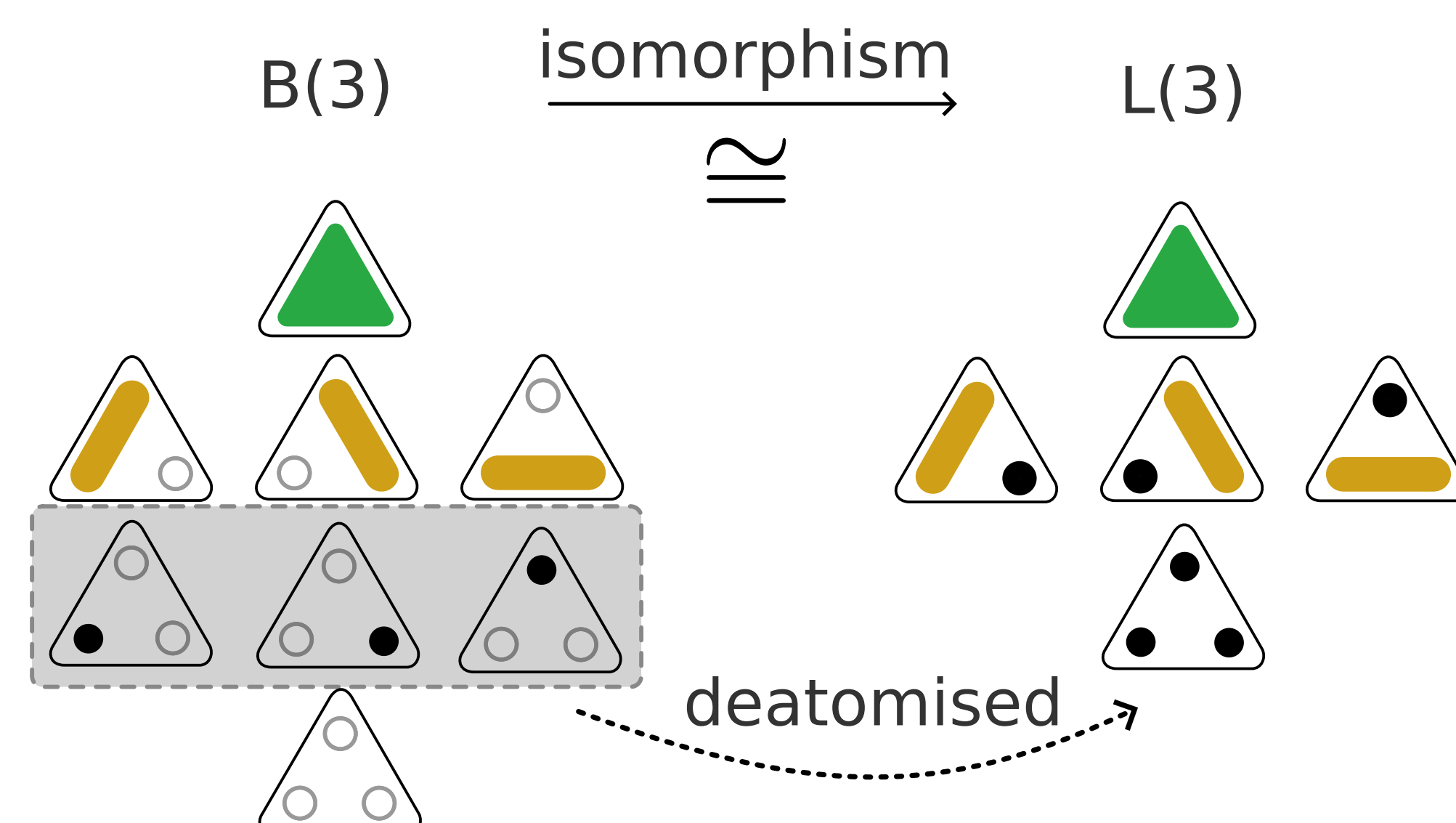
$$C_2(2) \cong C_2(3) \cong C_2(4) \cong \dots \cong C_2(d)$$



Boolean Lattice - Natural

Interaction information can be generated from the Boolean lattice $B(d)$.

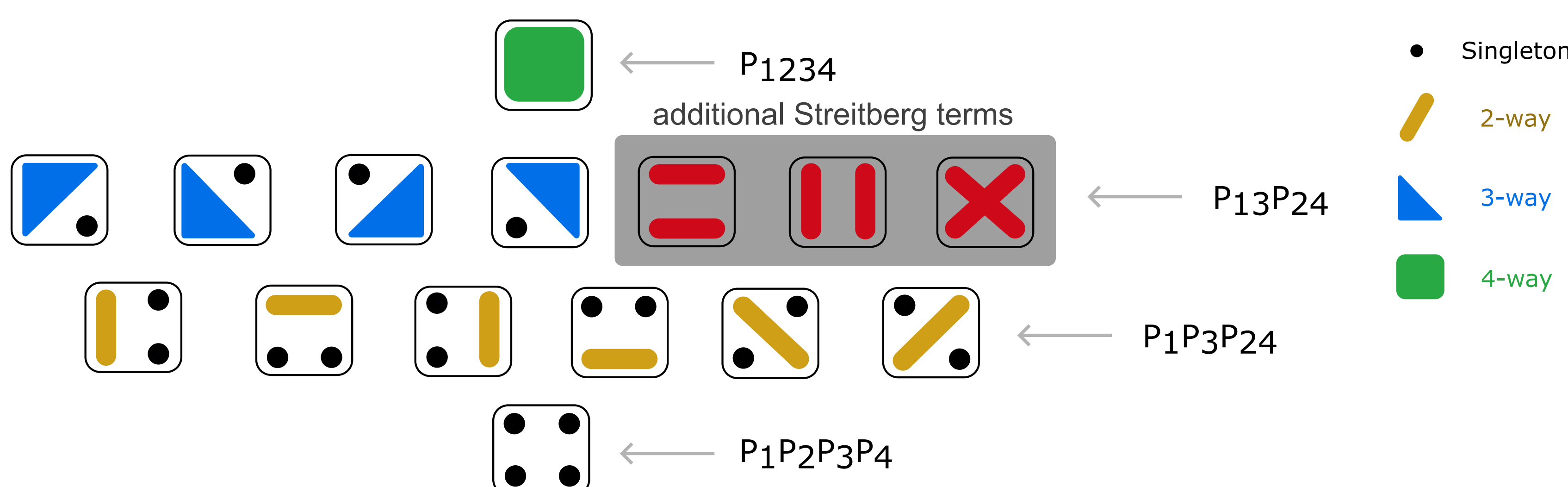
After removing the singletons, $B(d)$ is isomorphic to Lancaster lattice $L(d)$, a sublattice of partition lattice.



Partition Lattice - General

"All finite lattices can be embedded into a finite partition lattice."

This suggests an information measure induced by partition lattice can be more powerful.



STREITBERG INFORMATION

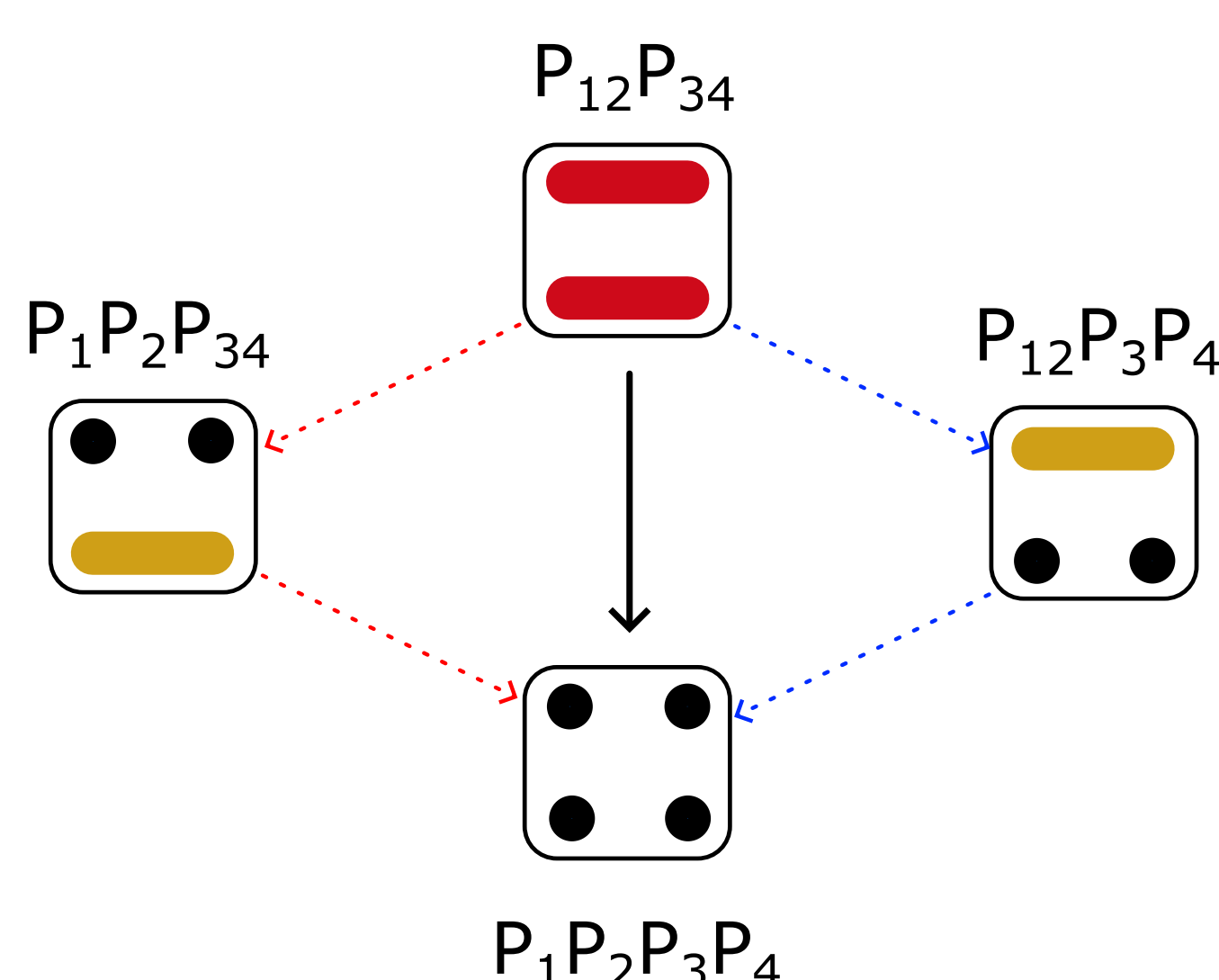
Streitberg interaction from partition lattice:

$$\Delta_S^d P = \sum_{\pi} (-1)^{|\pi|-1} (|\pi| - 1)! P_{\pi}$$

If the joint distribution factorises in any way, then $\Delta_S^d P$ becomes **zero**.

KL formulation

KL-divergence formulation is not effective as it destroys the structure of the lattice. E.g., $D_{KL}(P_{12}P_{34} || P_1 P_2 P_3 P_4) = D_{KL}(P_{34} || P_3 P_4) + D_{KL}(P_{12} || P_1 P_2)$.



Tsallis Alpha Divergence

The Streitberg information $SI(d)$ is defined as:

$$SI(d) = \sum_{\pi} (|\pi| - 1)! (-1)^{|\pi|-1} D_{TA} \left(p_{\pi} || \prod_{i=1}^d p_i \right)$$

PROPERTIES

Through experiments we validate:

1. Clear vanishing condition
2. Recursiveness
3. Monotonicity
4. Symmetry
5. Interpretability
6. Emergence

We provide a consistent estimator using k NN.

APPLICATIONS

Finance

Identify the high-order interactions between the stocks in SP500, which can lead to more **diverse investment portfolio**.

Neuroscience

Outperforms pairwise-based methods at **decoding neural signals** as high-order information is prevalent in brain signals.

Machine Learning

Uncover the **true features** that generate the target variable through high-order interaction whereas standard feature selection methods like SHAP all fail.