

The Undecidability of Probabilistic Conditional Independence Implication

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Random Variables

- A **random variable (RV)** $X : \Omega \rightarrow \mathcal{X}$ is a measurable function from a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ to a measurable space
- We focus on **discrete** random variables, i.e., the support \mathcal{X} is finite or countable

- Suffices to consider $\Omega = [0, 1]$ to be the standard probability space, i.e., $[0, 1]$ with the Lebesgue measure as the probability

- X, Y are (unconditionally) independent, denoted as $X \perp\!\!\!\perp Y$, if for all x, y ,

$$\mathbb{P}((X, Y) = (x, y)) = \mathbb{P}(X = x)\mathbb{P}(Y = y)$$

- X, Y are conditionally independent given Z , denoted as $X \perp\!\!\!\perp Y | Z$, if for all x, y, z ,

$$\mathbb{P}((X, Y, Z) = (x, y, z))\mathbb{P}(Z = z) = \mathbb{P}((X, Z) = (x, z))\mathbb{P}((Y, Z) = (y, z))$$

- WLOG assume all random variables are positive-integer-valued, i.e., measurable functions $X : [0, 1] \rightarrow \mathbb{N}$

First-order Theory of Random Variables

- Consider first-order formulae (with logical symbols $\forall, \exists, \wedge, \vee, \neg$), with non-logical symbols $\cdot \perp\!\!\!\perp \cdot$ (unconditional independence) and $\cdot \perp\!\!\!\perp \cdot | \cdot$ (conditional independence)
- Variables in the formulae are random variables, i.e., measurable functions $[0, 1] \rightarrow \mathbb{N}$
 - Just the ordinary first-order logic over the domain of measurable functions $[0, 1] \rightarrow \mathbb{N}$, with the usual semantics
- Relation with probabilistic team semantics [Durand et al., 2018, Hannula et al., 2023]:
 - A probabilistic team \mathbb{X} can be regarded as a joint distribution of the variables
 - Conditional independence $\mathfrak{A} \models_{\mathbb{X}} x \perp\!\!\!\perp_z y$ means $X \perp\!\!\!\perp Y | Z$ as RVs
 - Different semantics for \vee and \forall

Undecidable Problems

- Undecidable problems are decision problems that cannot be solved by any algorithm
 - E.g., Halting problem [Turing, 1936], Diophantine equations [Matiyasevich, 1993], Wang tiles [Berger, 1966], word problem of groups [Novikov, 1955]
- We discuss the undecidability of:
 - Conditional independence implication problem
 - First-order theory of random variables with probabilistic independence relation
 - Conditional information inequalities
 - Network coding

Probabilistic Independence Implication Problem

- Determine whether a probabilistic independence relation among several random variables follows from a list of other such relations [Geiger et al., 1991, Matúš, 1994]
- E.g. $X \perp\!\!\!\perp Y \wedge XY \perp\!\!\!\perp Z \Rightarrow X \perp\!\!\!\perp YZ$
 - i.e., $\forall X, Y, Z. ((X \perp\!\!\!\perp Y \wedge XY \perp\!\!\!\perp Z) \rightarrow X \perp\!\!\!\perp YZ)$
 - In the language of probabilistic team semantics:
 $\mathfrak{A} \models_{\mathbb{X}} (x \perp\!\!\!\perp y \wedge xy \perp\!\!\!\perp z) \Rightarrow \mathfrak{A} \models_{\mathbb{X}} x \perp\!\!\!\perp yz$
- Geiger et al. [1991] gave a complete set of axioms:
 - (Triviality) $X \perp\!\!\!\perp \emptyset$
 - (Symmetry) $X \perp\!\!\!\perp Y \Rightarrow Y \perp\!\!\!\perp X$
 - (Decomposition) $X \perp\!\!\!\perp YZ \Rightarrow X \perp\!\!\!\perp Y$
 - (Mixing) $X \perp\!\!\!\perp Y \wedge XY \perp\!\!\!\perp Z \Rightarrow X \perp\!\!\!\perp YZ$
- Complete – all true probabilistic independence implications can be deduced from these axioms
- Hence probabilistic independence implication is **decidable**

Conditional Independence Implication Problem

- Determine whether a conditional independence relation among several random variables follows from a list of other such relations [Dawid, 1979, Spohn, 1980, Mouchart and Rolin, 1984]
- E.g. $X \perp\!\!\!\perp Y|Z \wedge X \perp\!\!\!\perp W|YZ \Rightarrow X \perp\!\!\!\perp W|Z$
- **Decidable** if all random variables have bounded cardinalities [Geiger and Meek, 1999, Niepert, 2012]
 - Follows from the decidability of the real polynomial equations
 - Hannula et al. [2019] – in EXPSPACE if all RVs are binary
- What about the case where the cardinalities of the random variables are not bounded?

Semi-graphoid Axioms

- Pearl and Paz [1987] proposed the following 4 axioms:
 - (Symmetry) $X \perp\!\!\!\perp Y|Z \Rightarrow Y \perp\!\!\!\perp X|Z$
 - (Decomposition) $X \perp\!\!\!\perp YW|Z \Rightarrow X \perp\!\!\!\perp Y|Z$
 - (Weak union) $X \perp\!\!\!\perp YW|Z \Rightarrow X \perp\!\!\!\perp Y|ZW$
 - (Contraction) $X \perp\!\!\!\perp Y|Z \wedge X \perp\!\!\!\perp W|YZ \Rightarrow X \perp\!\!\!\perp YW|Z$
- CI implication would be decidable if semi-graphoid axioms are complete (i.e., all true CI implications can be deduced from these axioms)
 - Simply apply the axioms repeatedly on every combination of random variables until we obtain the desired CI statement
- For the special case where every CI statement involves all random variables (saturated CI), semi-graphoid axioms are complete, and hence **decidable** [Malvestuto, 1992, Geiger and Pearl, 1993]
- Unfortunately, semi-graphoid axioms are incomplete [Studený, 1989]
- Is conditional independence implication decidable in general?

Undecidability of Conditional Independence Implication

- Studený [1989]: Semi-graphoid axioms [Pearl and Paz, 1987] are incomplete
 - Is it possible to add more axioms to make it complete?
- Studený [1992]: No, conditional independence has no finite axiomization
 - Does not rule out other kinds of algorithms
- Herrmann [1995]: Embedded multivalued database dependency is undecidable
- Li [2021]: CI implication is undecidable if one of the RVs is binary
- Li [2022a]: First-order theory of random variables with probabilistic independence relation is undecidable
 - Allow any combination of $\perp\!\!\!\perp, \forall, \exists, \wedge, \vee, \neg$, not only implication
- Li [2022b]: CI implication is **undecidable**
 - Uses the ideas of Herrmann [1995]
- Kühne and Yashfe [2022]: Another concurrent proof of undecidability via matroid theory

First-order Theory of Probabilistic Independence

- Consider first-order formulae with only one non-logical symbol $\perp\!\!\!\perp$ (probabilistic independence)
 - Variables are random variables (X, Y, \dots)
- How to define condition that X is constant, written as $X \stackrel{\perp}{=} \emptyset$?
 - $X \stackrel{\perp}{=} \emptyset \Leftrightarrow X \perp\!\!\!\perp X$
- How to define relation that X is a function of Y , written as $X \stackrel{\leq}{\leq} Y$?
 - $X \stackrel{\leq}{\leq} Y \Leftrightarrow \forall U. (U \perp\!\!\!\perp Y \rightarrow U \perp\!\!\!\perp X)$
 - Write $X \stackrel{=}{=} Y \Leftrightarrow X \stackrel{\leq}{\leq} Y \wedge Y \stackrel{\leq}{\leq} X$ and
 $X \stackrel{<}{<} Y \Leftrightarrow X \stackrel{\leq}{\leq} Y \wedge \neg(Y \stackrel{\leq}{\leq} X)$
- How to define the joint random variable of X, Y , written as XY ?
 - $Z \stackrel{=}{=} XY \Leftrightarrow X \stackrel{\leq}{\leq} Z \wedge Y \stackrel{\leq}{\leq} Z \wedge \forall U. ((X \stackrel{\leq}{\leq} U \wedge Y \stackrel{\leq}{\leq} U) \rightarrow Z \stackrel{\leq}{\leq} U)$
- How to define conditional independence, written as $X \perp\!\!\!\perp Y|Z$?
 - $X \perp\!\!\!\perp Y|Z \Leftrightarrow \exists U. U \perp\!\!\!\perp XZ \wedge Y \stackrel{\leq}{\leq} ZU$

- Check X is (at most) a binary random variable (i.e., $|\mathcal{X}| \leq 2$):

$$\text{card}_{\leq 2}(X) \Leftrightarrow \forall U (U \stackrel{I}{<} X \rightarrow U \stackrel{I}{=} \emptyset)$$

- Any random variable with strictly less information than X is degenerate
- The condition that $|\mathcal{X}| \leq n$:

$$\text{card}_{\leq n}(X) \Leftrightarrow \forall U (U \stackrel{I}{<} X \rightarrow \text{card}_{\leq n-1}(U))$$

$$\text{card}_{\leq 1}(X) \Leftrightarrow (X \stackrel{I}{=} \emptyset)$$

- Define

$$\text{card}_{=n}(X) \Leftrightarrow \text{card}_{\leq n}(X) \wedge \neg \text{card}_{\leq n-1}(X)$$

$$\text{card}_{\geq n}(X) \Leftrightarrow \neg \text{card}_{\leq n-1}(X)$$

- If X, Y, Z are discrete random variables such that any one of them is a function of the other two, and they are pairwise independent, then they are all uniformly distributed over their supports, which have the same size [Zhang and Yeung, 1997]
- The condition that X is uniformly distributed over its support:

$$\text{unif}(X) \Leftrightarrow \exists Y, Z. \text{triple}(X, Y, Z),$$

where

$$\begin{aligned} \text{triple}(X, Y, Z) \Leftrightarrow & X \stackrel{!}{\leq} YZ \wedge Y \stackrel{!}{\leq} XZ \wedge Z \stackrel{!}{\leq} XY \\ & \wedge X \perp\!\!\!\perp Y \wedge X \perp\!\!\!\perp Z \wedge Y \perp\!\!\!\perp Z \end{aligned}$$

- Satisfied when $X, Y \sim \text{Unif}\{0, \dots, k-1\}$, $Z = X + Y \bmod k$

Representation of Integers

- Represent $k \in \mathbb{Z}_{>0}$ as a uniform random variable X with $|\mathcal{X}| = k$
- **Equality.** Formula for checking $|\mathcal{X}| = |\mathcal{Y}|$ for uniform X, Y [Li, 2021]:

$$\text{ueq}(X, Y) \Leftrightarrow \exists U_1, U_2, U_3. \\ \text{triple}(X, U_1, U_2) \wedge \text{triple}(Y, U_1, U_3)$$

- To check for equality against constants:

$$\text{ueq}_n(X) \Leftrightarrow \text{unif}(X) \wedge \text{card}_{=n}(X)$$

- **Multiplication.** Formula for $|\mathcal{X}||\mathcal{Y}| = |\mathcal{Z}|$ for uniform X, Y, Z :

$$\text{uprod}(X, Y, Z) \Leftrightarrow \exists \tilde{X}, \tilde{Y}. (\text{ueq}(X, \tilde{X}) \wedge \text{ueq}(Y, \tilde{Y}) \\ \wedge \tilde{X} \perp\!\!\!\perp \tilde{Y} \wedge \tilde{X}\tilde{Y} \stackrel{!}{=} Z)$$

- **Comparison.** Formula for $|\mathcal{X}| \leq |\mathcal{Y}|$ for uniform X, Y [Li, 2021]:

$$\text{ule}(X, Y) \Leftrightarrow \exists G, \tilde{Y}. (\text{uprod}(X, Y, G) \wedge \text{ueq}(Y, \tilde{Y}) \wedge G \stackrel{!}{\leq} Y\tilde{Y})$$

- “ \Leftarrow ”: $G \stackrel{!}{\leq} Y\tilde{Y} \Rightarrow |\mathcal{G}| \leq |\mathcal{Y}||\tilde{\mathcal{Y}}| \Rightarrow |\mathcal{X}||\mathcal{Y}| \leq |\mathcal{Y}|^2$
- “ \Rightarrow ”: $X \sim \text{Unif}\{0, \dots, a-1\}$, $Y \sim \text{Unif}\{0, \dots, b-1\}$, $G = (X, Y)$, $\tilde{Y} = X + Y \bmod b$

Addition between Integers

- To define addition, the main idea is that if Z is uniform with $|\mathcal{Z}| = |\mathcal{X}| + |\mathcal{Y}|$, then we can partition \mathcal{Z} into two sets with sizes $|\mathcal{X}|, |\mathcal{Y}|$ respectively
 - If $U \in \{0, 1\}$ is the indicator of whether Z is in the first set, then $U \sim \text{Bern}(|\mathcal{X}|/(|\mathcal{X}| + |\mathcal{Y}|))$
- The following checks that X, Y, Z are uniform, $|\mathcal{Z}| = |\mathcal{X}| + |\mathcal{Y}|$, and $U \sim \text{Bern}(|\mathcal{X}|/(|\mathcal{X}| + |\mathcal{Y}|))$:

$$\begin{aligned} \text{frac}(X, Y, Z, U) \Leftrightarrow & (\text{ueq}_2(U) \wedge \text{uprod}(X, U, Z) \wedge \text{uprod}(Y, U, Z)) \\ & \vee \exists \tilde{X}, \tilde{Y}. (\text{ueq}(X, \tilde{X}) \wedge \text{ueq}(Y, \tilde{Y}) \wedge \text{unif}(Z) \\ & \wedge \text{card}_{=2}(U) \wedge \neg \text{unif}(U) \wedge U \stackrel{\iota}{\leq} Z \wedge \tilde{X} \perp\!\!\!\perp \tilde{Y} \perp\!\!\!\perp U \wedge Z \stackrel{\iota}{\leq} \tilde{X} \tilde{Y} U \\ & \wedge \forall V. (\text{smi}(Z, V) \rightarrow \text{smi}(\tilde{X}U, V) \vee \text{smi}(\tilde{Y}U, V))) \end{aligned}$$

where

$$\begin{aligned} \text{smi}(X, Y) \Leftrightarrow & (X \stackrel{\iota}{=} Y \stackrel{\iota}{=} \emptyset) \vee (Y \stackrel{\iota}{\leq} X \wedge \text{card}_{=2}(Y) \\ & \wedge \forall U. (U \stackrel{\iota}{\leq} X \wedge \text{card}_{=4}(U) \rightarrow \neg \exists V. (\text{card}_{\leq 2}(V) \wedge U \stackrel{\iota}{\leq} YV))) \end{aligned}$$

- We then have $\text{usum}(X, Y, Z) \Leftrightarrow \exists U. \text{frac}(X, Y, Z, U)$

Theorem (Li [2022a])

The first-order theory of probabilistic independence is undecidable, i.e., no algorithm can determine whether a statement in FOTPI holds

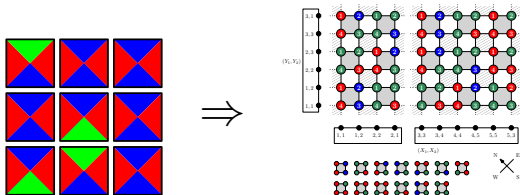
- Direct consequence of the fact that true arithmetic (over natural numbers) is interpretable in FOTPI, and that true arithmetic is undecidable [Tarski, 1933]

Undecidability of CI when one RV is Binary [Li, 2021]

- It is **undecidable** to determine whether

$$|\mathcal{X}_1| \leq 2 \wedge \bigwedge_{i=1}^k X_{A_i} \perp\!\!\!\perp X_{B_i} | X_{C_i} \Rightarrow X_{A_0} \perp\!\!\!\perp X_{B_0} | X_{C_0}$$

- Use $\text{unif}(X_1)$ to force X_1 to be uniform, and make independent copies
- Use comparison to force any RV to have any cardinality
 - E.g. $a = 5$ is the only solution to $2^9 \leq a^4 \leq 2^{10}$
- Reduction from periodic tiling problem [Gurevich and Koryakov, 1972]: deciding whether a set of square tiles can tile a torus
- Use uniform RVs to represent coordinates and colors



- It is **undecidable** to determine whether

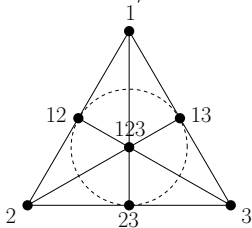
$$\bigwedge_{i=1}^k X_{A_i} \perp\!\!\!\perp X_{B_i} | X_{C_i} \Rightarrow X_{A_0} \perp\!\!\!\perp X_{B_0} | X_{C_0}$$

for given $(A_i)_i, (B_i)_i, (C_i)_i$

- Use the strategy in undecidability of embedded multivalued dependency [Herrmann, 1995]
- Show undecidability by reduction from uniform word problem for finite monoids [Gurevich, 1966]
- Problem – there is no algebraic structure in the RVs X_i !
- Have to impose some algebraic structure using conditional independence

Undecidability of CI Implication [Li, 2022b]

- RVs $A_1, A_2, A_3, A_{12}, A_{13}, A_{23}, A_{123}$
- Impose the “Fano-non-Fano condition”:
 - For any three RVs on same solid line, any one is a function of other two
 - Any three RVs not on same solid/dotted line are independent



Lemma

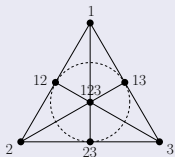
Fano-non-Fano condition holds iff A_1, A_2, A_3 are uniform elements in abelian group, and $A_{12} = A_1 + A_2$, $A_{13} = A_1 + A_3$, $A_{23} = A_2 + A_3$, $A_{123} = A_1 + A_2 + A_3$, up to relabeling

- Equivalent form used in [Herrmann, 1995] for undecidability of EMVD
- Appeared in [Dougherty et al., 2006a] to show unachievability of network coding capacity

Fano-non-Fano Condition

Lemma

Fano-non-Fano condition holds iff A_1, A_2, A_3 are uniform elements in abelian group, $A_{12} = A_1 + A_2$, $A_{13} = A_1 + A_3$, $A_{23} = A_2 + A_3$, $A_{123} = A_1 + A_2 + A_3$, up to relabeling



- A_k is a function of A_i, A_j , let this function be $f_k^{i,j}(a_i, a_j)$
- Bijection between independent (A_i, A_j, A_k) and $(A_i)_i$, let function from (A_i, A_j, A_k) to A_l be $f_l^{i,j,k}(a_i, a_j, a_k)$

Lemma

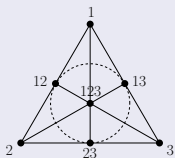
We have

- $f_k^{i,j}(a, b) = f_k^{j,i}(b, a)$, and $f_l^{i,j,k}(a, b, c) = f_l^{j,k,i}(b, c, a)$
- $f_i^{k,j}(f_k^{i,j}(a, b), b) = a$
- $f_l^{i,j,k}(a, b, c) = f_l^{m,k}(f_m^{i,j}(a, b), c)$

Fano-non-Fano Condition

Lemma

Fano-non-Fano condition holds iff A_1, A_2, A_3 are uniform elements in abelian group, $A_{12} = A_1 + A_2$, $A_{13} = A_1 + A_3$, $A_{23} = A_2 + A_3$, $A_{123} = A_1 + A_2 + A_3$, up to relabeling



- A_k is a function of A_i, A_j , let this function be $f_k^{i,j}(a_i, a_j)$
- Bijection between independent (A_i, A_j, A_k) and $(A_i)_i$, let function from (A_i, A_j, A_k) to A_l be $f_l^{i,j,k}(a_i, a_j, a_k)$
- $f_{12}^{1,2} = f_{12}^{2,1} = f_{13}^{1,3} = f_{13}^{3,1} = f_{23}^{2,3} = f_{23}^{3,2} = f_{123}^{1,23} = f_{123}^{2,13} = f_{123}^{3,12} = f_{123}^{23,1} = f_{123}^{13,2} = f_{123}^{12,3}$
- Define abelian group over \mathcal{A} by $a + b := f_{12}^{1,2}(a, b)$, $-a := f_2^{1,12}(a, 0)$

Undecidability of CI Implication

- Strategy proposed by Dougherty [2009] – reduction from the identity problem for finite groups
 - Identity – equality that holds **for all** values of the variables, e.g.,
 $\forall x, y. xy = yx$ (iff group is abelian)
 - Identity problem – whether a list of identities implies another identity

$$\bigwedge_{i=1}^l (\forall x_{1..k}. P_i(x_{1..k})) \rightarrow \forall x_{1..k}. P_0(x_{1..k})$$

- Uniform RVs act as the universally-quantified variables
 - However, identity problem for finite groups is not known to be decidable or undecidable [Albert et al., 1992]!
- Herrmann [1995], Li [2022b]: instead use uniform word problem for finite monoids [Gurevich, 1966]
 - Whether a list of equalities implies another equality

$$\forall x_{1..k}. \left(\bigwedge_{i=1}^l P_i(x_{1..k}) \rightarrow P_0(x_{1..k}) \right)$$

- Need to use uniform RVs to represent specific monoid elements

Word Problem and Endomorphism Monoid

- Uniform word problem for finite monoids [Gurevich, 1966] – Given $a_i, b_i, c_i \in \{1, \dots, k\}$ for $i = 0, \dots, l$, determine whether the implication

$$\bigwedge_{i=1}^l (x_{a_i} \cdot x_{b_i} = x_{c_i}) \rightarrow (x_{a_0} = x_{c_0})$$

holds for all finite monoids \mathcal{M} and all k -tuples $x_1, \dots, x_k \in \mathcal{M}$

- Consider endomorphism monoid of abelian group
 - Homomorphism $g : \mathcal{A} \rightarrow \mathcal{B}$ between abelian groups \mathcal{A}, \mathcal{B} is a function satisfying $g(a + b) = g(a) + g(b)$
 - Endomorphism in \mathcal{A} is a homomorphism $g : \mathcal{A} \rightarrow \mathcal{A}$
 - The *endomorphism monoid* $\text{End}(\mathcal{A})$ is the set of endomorphisms in \mathcal{A} , equipped with the operation $g \cdot h : \mathcal{A} \rightarrow \mathcal{A}$ where $g \cdot h(a) = g(h(a))$
- Kurosh [1963] – For any finite monoid, there exists embedding from that monoid into $\text{End}(\mathcal{A})$ for some finite abelian group \mathcal{A}
 - No loss of generality of considering only endomorphism monoids

Representing Endomorphism by RV

- A_1, A_2, A_3 are uniform elements in abelian group \mathcal{A} , and $A_{12} = A_1 + A_2$, $A_{13} = A_1 + A_3$, $A_{23} = A_2 + A_3$, $A_{123} = A_1 + A_2 + A_3$
- Represent an endomorphism $g : \mathcal{A} \rightarrow \mathcal{A}$ by $U = A_1 - g(A_2)$
- Check whether U corresponds to an endomorphism [Herrmann, 1995, Li, 2023]:

$$\begin{aligned} \text{end}_{1,2}((A_i)_i, U) &\Leftrightarrow \exists V, W : \text{FanoNonFano}((A_i)_i) \wedge \text{ueq}(U, A_1) \\ &\quad \wedge \text{ueq}(V, A_1) \wedge \text{ueq}(W, A_1) \wedge U \stackrel{\ell}{=} A_1|A_2 \\ &\quad \wedge V \stackrel{\ell}{=} A_1|A_{23} \wedge U \stackrel{\ell}{=} V|A_3 \wedge W \stackrel{\ell}{=} A_{13}|A_2 \wedge U \stackrel{\ell}{=} W|A_3, \end{aligned}$$

where $X \stackrel{\ell}{=} Y|Z \Leftrightarrow X \stackrel{\ell}{\leq} ZY \wedge Y \stackrel{\ell}{\leq} ZX$, i.e., if we are given Z , then X has the same information as Y , and $\text{ueq}(X, Y)$ checks whether X, Y are both uniform and have the same cardinality

- “ \Rightarrow ”: $V = A_1 - g(A_2 + A_3)$, $W = A_1 - g(A_2) + A_3$
- Representing composition – if $\text{end}_{1,2}((A_i)_i, U_1)$, $\text{end}_{2,3}((A_i)_i, U_2)$, $\text{end}_{1,3}((A_i)_i, U_3)$, we have $U_3 \stackrel{\ell}{\leq} U_1 U_2$ iff $g_3 = g_1 \cdot g_2$
 - “ \Leftarrow ”: $U_3 = A_1 - g_1 \cdot g_2(A_3) = A_1 - g_1(A_2) + g_1(A_2 - g_2(A_3))$

Representing Endomorphism by RV

- A_1, A_2, A_3 are uniform elements in abelian group \mathcal{A} , and $A_{12} = A_1 + A_2$, $A_{13} = A_1 + A_3$, $A_{23} = A_2 + A_3$, $A_{123} = A_1 + A_2 + A_3$
- Represent an endomorphism $g : \mathcal{A} \rightarrow \mathcal{A}$ by $U = A_1 - g(A_2)$
- Check whether U corresponds to an endomorphism [Herrmann, 1995, Li, 2023]: $\text{end}_{1,2}((A_i)_{i \in \mathcal{E}}, U)$
- Representing composition – if $\text{end}_{1,2}((A_i)_i, U_1)$, $\text{end}_{2,3}((A_i)_i, U_2)$, $\text{end}_{1,3}((A_i)_i, U_3)$, we have $U_3 \stackrel{\ell}{\leq} U_1 U_2$ iff $g_3 = g_1 \cdot g_2$
- Need to convert $\text{end}_{2,3}, \text{end}_{1,3}$ to $\text{end}_{1,2}$
- Convert $\text{end}_{i,j}$ for different i, j :

$$\begin{aligned} \text{conv}_{1,3}^{1,2}((A_i)_i, U, V) &\Leftrightarrow \exists W : \text{end}_{1,2}((A_i)_i, U) \\ &\quad \wedge \text{end}_{1,3}((A_i)_i, V) \wedge \text{end}_{2,3}((A_i)_i, W) \\ &\quad \wedge A_{13} \stackrel{\ell}{\leq} A_{12} W \wedge V \stackrel{\ell}{\leq} U W \end{aligned}$$

Representing Endomorphism by RV

- A_1, A_2, A_3 are uniform elements in abelian group \mathcal{A} , and $A_{12} = A_1 + A_2$, $A_{13} = A_1 + A_3$, $A_{23} = A_2 + A_3$, $A_{123} = A_1 + A_2 + A_3$
- Represent an endomorphism $g : \mathcal{A} \rightarrow \mathcal{A}$ by $U = A_1 - g(A_2)$
- Check whether U corresponds to an endomorphism [Herrmann, 1995, Li, 2023]: $\text{end}_{1,2}((A_i)_{i \in \mathcal{E}}, U)$
- Representing composition – if $\text{end}_{1,2}((A_i)_i, U_1)$, $\text{end}_{2,3}((A_i)_i, U_2)$, $\text{end}_{1,3}((A_i)_i, U_3)$, we have $U_3 \stackrel{t}{\leq} U_1 U_2$ iff $g_3 = g_1 \cdot g_2$
- Convert $\text{end}_{i,j}$ for different i, j : $\text{conv}_{1,3}^{1,2}((A_i)_i, U, V)$
- Check whether U_1, U_2, U_3 with $\text{end}_{1,2}((A_i)_i, U_j)$ satisfy $g_3 = g_1 \cdot g_2$:

$$\text{comp}_{1,2}((A_i)_i, U_1, U_2, U_3) \Leftrightarrow$$

$$\exists V_1, V_2 : \bigwedge_{j=1}^3 \text{end}_{1,2}((A_i)_i, U_j) \wedge \text{conv}_{1,3}^{1,2}((A_i)_i, U_1, V_1)$$

$$\wedge \text{conv}_{3,2}^{1,2}((A_i)_i, U_2, V_2) \wedge U_3 \stackrel{t}{\leq} V_1 V_2$$

Undecidability of CI Implication [Li, 2022b]

- Uniform word problem for finite monoids [Gurevich, 1966]

$$\bigwedge_{i=1}^l (x_{a_i} \cdot x_{b_i} = x_{c_i}) \rightarrow (x_{a_0} = x_{c_0})$$

holds for all finite monoid \mathcal{M} and all k -tuples $x_1, \dots, x_k \in \mathcal{M}$ is true iff...

-

$$\left(\bigwedge_{j=1}^k \text{end}_{1,2}((A_i)_i, U_j) \wedge \bigwedge_{j=1}^l \text{comp}_{1,2}((A_i)_i, U_{a_j}, U_{b_j}, U_{c_j}) \right) \rightarrow (U_{a_0} \stackrel{v}{\leq} U_{c_0})$$

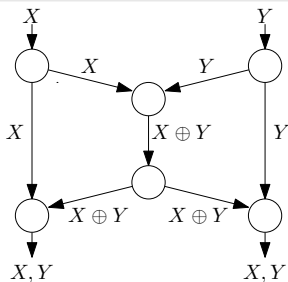
holds for all finite random variables $(A_i)_i, U_1, \dots, U_k$

- Since uniform word problem for finite monoids is undecidable, CI implication is undecidable as well

Related Problems: Linear Information Inequalities

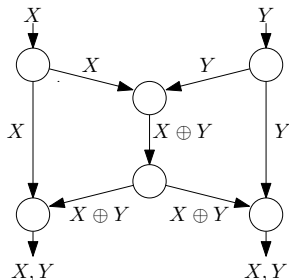
- Sequence of random variables $X^n = (X_1, \dots, X_n)$
- Entropic vector [Zhang and Yeung, 1997] $\mathbf{h}(X^n) = \mathbf{h} \in \mathbb{R}^{2^n - 1}$, where entries of \mathbf{h} are indexed by nonempty subsets of $[n]$, and $\mathbf{h}_S := H(X_S)$
- Entropic region $\Gamma_n^* := \bigcup_{\rho_{X^n}} \{\mathbf{h}(X^n)\}$ [Zhang and Yeung, 1997]
- Non-Shannon inequalities (cannot be deduced from $I(X; Y|Z) \geq 0$) were given in [Zhang and Yeung, 1998, Makarychev et al., 2002, Dougherty et al., 2006b]
- Matúš [2007] showed that $\overline{\Gamma_n^*}$ is not polyhedral
- Conditional information inequalities: whether a linear inequality follows from a list of inequalities
 - Can encode conditional independence implication, and hence **undecidable** [Li, 2022b]
- Decidability of unconditional information inequalities is open

Related Problems: Network Coding



- Network coding [Ahlsweede et al., 2000, Li et al., 2003]
 - Network of nodes connected by noiseless links with same capacity
 - Each source node has a message, and each destination node desires a set of messages
 - Each node is capable of performing coding, not only routing
- If there is one source and multiple destinations, the capacity (number of message bits per link capacity) is given by the maximum network flow [Ahlsweede et al., 2000]
 - Single-source multicast network coding is **decidable**
- Significantly harder if there are multiple sources

Related Problems: Network Coding



- NP-hardness results: Lehman [2005], Langberg et al. [2006], Langberg and Sprintson [2011]
- Is network coding decidable?
 - Given a network, if the message size and the link capacity are the same, does there exist a valid coding scheme?
 - Partial result: whether a network admits a vector linear network code is undecidable [Kühne and Yashfe, 2019]
 - Shown to be **undecidable** in [Li, 2022b]
- Decidability of whether the capacity can be approached is unknown

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