

VENTURING FURTHER INTO EPISTEMIC TOPOLOGY

Johan van Benthem

joint work with Alexandru Baltag

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Background: Information in logic



Knowledge and Information in Logic

information as range

epistemic logic



Jaakko Hintikka

information as correlation

situation theory



Jon Barwise



Further Varieties of Information in Logic

also **procedural information** (about process of inquiry) **information as code** (deduction, computing)

and more …











Enter Dependence



Dependence and Information

physical dependence of y on X in the world

determined, or correlated behavior

covers a family of notions/intuitions

epistemic dependence

if you know the values of X you know the value of y

related epistemic view: implication between questions



A Dependence Table

Toy example: assignment of objects to variables

global dependence fixing value of *x* fixes that of *y*

- х у г
- 0 0 0
- 0 0 1
- 1 1 1 2 1 0

 $x \rightarrow y$: y depends on x

dependence graph

 $x \rightarrow y$, not $y \rightarrow x$

not
$$z \rightarrow x$$
, $\{y, z\} \rightarrow x$

local dependence more basic!

x depends on *y* **at** (**0**, **0**, **0**) but not at (**1**, **1**, **1**)



Dependence Models

 $\mathcal{M}=(V,\;O,\;S,\;P)$

V variables

,

- *O* objects (possible values of variables)
- S states: family of functions from V to O need not be full O^V : gaps = dependence

- $D^{s}_{X}y \quad \forall t \in S: \text{ if } s =_{X} t, \text{ then } s =_{Y} t \quad D_{X}y \quad \forall s \in S: D^{s}_{X}y$
- with $s =_x t : s(x) = t(x)$, $s =_X t : \forall x \in X : s =_x t$
- also $D_X Y : \forall y \in Y : D_x y$



Functional Definability

useful shift in perspective: x(s) := s(x)
variables as maps from states to objects
states might now be more than assignments
a simple factorization result

,

D_X**y** holds globally in a dependence model iff there exists a **(partial) function F** from tuples of X-values to y-values s.t. for all s:

 $\mathbf{y}(\mathbf{s}) = \mathbf{F}(\mathbf{X}(\mathbf{s})) \qquad \quad \mathbf{'y} = \mathbf{F} \circ \mathbf{X'}$







Modal Dependence Logic LFD Language and Semantics





LFD Baltag & van Benthem 2021



Journal of Philosophical Logic 50, 939–1005 (2021) Cite this article

a modal logic of **functional dependence**



Basic Modal-Style Dependence Logic

assignment of values to variables: state of the world dependence because of gaps in full function space dependence model: family of assignments propositional language plus $D_{x}\phi \mid D_{x}y$ **D_xy** local dependence at state s: if s = x, then s = t $D_{x}\phi$ local modality/quantifier: for all t s.t. s = t, t = ϕ universal modality, global dependence definable



Expressive Power

• Changing x implies changing y $D_y x$

more complex versions with '(pre-causal) influence'

- $D_{\varnothing}\varphi$ is the **universal modality** $U\varphi$
- Global dependence defined from local: UD_Xy
 - Dependence as value restriction:

if x lies within some range, so does y: $U(Q_1x \rightarrow Q_2y)$

• universal quantifier $\forall X \varphi$: **D**_{VAR-X} φ



Given States and DL

Extends CRS:

Generalized FOL semantics drops independence assumptions:

decidable sublogic remains

now add explicit dependence atoms

Different take from **DL**:

not second-order logic on teams

classical propositional base

(switching teams would require dynamic modalities)



Dependence models : 'generalized assignment models' M = (D, V, I) with V set of 'available' assignments

M, $s \models \exists x. \varphi$ iff there exists t in \mathcal{V} with $s \models t$ and *M*, $t \models \varphi$ $s \models t : s(y) \models t(y)$ for all variables y distinct from x

Theorem The first-order validities on generalized assignment models are recursively axiomatizable and decidable.

Drops **independence** principles like $\exists x. \exists y. \varphi \rightarrow \exists y. \exists x. \varphi$: these impose existential confluence properties on the set \mathcal{V} .



É Translation Into FOL on Standard Models

- Thm There is a translation *tr* from the language of LFD into first-order logic making the following equivalent for modal formulas *φ*:
 (a) *φ* is satisfiable in a dependence model,
 (b) *tr*(*φ*) is satisfiable in a *standard* first-order model.
- Trick: finitely many variables **x**, code that a tuple of values for **x** is an available assignment with new **dedicated predicate** U**x**.
- **Corollary** All logics that we will consider are RE (axiomatizable). **Caveat** Not reduction of object-language, but meta-language.



Axiom System for LFD

The proof system LFD consist of

(a) The principles of modal S5 for each separate D_Xφ
(b) Monotonicity D_Xφ → D_{X∪Y}φ
(c) Reflexivity, Transitivity, Monotonicity for atoms D_Xy
(d) Transfer axiom (D_XY ∧ D_Yφ) → D_Xφ
(e) Invariance (¬)Qx → D_X(¬)Qx, (¬)D_Xy → D_X(¬)D_Xy









Independence and Correlation



Independence and Correlation

Independence is not the negation of dependence $\neg D_{\chi}y$.

Natural sense of **independence** of **y** from X: fixing the values of X leaves y free to take on any value it can take in the model ('knowing X implies no knowledge about y'). Can be formalized as an **independence atom I_xy**.

¬l_xy some local correlation, D_xy total correlation spectrum of (quantitative) correlations in between







Completeness and Other System Results



Some Results About the System

Thm LFD is sound, complete, decidable

Thm LFD + I axiomatizable, undecidable

Open PL + just I decidable?

some follow up results

LFD + terms + = undecidable Graedel & Puetzstueck includes general analysis of localization and decidability bisimulation analysis LFD Puetzstueck, Koudijs LFD has Finite Model Property Koudijs



mutual interpretability LFD and **Guarded Fragment** (solves computational complexity LFD) Koudijs, ten Cate

dependence in dynamical systems

link up with temporal logic Ba & vB with Dazhu Li

dependence & independence in Linear Algebra/Matroid Theory modal logics of **vector spaces** vB & Nick Bezhanishvili

> Why mention all these technical LFD topics? Interesting to see later which ones make sense (and if so, how) in a topological setting







Empirical Inquiry, Measurement, Topology



From Sharp Values to Empirical Measurement

LFD semantics assumes sharp values and suggests that we can know these

this works for many epistemic puzzles and scenarios but it is highly idealized in many settings

empirical inquiry yields only approximate measurements



open sets

outcomes of possible **observations approximation** of values now essential

many versions of this idea:

Vickers, Parikh & Moss

also Intuitionistic Logic since 1930s

Topology of information states, or: stages in temporal history of inquiry



Felix Hausdorff



Marshall Stone





General Background: Logics of Space









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Topological Dependence Logic



Topo-Dependence Models

 $M = (S, O, \{O_x\}_{x \text{ in VAR}}, V)$

set of admissible states **S** variables **x** map states to objects x(s) in their ranges O_x these ranges carry topologies O_x measurements of **x** yield opens in O_x

- * separation axioms make epistemic sense: T_0
- * lift to finite sets of variables using product topology

topologies on values can be retracted to O_x on states: the smallest topology that makes the map x continuous often makes for more intuitive formulations



Knowledge and Knowability

local modal semantic format **M**, **s** $\models \varphi$

Uφ what is true in all states is **hard knowledge** [can be made more sophisticated in various standard ways]

 $K_{\chi}\phi$ local **knowability** under **new information** ϕ true in some open X-neighborhood of s

the standard topological interpretation of modal logicS4 instead of S5 lots of theory in epistemic logic(also on modal logics for product topologies)



Important Special Topologies

discrete topology contains our earlier LFD approach

Alexandrov topology each point has smallest open NBD ~ standard relational models: specialization pre-orders

Proposition In an Alexandroff topo-dependence model M, the above semantical clauses are equivalent to the following relational versions:

 $s \models_{\mathbf{M}} K_X \varphi \quad iff \quad \forall t \in S(s \leq_X t \Rightarrow t \models \varphi),$

also for later notions $s \models_{\mathbf{M}} K_X Y$ iff $\forall t, w \in S(s \leq_X t \leq_X w \Rightarrow t \leq_Y w)$,

where \leq_X is the specialization preorder for the X-topology $\tau_{\mathbf{X}}^S$ on states.

technical uses our completeness proofs all pass via **representation theorems** for such relational models



Alternative/Richer Frameworks

topo-logic (Parikh & Moss)

dynamic-epistemic update logics like PAL and DEL for explicit changes of topological models

temporal logics with topological structure (dynamic topological logic **DTL**)

here we will stick with the simplest base setting









Continuity and Learnability



Continuity as Information-Carrying Dependence

we may know that a dependence exists but it may not be useful for approximation

k_xy approximable dependence of y on X:
get arbitrarily close to y(s) by measuring X(s)
locally₁ continuous function at state s

may still not yield knowledge:

we do not know which state s we are at

local continuity can be fragile



- maybe the more natural topological sense is **locality**₂: truth **throughout** some open neighborhood of s
- **knowable dependence K_xy** y depends on X continuously over some open neighborhood U of s
- **I.e.** dependence function y(X) on states factors over continuous map from X-values to y-values on the set U

similar factoring views for our other notions to come





Connections to Formal Learning Theory

if the spaces are first-countable
(countable basis for open NBHs)
continuous y is gradually learnable
from X under some X-data stream



Proposition Let $X : S \to (\mathbb{D}_X, \tau_X)$ and $Y : S \to (\mathbb{D}_Y, \tau_Y)$ be observational variables The following are equivalent:

- 1. Y is gradually learnable from X-observations;
- 2. there exists a continuous function $F : (\mathbb{D}_X, \tau_X) \to (\mathbb{D}_Y, \tau_Y)$ s.t. $F \circ X = Y$;
- 3. there is a known epistemic dependence between the variables $X : S \to (\mathbb{D}_X, \tau_X)$ and $Y : S \to (\mathbb{D}_Y, \tau_Y)$, i.e. we have $S \models K(X; Y)$.

here $K(X; Y) = U k_x y$







The System LCD


LCD Documents since 2020

The Logic of Continuous Dependence	
Alexandru Baltag & Johan van Benthem	
	Knowability as Continuous Dependence
	Alexandru Baltag, ILLC, Amsterdam

Based on joint work with Johan van Benthem.

conceptual/mathematical analysis of epistemic notions design of simple modal base logics sequence of completeness and decidability results uniform approach via representation theorems



LCD Language and Semantics

M, s \models K_X ϕ iff \exists U \in O_X: X(s) \in U & \forall t with t(X) \in U: **M**, t \models ϕ

equivalent in our earlier terms: some open NBH in the retracted O_X -topology at s is contained in [[ϕ]] the usual topological interpretation of modal logic

M, s \models k_Xy iff \forall U \in O_y s.t. s(y) \in U \exists V: \forall t with t(X) \in V: t(y) \in U

or: every y-open NBD of s contains an X-open NBD of s again too fragile? replace by local₂ version

M, s \models K_Xy iff **M**, s \models K_Xk_Xy





É Digression: First-Order Perspective

what can we expect a priori?

LFD and its related systems were all translatable effectively into first-order logic explains the axiomatizability of these logics

definition of topological space is second-order, but: if we think of the outcomes of measurements as a base attractive **three-sorted FOL-translatable** versions exist



Axioms and rules of LFD **(I)** (II) S4 Axioms for knowability: (K-Necessitation) From φ , infer $K_X \varphi$ $K_X(\varphi \to \psi) \to (K_X \varphi \to K_X \psi)$ (K-Distribution) (Veracity) $K_X \varphi \rightarrow \varphi$ $K_X \varphi \to K_X K_X \varphi$ (Positive Introspection) $\{t \mid s =_{x} t\}$ contained in every X-open (Knowable Determination) $K_X \varphi \Rightarrow D_X \varphi$ around s, but need not itself be open. $A\varphi \Rightarrow K\varphi$ (Knowledge of Necessity) (IV) Axioms for knowable dependence: $K_X Y$, provided that $Y \subseteq X$ (Inclusion) $(K_XY \wedge K_XZ) \rightarrow K_X(Y \cup Z)$ (Additivity) $(K_XY \wedge K_YZ) \rightarrow K_XZ$ (Transitivity) (Knowable Dependence) $K_X Y \rightarrow (D_X Y \wedge K_X K_X Y)$ KT expresses continuity of (Knowledge of Constants) $C(Y) \Rightarrow KY$ dependence, even modal (Knowability Transfer) $K_X Y \to (K_V \varphi \to K_X \varphi)$ corrrespondence-style.

Table 1: The proof system **LCD**, where we used the notations $A\varphi := D_{\emptyset}\varphi$, $C(Y) := D_{\emptyset}Y$, and $K := K_{\emptyset}$ (for both formulas φ and sets $Y \subseteq V$).

explanation: look like LFD axioms but different content



LCD modal logic of knowability plus knowable continuity

Thm LCD is complete for validity

Open Axiomatize full language with local continuity

Thm LCD is decidable

Key proof steps Standard modal completeness for 'general relational models'. Representation theorem general as standard relational models: yields Alexandrov topologies. Decidability via finite general relational model property.

mathematical logic Generalize Tarski-McKinsey Thm for S4 now not just produce the reals, but also continuous maps over them







Computability and Domain Theory



Toward Computable Dependence

Specialization Computability in **Domain Theory**

complete lattices of 'information pieces'

Scott topology with base of finite information

pieces (not the upset topology for inclusion)

Scott continuity ~ abstract **computability**



Dana Scott

Thm Complete and decidable logic is **LCD +** one extra axiom for bottom elements in domain structure

 $K(K_X \varphi \lor K_X \psi) \to (K \varphi \lor K \psi).$

K is universal modality, look at bottom element for X-value domain

Representation proof more complex so as to produce domains!









Independence in Topological Models



Independence in Epistemic Topology

topological independence

$$s \models Ig_X Y$$
 iff $Y(U) = \mathbb{D}_Y$ for all $U \in \tau_X^S(s)$.

subtly different weaker than LFD-style independence which wants to see all y-values on the X-equal states

also symmetric(!) **global versions**

 $I(X;Y) \Leftrightarrow I(Y;X), \qquad Ig(X;Y) \Leftrightarrow Ig(Y;X).$

scope for modal logic: some connections

- $I_X Y \Rightarrow Ig_X Y, \qquad I(X;Y) \Rightarrow Ig(X;Y)$
- $I(X;Y) \Rightarrow I_X Y, \qquad Ig(X;Y) \Rightarrow Ig_X Y$

open problem Topological I-logic axiomatizable, decidable?



Everywhere Surjective Functions

Y as function of X is everywhere surjective Lebesque 1904

sort of opposite to continuous functions

Thm The following are equivalent:

1. $D(X;Y) \wedge Ig(X;Y)$ holds in the model **M** (at any/all states);

2. there exists some everywhere-surjective map $F : \mathbb{D}_X \to \mathbb{D}_Y$ s.t. $F \circ X = Y$.

interesting recent mathematical theory Bernardi and Rainaldi 2018





How to logicize **more general accuracy correlations** in between independence and functional dependence?

And also: inverse accuracy correlations like that between position and momentum in Heisenberg's Uncertainty Principle?







XII

Epistemic Know-How: Uniform Continuity



From Continuity to Uniform Continuity

epistemic know-how in empirical inquiry knowing an approximation that works: uniform continuity $\forall \varepsilon > 0 \forall s \in S \exists \delta > 0 \forall t \in S (d_X(X(s), X(t)) \leq \delta \Rightarrow d_Y(Y(s), Y(t)) \leq \varepsilon).$ $\forall \varepsilon > 0 \exists \delta > 0 \forall s \in S \forall t \in S (d_X(X(s), X(t)) \leq \delta \Rightarrow d_Y(Y(s), Y(t)) \leq \varepsilon).$ LCD-style modal logic on metric dependence models U(X; Y) uniformly continuous dependence map Y(X)

U_XY locally uniform continuous dependence

Proposition Given empirical variables $X : S \to (\mathbb{D}_X, d_X)$ and $Y : S \to (\mathbb{D}_Y, d_Y)$, the following are equivalent:

- 1. U(X;Y) holds;
- 2. there exists a uniformly continuous map $F_{X;Y} : (\mathbb{D}_X, d_X) \to (\mathbb{D}_Y, d_Y)$ s.t. $F \circ X = Y$ holds on S.



The Complete Logic LUD

(I)All axioms and rule(II)Axioms for uniform(U-Inclusion)U(X;Y), provided that(U-Additivity) $U(X;Y) \wedge U(X;Z)$ (U-Transitivity) $(U(X;Y) \wedge U(Y;Z))$ (Uniform Dependence is Known) $U(X;Y) \rightarrow KU(X;Y)$ (Uniformity implies Continuity) $U(X;Y) \rightarrow KU(X;Y)$

All axioms and rules of the system LCD Axioms for uniform dependence: U(X;Y), provided that $Y \subseteq X$ $(U(X;Y) \land U(X;Z)) \rightarrow U(X;Y \cup Z)$ $(U(X;Y) \land U(Y;Z)) \rightarrow U(X;Z)$ $U(X;Y) \land U(Y;Z)) \rightarrow U(X;Z)$ $U(X;Y) \rightarrow KU(X;Y)$

some explanations

Thm LUAD is complete and decidable.

proof runs via 'pseudo metric models':

requires yet more complex representation argument

open problem how to deal with **U_XY**



uniform spaces with entourages Weil 1937

Family \mathcal{U} of ref-sym relations ~ plus refinement closure:

for every $R \in \mathcal{U}$ there is $S \in \mathcal{U}$ with $S \circ S \subseteq R$

Sets {y | y ~ x} generate a topology

qualitative **uniform continuity**

 $\forall U \in \mathcal{U} \exists V \in \mathcal{V} \, \forall x \in D \, \forall y \in D \, ((x, y) \in V \Rightarrow (F(x), F(y)) \in U)$

Analogy with margin-of-error relations in epistemology How to best bring into our kind of logic?





One Option: Modal Logic of Accuracy Dynamics

u assigns binary relations \sim_x to variables **x**

- $r \leq_X r'$ more refined relations on all X-variables
 - $r \models S_{\chi} \phi$ ϕ true in all U(X)-close points

U_Xy y uniformly continuous in X
dynamic modalities [↓_X]φ

 $\phi\,$ true for all X-refinements of current r

logic now includes reasoning about accuracy refinements

~ modal logics of relation change, e.g.:

The Modal Logic of Stepwise Removal

Johan van Benthem, Krzysztof Mierzewski & Francesca Zaffora Blando Review of Symbolic Logic.









Discussion and Conclusions



Our topological models consider global objects 'known' a global dependence function is known in this sense But what if we want to **learn a dependence function**? need to lift our semantics to families of dependence models LFD paper: 'dependence universes' optimal logic still to be developed analogies with learning the content of a set:

suggests lifting of public announcement logic

plus richer static base language



our topological models assume sharply defined functions but is not this at odds with the imprecise measurement setting? possible alternative: **point-free topology** primitive approximation maps run backwards 'as if' representation results create points out of primitive opens, produce continuous function out of approximations





General Enterprise: Epistemize Analysis

develop parallel mathematical notions ~ epistemic notions

various directions:

• common sense analogues of math notions

continuity/uniform continuity: move to knowing how?

• common sense epistemic principles as

high-level expressions of mathematical notions

e.g., $K_x y \rightarrow U_x y$ (epistemic de dicto to de re)/local compactness?





Ú Digression: Where is the Dynamics?

epistemic actions **model change**, LFD + PAL dependence universes

in the topology/metrics: DEL **accuracy** dynamics

causality: dependence plus **interventions** that change causal models



BMS



Amsterdam Dynamics Portal





Summary: Richer Epistemic Topology

Exploit more topology explicitly in epistemic logic

Unify topological and computational aspects

Venture beyond topology into Analysis

not discussed today:

Add further structure beyond dependence: causality

Step up abstraction: category theoretic-framework for LCD style logics (current work by Ye Lingyuan)



Questions and Answers: Just a Few Points

 K_{xy} Why is 'knowable' k_{xy} only based on more X-info?

Yes, could be based on measuring other variables Z. Simple case: when X depends on Z. In general, we would need $K_Z k_X y$ which is not yet in our language...

What are we learning?

Just the actual state **s** of the system. Like the actual world in epistemic scenarios. Learning the state space itself, or a dependence function, requires richer models. We can see this as LCD over more structured states, but we might want a richer language describing the structural content of states.



Questions and Answers: Just a Few Points

Could not we also learn about y by other means than measurement of other variables? [Hope I got this right]

Even in LCD, measuring X gives information about y in combination with another source of information: the structure of the state space and its assignment gaps. And yes of course, other informational events could take place, PAL- or DEL-style, not based on measurement.

How to capture more general intuitions about approximate knowledge in terms of 'getting enough information about X produces enough information about y'?





Questions and Answers: Continued

Spelling out the 'enough' is interesting as a typical qualitative device in stating the gist of mathematical notions. In which direction does it run here: from *y* to *X*, or vice versa? Could entourages offer a good way of studying this?

Is 'knowability' a felicitous term? Better 'will be known' or some other term from current learning-theoretic topology?

Our framework is more about information than knowledge, and yes, there may be better names for what we study.

Connections with knowability in intuitionistic logic?

May be helpful to bring out intuitionistic structures inside LCD.



Questions and Answers: Continued

Note: intuitionistic mathematics make heavy use of (uniform) continuity in its logical analysis of the informational surplus in constructive statements and proofs.

Why is LFD independence complex qua logic? What about topological independence?

In LFD: independence atoms can enforce a full Cartesian product for three variables, and FO(3) is undecidable. Topo-independence is weaker, and no such reduction may exist, leaving room for a decidable D + I logic. Aside: independence makes computational **practice** easier, but this regular structure is precisely what can make for a more complex logical **theory**.



Questions and Answers: Aftermath

for posing further questions or getting better answers:

johan@stanford.edu