

Complexity of Deep Computations via Topology of Function Spaces

Eduardo Dueñez¹

J. Iovino¹, T. Matos-Wiederhold², L. Salvetti², F. D. Tall²

¹University of Texas at San Antonio

²University of Toronto

ISQGD Special Session on Ergodic Theory
and Topological Dynamics

22 March 2026

Stone-Weierstrass: A computational theorem?!

- **Computations** involving real numbers hinge on evaluating certain basic **continuous** operations (e.g., arithmetic) to within tolerances specified in advance.
- Basic operations are implemented by a *floating-point library*.
 - **Caveat!** Any floating point library implements a (*digital!*) algorithm which *only* approximates real-valued arithmetic on *bounded* intervals: each library only works on a **specific compact subset of \mathbb{R}** .

Theorem (Stone-Weierstrass —a computational view)

Given: (1) a continuous map $f : V \rightarrow \mathbb{R}$ on a *finite-dimensional* real vector space V , (2) a **compact subset** $K \subseteq V$ and (3) $\varepsilon > 0$ arbitrary, there is an algorithm $\mathbf{A} = \mathbf{A}_{f,K,\varepsilon}$ which ε -uniformly approximates f on K in the sense that $f(x)$ is ε -approximately evaluated relying only on composition of arithmetic operations, applied to the **coordinates** of x (in some basis of V), as implemented by some *fixed-precision floating-point library*.

Note: The *Universal Approximation Theorem* for neural networks is a strengthening of Stone-Weierstrass narrowing the operations the algorithm \mathbf{A} is allowed to compose.

What does a choice of algorithm depend on?

Take-homes from last slide:

- An **algorithm** to approximate a function on a vector space V depends on the choices of:
 - a **precision** $\varepsilon > 0$;
 - the **compactum** $K \subseteq V$ on which the function is to be ε -approximated;
 - the **coordinatization of V** .

There are many circumstances where a function is naturally defined **only** on some **subset** $L \subseteq V$ of a real vector space V .

- E.g., **log** is defined only on $(0, +\infty) \subseteq \mathbb{R}$.
 - We wish to be able to compute such functions!

Lead-in to real-valued structures

The **finite-dimensionality** of V is rather a **red-herring**:

- Even if V is infinite-dimensional, any algorithm \mathbf{A} on compacta $K \subseteq V$ –or perhaps $K \subseteq L$ for a fixed $L \subseteq V$ – can only use **finitely many real-valued “features”** of points $x \in V$.
- Such “real features” might perhaps be coordinates—but not necessarily so:
 - **Example:** V may carry a norm $\|\cdot\|$. An algorithm (on some subset of V) may depend on the value of $\|x\|$; thus, the algorithm might take the **norm as an implicitly known feature** of its input argument rather than as an intermediate quantity to be computed.

We take **purely relational real-valued structures** from model theory as the class of domains L on which real-valued computations “make sense”.

Real-valued structures and computation states

Definition (Computation States Structure (CSS))

A *Computation states structure with countably many predicates* is of the form

$$\underline{L} = \langle L, P_i \rangle_{i \in \mathbb{N}},$$

where

- L is a **nonempty topological space**;
- P_i is a continuous real function on L (called a *structural predicate*) for each $i \in \mathbb{N}$,

and such that the following **faithful embedding axiom** is satisfied:

The **type map** $\text{tp} : L \rightarrow \mathbb{R}^{\mathbb{N}} : x \mapsto (P_i(x))_{i \in \mathbb{N}}$ is a homeomorphic embedding into the product space $\mathbb{R}^{\mathbb{N}}$.

One may identify L with its type image $\text{tp}[L] \subseteq \mathbb{R}^{\mathbb{N}}$.

- The predicate P_i becomes the projection $x \mapsto x_i$.

Remarks on computation states structures

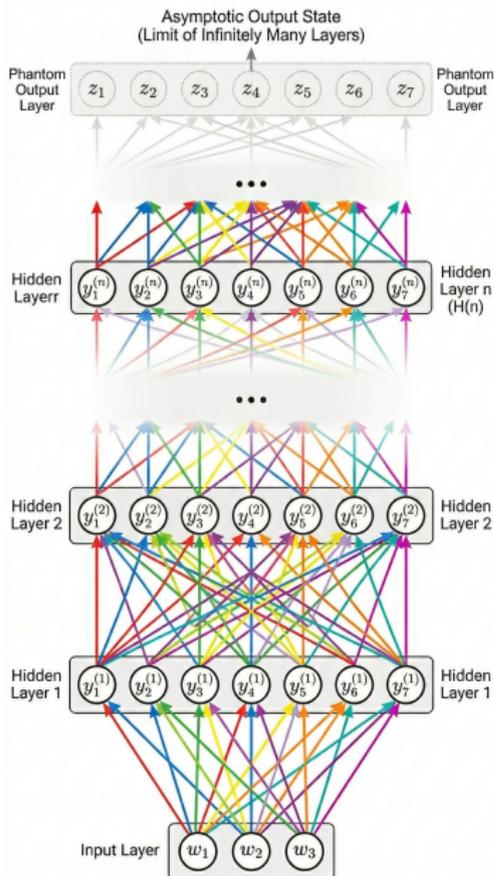
A CSS \underline{L} carries the **topology of coordinate-wise convergence**.

- A sequence/net in L , or filter on L , converges to $x \in L$ iff the pushforward of the sequence/net/filter by P_i converges to $P_i(x)$, for all i .

The model-theoretic (structural) perspective shines light on:

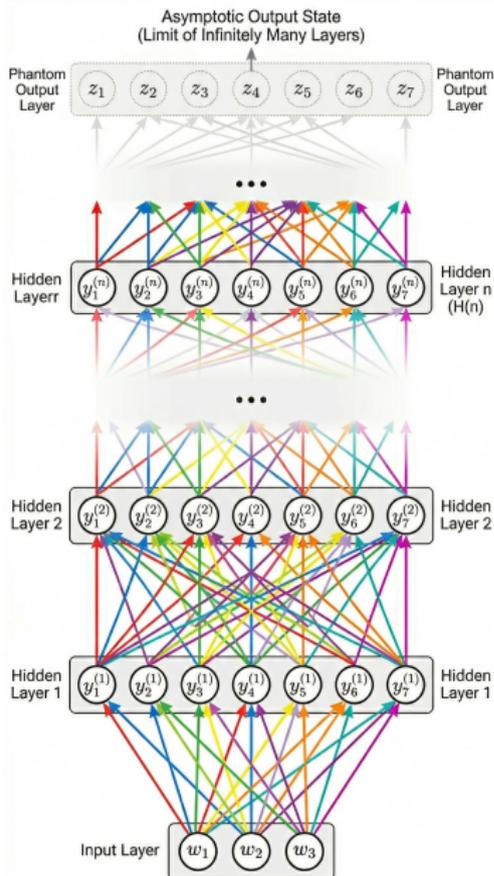
- 1 **Theories**: Relationships between predicate values,
 - such theories may be at the heart of specific algorithms;
- 2 **Explicit definability** of specific real functions on the states space,
 - definability is the model-theoretic notion capturing the existence of algorithms.

Deep Equilibria of Neural Networks



- A deep (but finite) neural network is an example of a CSS with finitely many predicates.
- An “infinitely deep” neural network is a CSS with countably many predicates.
 - Such infinitely deep networks are obtained by “**tying weights**”, i.e., by iterating a fixed map (single-layer transition) *ad infinitum*.
 - A dynamic viewpoint leads to the notion of **Compositional Computations Structure (CCS)**.
 - *Simplest CCS*: (\underline{L}, T) where \underline{L} is a CSS and $T : L \rightarrow L$ is continuous.

Deep Equilibria of Neural Networks



- A “**deep state**” of (\underline{L}, T) is any function $\tau : L \rightarrow L$ arising as pointwise (ultra)limit of maps T^n as $n \rightarrow \infty$.
- Some such deep state $\tau \in L^L$ often arise (only) as *ultralimits*, i.e., as *accumulation points* of $\{T^n\}_{n \in \mathbb{N}}$ in the product topology, called **ultracomputations**.

Theorem (Alva, D, Iovino, Walton '24)

If \underline{L} is realcompact and, for each fixed i , the iterates set $\{P_i \circ T^n\}_{n \in \mathbb{N}}$ is bounded on compacta $K \subseteq L$, then (\underline{L}, T) has **deep equilibria**, i.e., **idempotent ultracomputations** $\tilde{T} : L \rightarrow L$.

Proof: Ellis-Numakura Lemma.

CCS Example: Newton-Raphson Dynamics

Task: Find roots of a **complex polynomial** $p(z)$.

- **States:** $L = \hat{\mathbb{C}} \cong S^2$ (compact).
- **Predicates:** Inverse stereographic coordinates (P_1, P_2, P_3) .
- **Evolution:** Iterates of the **Newton map** $N_p(z) = z - \frac{p(z)}{p'(z)}$.

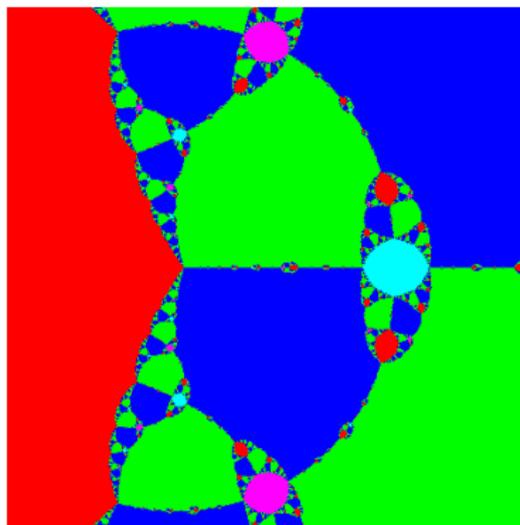
Deep Computations: **not Baire-1!**

A **deep computation** \tilde{T} “chooses” one specific **sub-sequential limit**

$\tilde{T}(z_0)$ of $(N_p^{(n)}(z_0))_n$ for each z_0 .

The limit for **every** z_0 is **not** realizable sequentially globally:

- such \tilde{T} is **not** the pointwise limit of any subsequence of $(N_p^{(n)})_n$.



Newton Fractal for $z^3 - 2z + 2$

Visualizing Newton-Raphson Dynamics

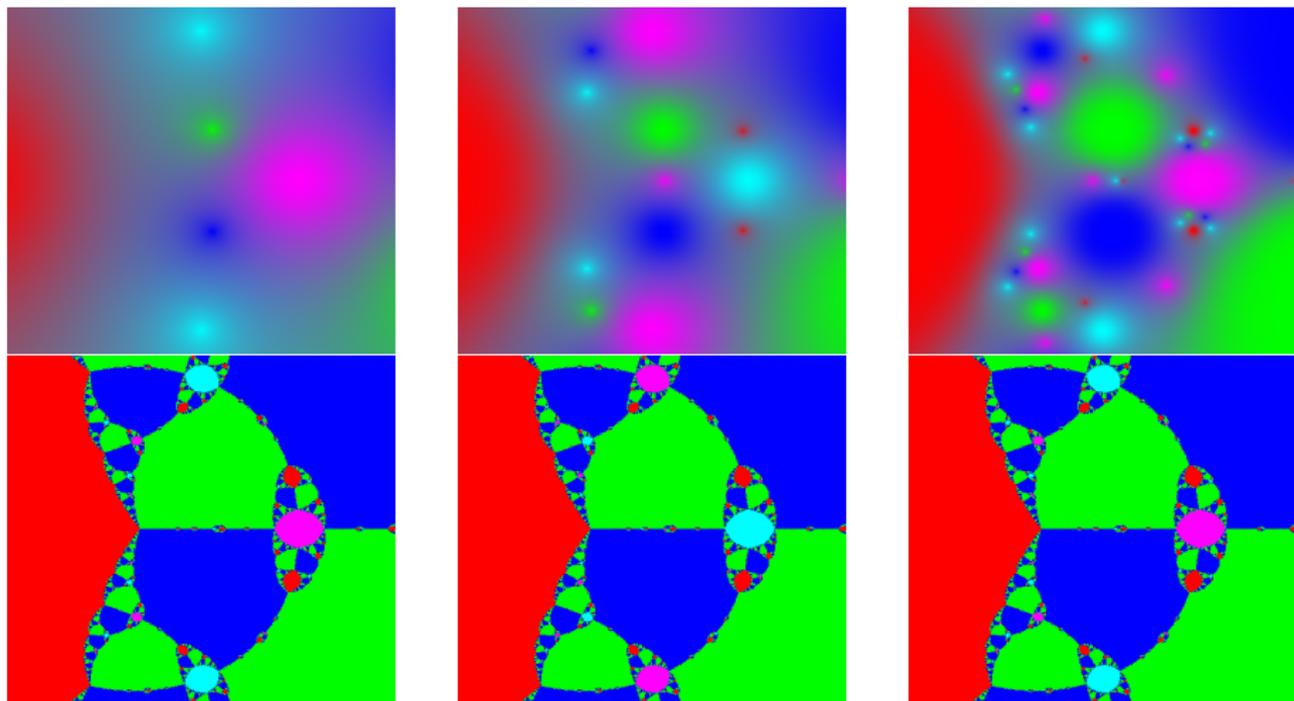
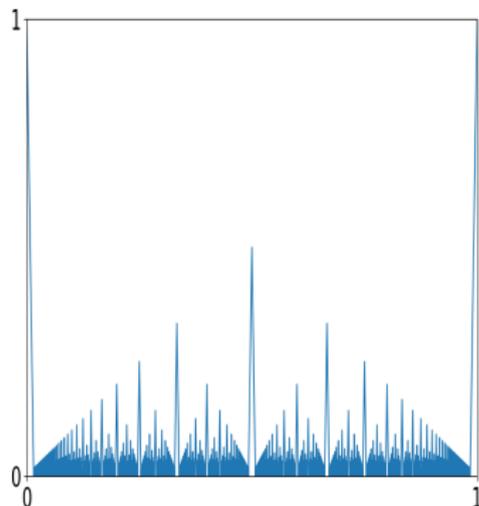


Figure: Newton iterations ($n = 1, 2, 3, 99, 100, 101$). Red/Green/Blue: Convergence to roots. Magenta/Cyan: Limit 2-cycle $\{0, 1\}$. Boundaries: Julia (fractal) set.

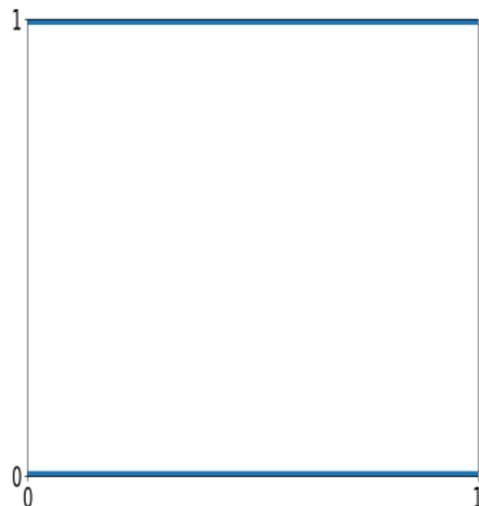
Baire Class 1

$B_1(X, Y)$: Baire Class 1 functions $X \rightarrow Y$

Set of (“Baire-1”) functions: pointwise limits of continuous $X \rightarrow Y$.



64-Farev Approximant to Thomae's function



Dirichlet Function $\mathbf{1}_{\mathbb{Q}}$

Figure: **Left:** Thomae's (“popcorn”) function $x \mapsto \llbracket x = \frac{m}{n} \in \mathbb{Q} \rrbracket \cdot \frac{1}{n}$ is Baire class 1.
Right: Dirichlet's indicator function $\mathbf{1}_{\mathbb{Q}}$ is not Baire-1 (but it is Baire-2).

Deep computations arise as limits. But how wild are these limits?

Theorem (Fundamental Dichotomy (BFT / NIP))

Let $\Delta \subseteq C_p(L, L)$ be a set of continuous transformations of a realcompact CSS \underline{L} . The following are equivalent:

- 1 $\bar{\Delta}$ is *relatively compact* in $B_1(L, L)$ (**tameness**).
- 2 Δ has the **No Independence Property (NIP)**.

If the above conditions hold, then every deep computation $\tau \in \bar{\Delta}$ is the pointwise limit of some sequence in $\bar{\Delta}$ over arbitrary compacta $K \subseteq L$.

NIP (intuition): Δ cannot not “shatter” arbitrarily large finite sets of inputs.
Discrete analogue: Finiteness of **Vapnis-Chervonenkis (VC) dimension**.

Ultracomputations: From tame to wild

If a system satisfies NIP, it is tame. But there are *levels of tameness*.

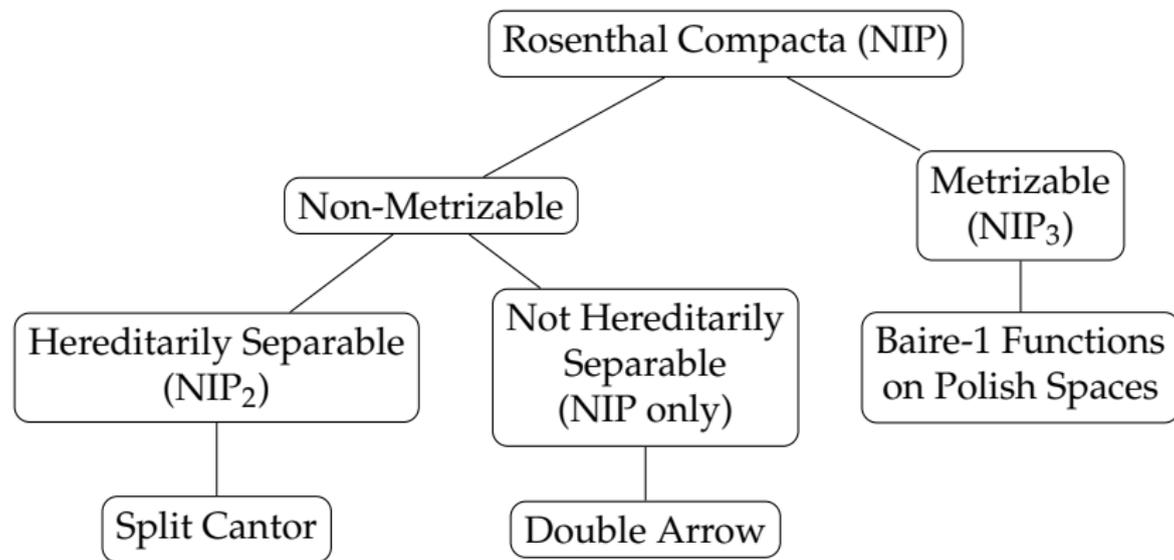
- **Todorčević's Trichotomy** for Rosenthal Compacta (compact subsets of $B_1(X)$) implies **complexity classes** of ultracomputations.
 - Philosophically at least, ultracomputations *not* Baire-1 are **wild**: they cannot be approximated by any algorithm relying solely on floating-point arithmetic over any compactum $K \subseteq L$ witnessing the failure of NIP. —i.e., wild/non-Baire-1 ultracomputations are largely incomputable over the entire states space L in any reasonable sense.
 - A Julia set is an example of a “bad” such compactum.

Remark: Glasner-Megrelishvili's results parallel some of ours.

Important technical differences include:

- Our framework applies to functions on the class of states spaces that are Tychonoff; our strongest results assume only realcompactness (rather than compact metrizable).
- We emphasize connections with the theory of real-valued computing.

The Hierarchy of Tameness



- NIP_3 : Metrizable (most tame). Limits are unique and well-behaved.
- NIP_2 : Hereditarily Separable but not Metrizable.
- NIP_1 : First Countable.

The Seven Prototypes (ADK Heptachotomy)

Argyros, Dodos, and Kanellopoulos identified 7 *minimal building blocks* for Rosenthal compacta. Any wilder behavior must contain one of these prototypes.

- 1 D_1 : Alexandrov compactification (Discrete)
- 2 D_2 : One-point compactification
- 3 D_3 : Split Cantor
- 4 ...
- 5 D_7 : Extended split Cantor

Implication for Computing

If a deep computation system satisfies NIP, its asymptotic behavior is effectively approximated by a **countable discrete set** of "archetypal" limits.

Summary

We introduce a framework to classify the complexity of asymptotic/limit real-valued computations.

| Complexity | Rosenthal Topology | NIP Class |
|--------------|------------------------|----------------|
| Most Tame | Metrizable | NIP_3 |
| Intermediate | Hereditarily Separable | NIP_2 |
| Tame | First Countable | NIP_1 |
| Wild | Independence Property | IP |

Essential Takeaway:

NIP is the **dividing line** between “tame” (Baire-1) transformations/functions that may be evaluated (up to a small error) **computationally feasibly**, and **unapproximable/wild** ones.

Thank You!

References

- S. Alva, E. Dueñez, J. Iovino, C. Walton. *Approximability of deep equilibria*. *arXiv preprint arXiv:2409.06064*, 2024. To appear in *Mathematical Structures in Computer Science*.
- Spiros A. Argyros, Pandelis Dodos, and Vassilis Kanellopoulos. A classification of separable Rosenthal compacta and its applications. *Dissertationes Mathematicae*, 449:1–55, 2008.
- E. Dueñez, J. Iovino, T. Matos-Wiederhold, L. Salvetti, F. D. Tall. Complexity of deep computations via topology of function spaces. *arXiv preprint arXiv:2601.00528*, 2026.
- Eli Glasner and Michael Megrelishvili. Todorčević' trichotomy and a hierarchy in the class of tame dynamical systems. *Transactions of the American Mathematical Society*, 375(7):4513–4548, 2022.
- Stevo Todorčević. Compact subsets of the first Baire class. *Journal of the American Mathematical Society*, 12(4):1179–1212, 1999.