

One Analog Neuron Cannot Recognize Deterministic Context-Free Languages

Jiří Šíma & Martin Plátek

sima@cs.cas.cz



**Institute of Computer Science
Czech Academy of Sciences, Prague**

The Computational Power of Neural Networks (NNs)

(discrete-time recurrent NNs with the saturated-linear activation function)

depends on the information contents of weight parameters:

1. **integer** weights: **finite automaton** (Minsky, 1967)
2. **rational** weights: **Turing machine** (Siegelmann, Sontag, 1995)
polynomial time \equiv complexity class P
polynomial time & increasing **Kolmogorov complexity** of real weights \equiv
a proper **hierarchy** of nonuniform complexity classes **between P and P/poly**
(Balcázar, Gavalda, Siegelmann, 1997)
3. arbitrary **real** weights: **“super-Turing”** computation (Siegelmann, Sontag, 1994)
polynomial time \equiv nonuniform complexity class P/poly
exponential time \equiv any I/O mapping

Motivation: filling the gap between **integer** and **rational** weights

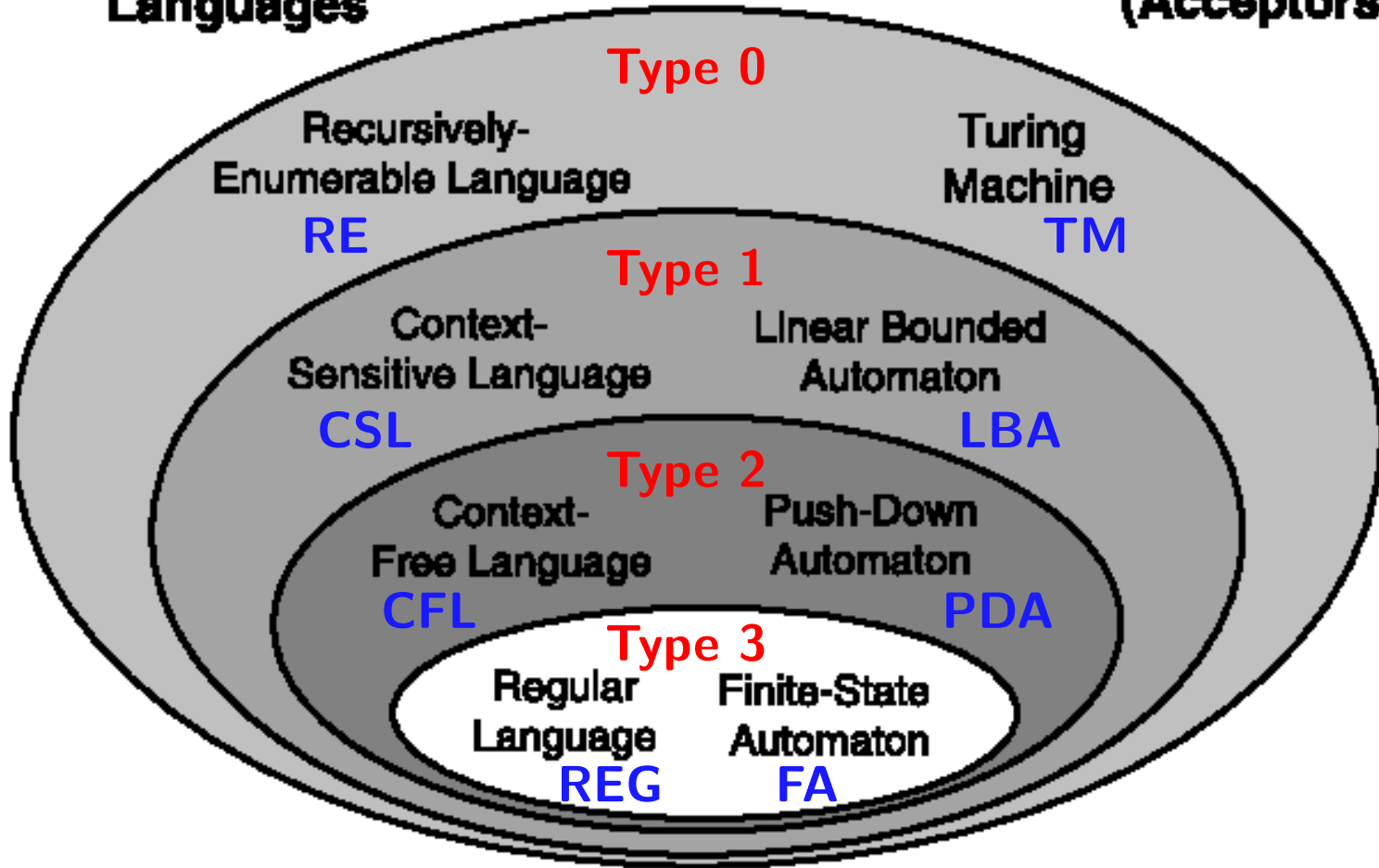
w.r.t. **Chomsky hierarchy**:

regular (Type 3) \times **recursively enumerable (Type 0)** languages

The Traditional Chomsky Hierarchy

Grammars (Generators) & Languages

Automata (Acceptors)



The Formal Language Hierarchy

Neural Networks with Increasing Analogicity

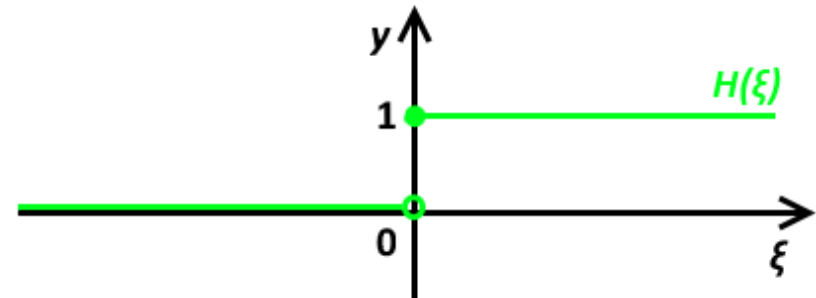
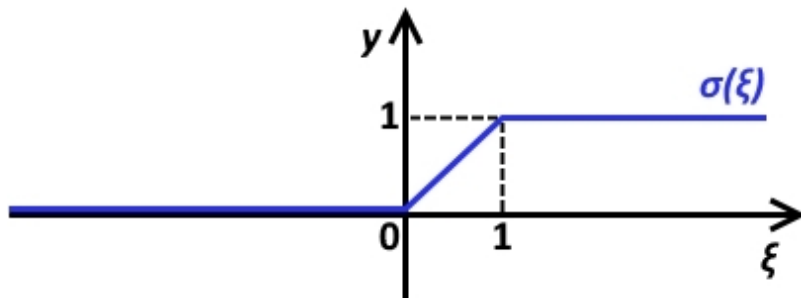
from **binary** ($\{0, 1\}$) to **analog** ($[0, 1]$) neurons' states

α ANN = a **binary-state** NN with **integer** weights

+ **α extra analog-state** neurons with **rational** weights

$$y_j^{(t+1)} = \sigma_j \left(\sum_{i=0}^s w_{ji} y_i^{(t)} \right) \quad j = 1, \dots, s \quad \text{updating the neuron states}$$

$$\sigma_j(\xi) = \begin{cases} \sigma(\xi) = \begin{cases} 1 & \text{for } \xi \geq 1 \\ \xi & \text{for } 0 < \xi < 1 \\ 0 & \text{for } \xi \leq 0 \end{cases} & j = 1, \dots, \alpha \quad \text{saturated-linear function} \\ H(\xi) = \begin{cases} 1 & \text{for } \xi \geq 0 \\ 0 & \text{for } \xi < 0 \end{cases} & j = \alpha + 1, \dots, s \quad \text{Heaviside function} \end{cases}$$

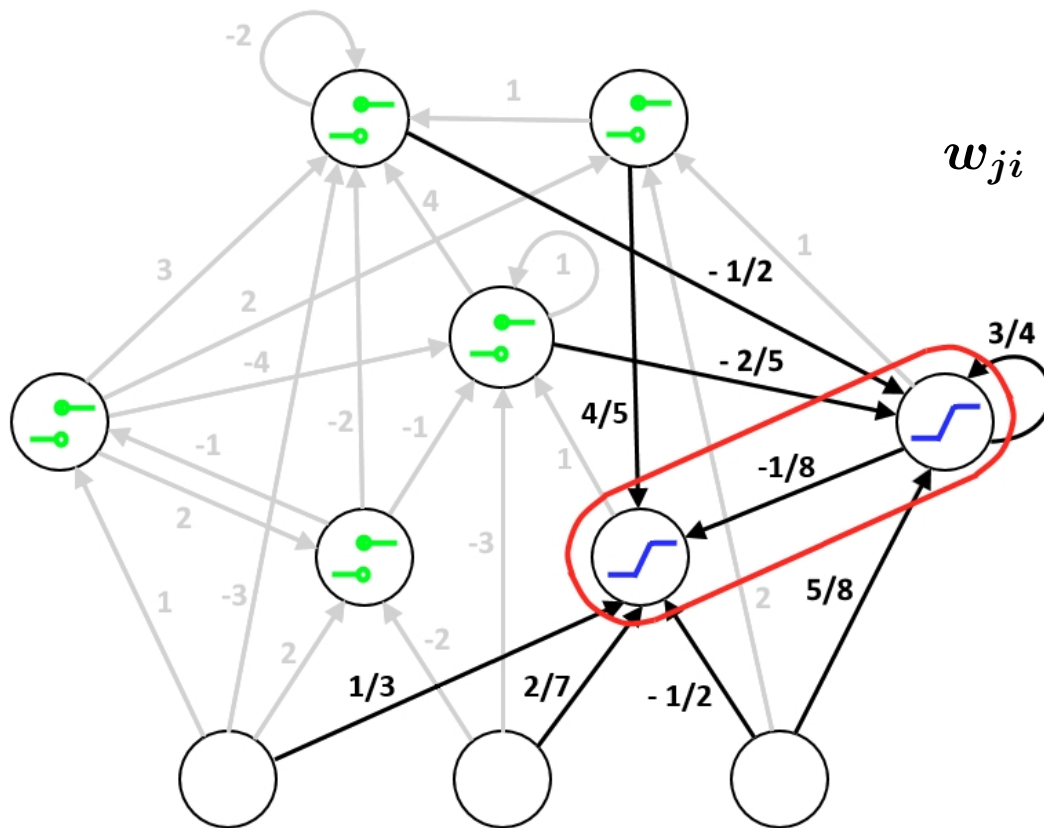


Neural Networks with Increasing Analogicity

equivalently from **integer** to **rational** weights

α ANN = a **binary-state** NN with **integer** weights

+ α **extra analog-state** neurons with **rational** weights

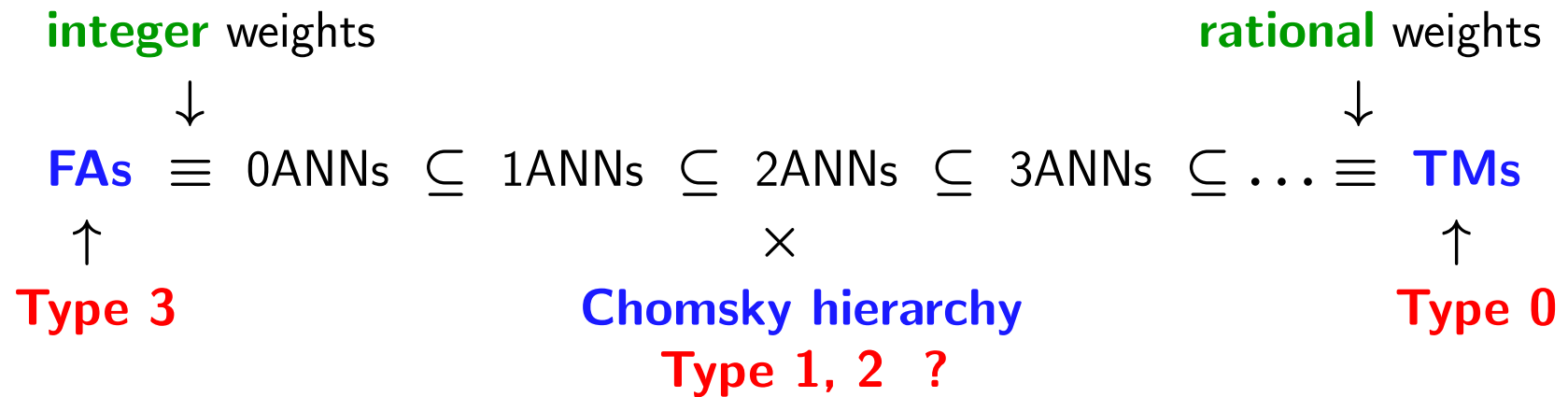


$$w_{ji} \in \begin{cases} \mathbb{Q} & j = 1, \dots, \alpha \\ \mathbb{Z} & j = \alpha + 1, \dots, s \end{cases}$$
$$i \in \{0, \dots, s\}$$

The Analog Neuron Hierarchy

the computational power of α ANNs

increases with the number α of extra analog-state neurons:



The Separation of 1ANNs: $0\text{ANNs} \subsetneq 1\text{ANNs}$ (Šíma, 2017):

- upper bound: $1\text{ANNs} \subset \text{LBAs} \equiv \text{CSLs}$ (Type 1)
- lower bound: $1\text{ANNs} \not\subset \text{PDAs} \equiv \text{CFLs}$ (Type 2)

$$L_1 = \left\{ x_1 \dots x_n \in \{0, 1\}^* \mid \sum_{k=1}^n x_{n-k+1} \left(\frac{3}{2}\right)^{-k} < 1 \right\} \in 1\text{ANNs} \setminus \text{CFLs}$$

Quasi-Periodic Numbers (Šíma, Savický, 2017):

for a fixed real **base (radix)** β ($|\beta| > 1$) and a finite set $A \neq \emptyset$ of real **digits**, we say that a real number x is **quasi-periodic** if every its **β -expansion**

$$x = (0 . a_1 a_2 a_3 \dots)_\beta = \sum_{k=1}^{\infty} a_k \beta^{-k} \quad \text{where } a_k \in A$$

(i.e. non-standard positional numeral system) is **eventually quasi-periodic**:

$$\left(0 . \underbrace{a_1 \dots a_{m_1}}_{\substack{\text{preperiodic} \\ \text{part}}} \underbrace{a_{m_1+1} \dots a_{m_2}}_{\text{quasi-repetend}} \underbrace{a_{m_2+1} \dots a_{m_3}}_{\text{quasi-repetend}} \underbrace{a_{m_3+1} \dots a_{m_4}}_{\text{quasi-repetend}} \dots \right)_\beta$$

such that

$$(0 . \overline{a_{m_1+1} \dots a_{m_2}})_\beta = (0 . \overline{a_{m_2+1} \dots a_{m_3}})_\beta = (0 . \overline{a_{m_3+1} \dots a_{m_4}})_\beta = \dots$$

Example: the plastic $\beta \approx 1.324718$ ($\beta^3 - \beta - 1 = 0$), $A = \{0, 1\}$

$$1 = (0 . 0 \underbrace{100}_{\text{quasi-repetend}} \underbrace{00110111}_{\text{quasi-repetend}} \underbrace{00111}_{\text{quasi-repetend}} \underbrace{100}_{\text{quasi-repetend}} \dots)_\beta$$

with quasi-repetends: $(0 . \overline{100})_\beta = (0 . \overline{0(011)^i 1})_\beta = \beta$ for every $i \geq 1$

1ANNs with Quasi-Periodic “Weights” (QP-1ANNs):

w_{11} is the **self-loop weight** of the one analog-state neuron ($0 < |w_{11}| < 1$)

$\beta = 1/w_{11}$ is the **base**

$A = \left\{ \sum_{i=0; i \neq 1}^s \frac{w_{1i}}{w_{11}} y_i \mid y_2, \dots, y_s \in \{0, 1\} \right\} \cup \{0, \beta\}$ are the **digits**

$X = \left\{ \sum_{i=0; i \neq 1}^s \frac{w_{ji}}{w_{j1}} y_i \mid j \neq 1, w_{j1} \neq 0, y_2, \dots, y_s \in \{0, 1\} \right\} \cup \{0, 1\}$

definition of a **QP-1ANN**: every $x \in X$ is **quasi-periodic**

(e.g. 1ANNs with **Pisot** β + other **weights** from $\mathbb{Q}(\beta)$ are **QP-1ANNs**)

Regular 1ANNs (even with real weights) (Šíma, 2017):

QP-1ANNs \equiv **0ANNs** \equiv **FAs** \equiv **REG (Type 3)**

Example: 1ANNs with **rational** weights + the **self-loop weight** $w_{11} = 1/\beta$

where e.g. β is an **integer** **or** the **plastic constant** (≈ 1.324718)

or the **golden ratio** (≈ 1.618034)

The Collapse of the Analog Neuron Hierarchy (Šíma, 2018)

3ANNs = 4ANNs = 5ANNs = ... \equiv TMs \equiv RE (Type 0)

three analog-state neurons can simulate any **TMs**

The Separation of 2ANNs (Šíma, 2019)

1ANNs \subsetneq 2ANNs

the “counting” language $L_{\#} = \{0^n 1^n \mid n \geq 1\} \in \mathbf{2ANNs} \setminus \mathbf{1ANNs}$

$L_{\#}$ is a (non-regular) **deterministic context-free language (DCFL)**

accepted by a **deterministic push-down automaton (DPDA)**

- $L_{\#} \in \mathbf{DCFLs} \equiv \mathbf{DPDAs} \subset \mathbf{2ANNs}$

two analog-state neurons can simulate any **DPDA**

- $L_{\#} \notin \mathbf{1ANNs}$

one analog-state neuron cannot count up to n (even with real weights)

$\longrightarrow \mathbf{DCFLs} \equiv \mathbf{DPDAs} \not\subset \mathbf{1ANNs}$

The Main Result: The Stronger Separation of 2ANNs

$$(\text{DCFLs} \setminus \text{REG}) \subseteq (\text{2ANNs} \setminus \text{1ANNs})$$

or equivalently $(\text{DCFLs} \setminus \text{REG}) \cap \text{1ANNs} = \emptyset$

$$\text{1ANNs} \cap \text{DCFLs} = \text{0ANNs} \equiv \text{REG}$$

Theorem. *Any non-regular deterministic context-free language L cannot be recognized by any 1ANN with one extra analog unit having real weights.*

Idea of Proof:

by contradiction: suppose $\mathcal{N} \in \text{1ANNs}$ recognizes $L \in \text{DCFLs} \setminus \text{REG}$

a construction of a bigger $\mathcal{N}_{\#} \in \text{1ANNs}$ which exploits \mathcal{N} as its subnetwork
(subroutine) for recognizing the counting language $L_{\#}$

which implies $L_{\#} \in \text{1ANNs}$ – a contradiction

The Simplest Non-Regular Deterministic CFLs

the counting language $L_{\#} = \{0^n 1^n \mid n \geq 1\}$ can be reduced through a Turing-like reduction to every language in the class **DCFLs** \ **REG**:

Theorem. For every non-regular deterministic context-free language $L \subset \Sigma^*$ over a finite alphabet $\Sigma \neq \emptyset$, there exist words $u, w, z \in \Sigma^*$, nonempty strings $x, y \in \Sigma^+$, an integer $\kappa \geq 0$, and languages $L_k \in \{L, \bar{L}\}$ for $k \in K = \{-\kappa, \dots, -1, 0, 1, \dots, \kappa\}$, such that for every pair of integers, $m \geq 0$ and $n \geq \kappa$,

$$(ux^m wy^{n+k} z \in L_k \text{ for all } k \in K) \quad \text{iff} \quad m = n.$$

Example: $L \subseteq \{0, 1\}^*$ is composed of words that contain more 0s than 1s

\longrightarrow u, w, z empty, $x = 0$, $y = 1$, $\kappa = 1$, $L_{-1} = L$, $L_0 = L_1 = \bar{L}$

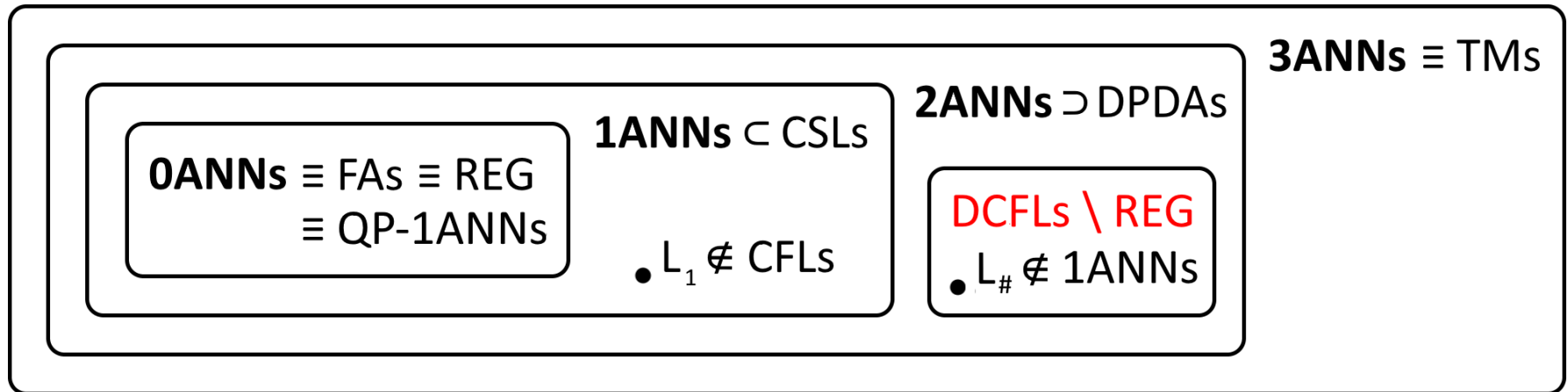
$$(0^m 1^{n-1} \in L_{-1} = L \ \& \ 0^m 1^n \in L_0 = \bar{L} \ \& \ 0^m 1^{n+1} \in L_1 = \bar{L})$$

$$\text{iff } (m > n - 1 \ \& \ m \leq n \ \& \ m \leq n + 1) \text{ iff } m = n.$$

contribution to complexity theory: a counterpart to the hardest problem in a complexity class to which every problem is reduced (e.g. NP-completeness)

A Summary of the Analog Neuron Hierarchy

$$\text{FAs} \equiv \text{0ANNs} \subsetneq \text{1ANNs} \subsetneq \text{2ANNs} \subseteq \text{3ANNs} \equiv \text{TMs}$$



Open Problems:

- the separation of the 3rd level: $\text{2ANNs} \subsetneq \text{3ANNs} ?$
- strengthening the 2nd level separation to the **nondeterministic CFLs**:
 $(\text{CFLs} \setminus \text{REG}) \cap \text{1ANNs} = \emptyset ?$
- a proper “natural” hierarchy of NNs between integer and rational weights which can be mapped to known infinite hierarchies of REG/CFLs ?