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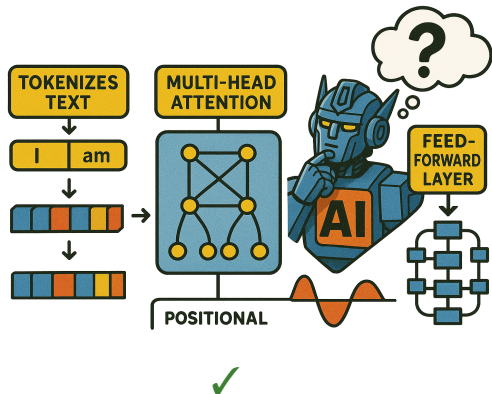
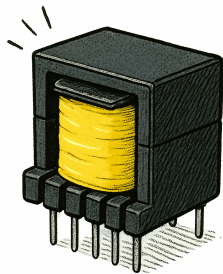
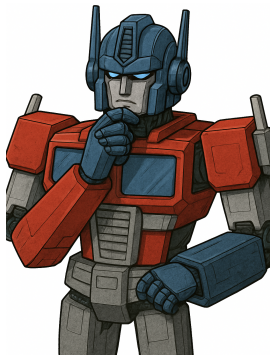
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<sup>2</sup> Centro Nacional de Inteligencia Artificial, Santiago, Chile

<sup>3</sup> Rheinland-Pfälzische Technische Universität, Kaiserslautern & Landau, Germany

September 4, 2025

# Transformers?



- Transformers are the basic model in machine learning used in recent LLMs

 ChatGPT

 Gemini

 Meta AI

# Motivation

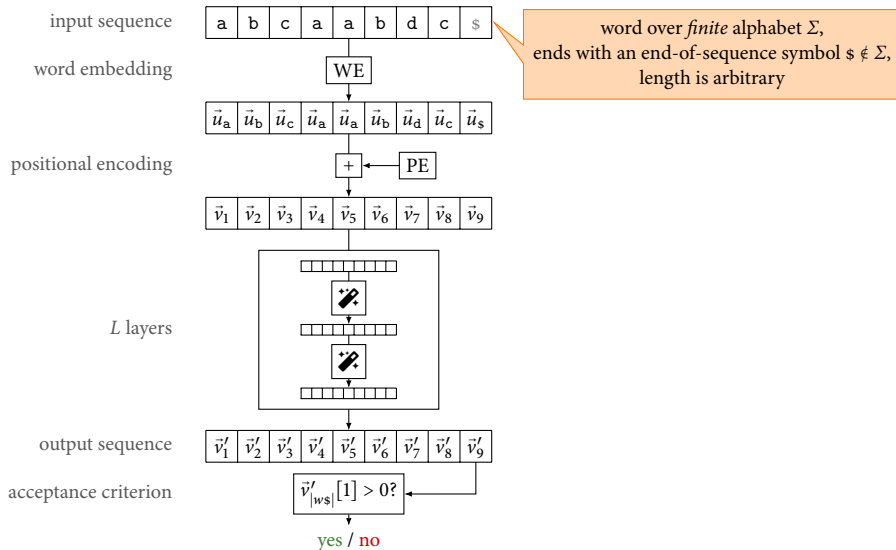
- Artificial intelligence is often not intelligent:



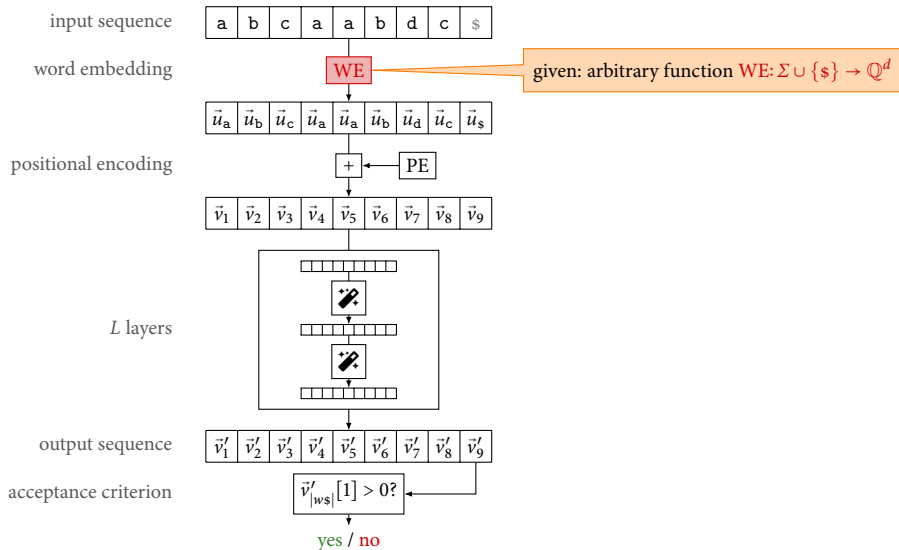
Source: Facebook, “15 most powerful passports in the world”.

- We try to understand what transformers actually can do.
  - A first step: study expressibility

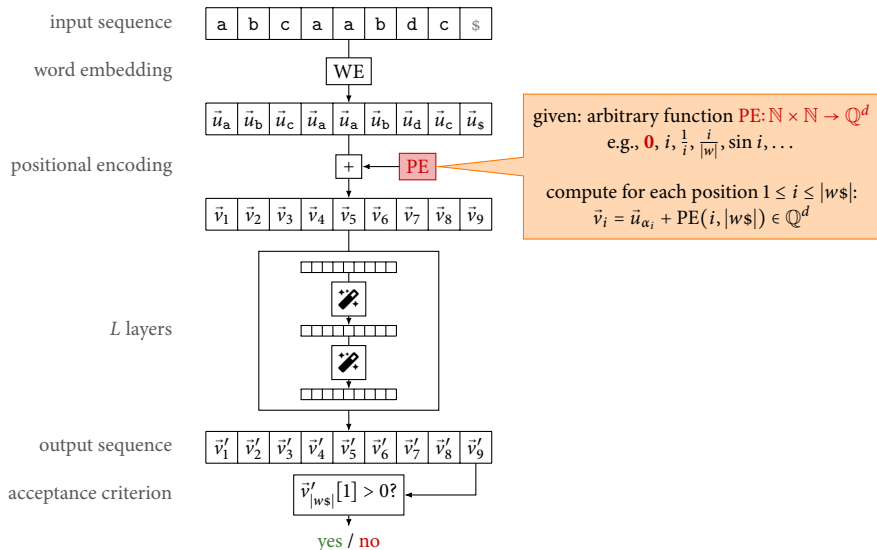
# Transformers: Scheme



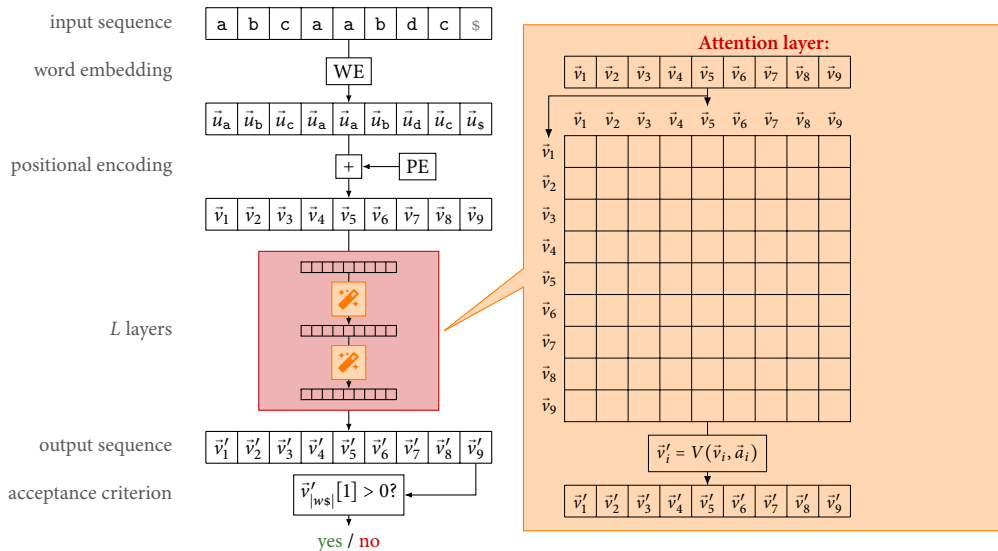
# Transformers: Scheme



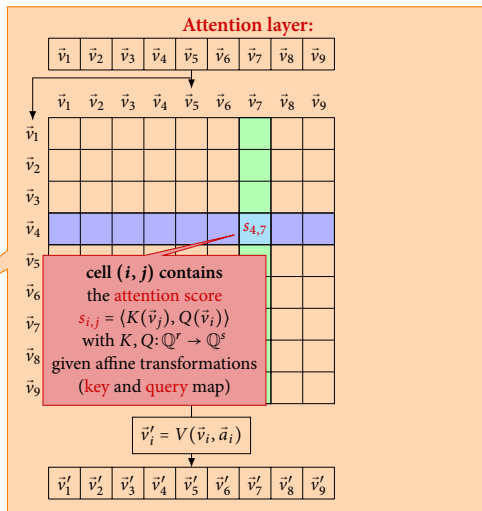
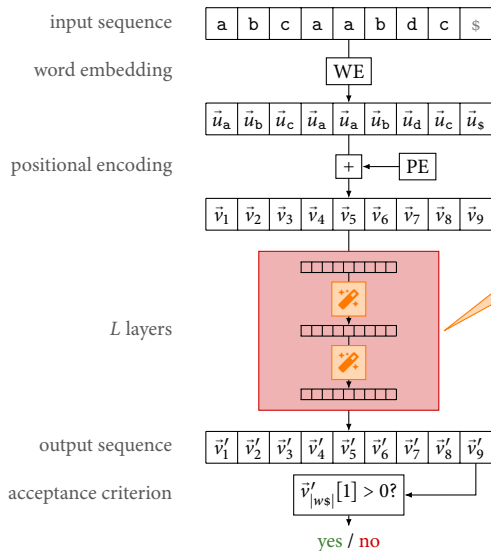
# Transformers: Scheme



# Transformers: Scheme

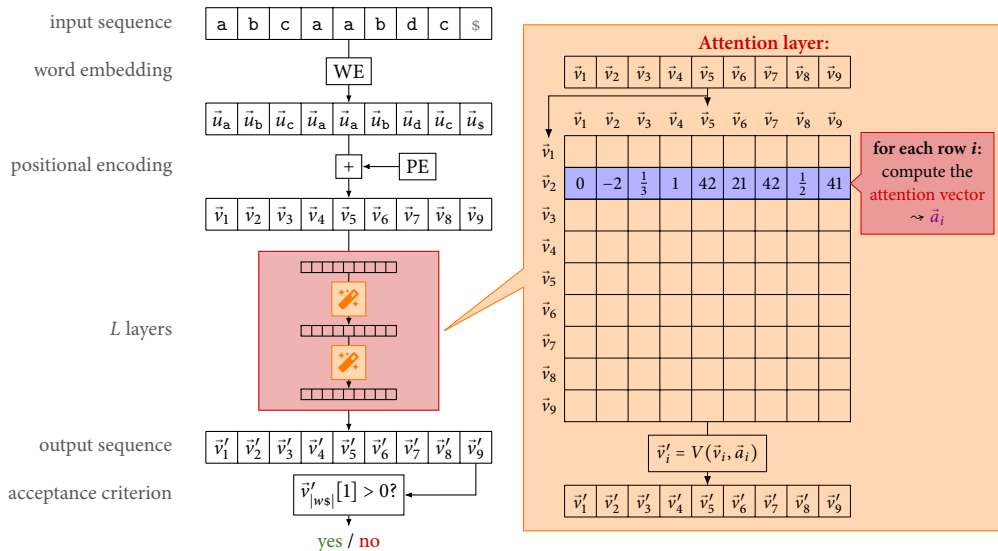


# Transformers: Scheme

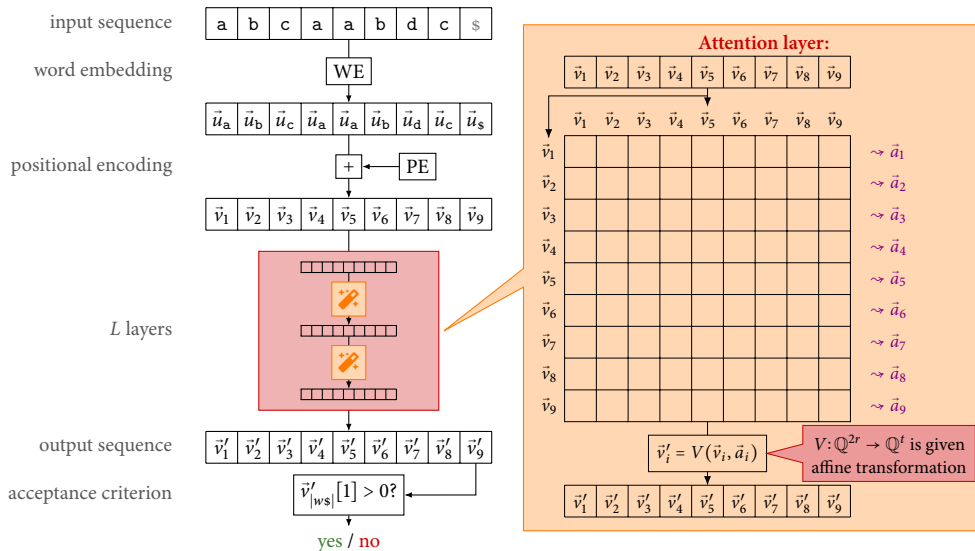




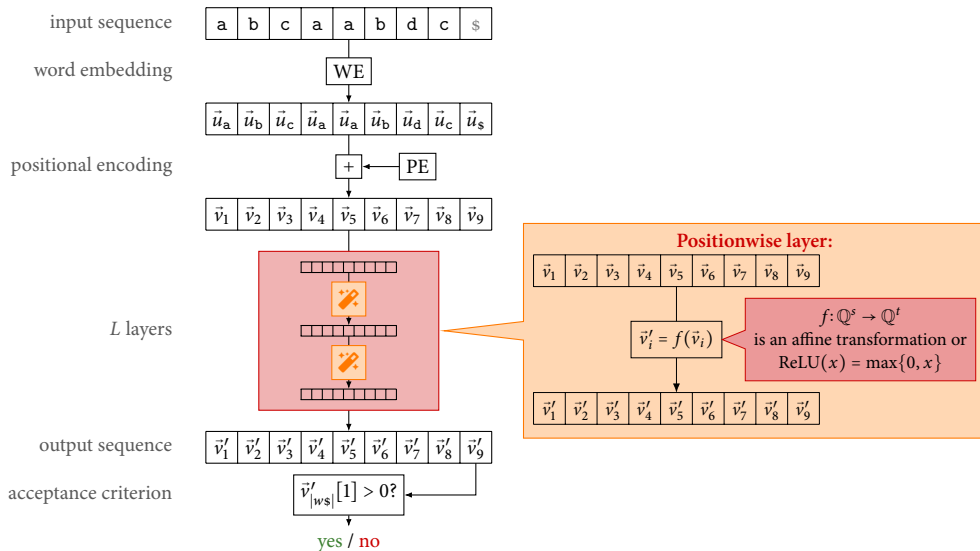
# Transformers: Scheme



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# Transformers: Scheme



# Transformers: Attention Mechanisms

## 1 Unique Hard Attention Transformers (UHAT)

|             | $\vec{v}_1$ | $\vec{v}_2$ | $\vec{v}_3$   | $\vec{v}_4$ | $\vec{v}_5$ | $\vec{v}_6$ | $\vec{v}_7$ | $\vec{v}_8$   | $\vec{v}_9$ |
|-------------|-------------|-------------|---------------|-------------|-------------|-------------|-------------|---------------|-------------|
| $\vec{v}_i$ | 0           | -2          | $\frac{1}{3}$ | 1           | 42          | 21          | 42          | $\frac{1}{2}$ | 41          |

 $\leadsto \vec{a}_i = \vec{v}_5$

## 2 Average Hard Attention Transformers (AHAT)

|             | $\vec{v}_1$ | $\vec{v}_2$ | $\vec{v}_3$   | $\vec{v}_4$ | $\vec{v}_5$ | $\vec{v}_6$ | $\vec{v}_7$ | $\vec{v}_8$   | $\vec{v}_9$ |
|-------------|-------------|-------------|---------------|-------------|-------------|-------------|-------------|---------------|-------------|
| $\vec{v}_i$ | 0           | -2          | $\frac{1}{3}$ | 1           | 42          | 21          | 42          | $\frac{1}{2}$ | 41          |

 $\leadsto \vec{a}_i = \frac{1}{2}\vec{v}_5 + \frac{1}{2}\vec{v}_7$

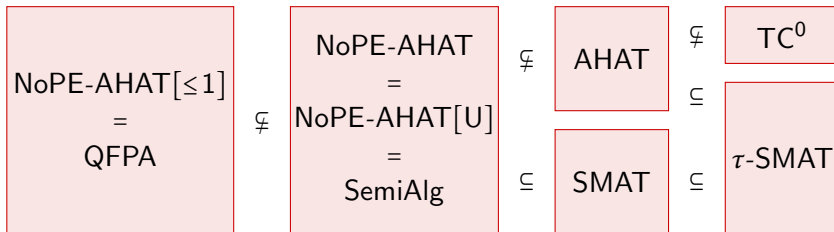
## 3 Softmax Attention Transformers with temperature scaling $\tau > 0$ ( $\tau$ -SMAT)

|             | $\vec{v}_1$ | $\vec{v}_2$ | $\vec{v}_3$   | $\vec{v}_4$ | $\vec{v}_5$ | $\vec{v}_6$ | $\vec{v}_7$ | $\vec{v}_8$   | $\vec{v}_9$ |
|-------------|-------------|-------------|---------------|-------------|-------------|-------------|-------------|---------------|-------------|
| $\vec{v}_i$ | 0           | -2          | $\frac{1}{3}$ | 1           | 42          | 21          | 42          | $\frac{1}{2}$ | 41          |

 $\leadsto \vec{a}_i = \sum_{j=1}^{|w\$|} \frac{e^{s_{i,j}/\tau}}{\sum_{k=1}^{|w\$|} e^{s_{i,k}/\tau}} \cdot \vec{v}_j$

# Transformers: Languages

- accepted language of a transformer  $\mathfrak{T}$ :  $L(\mathfrak{T}) = \{w \in \Sigma^* \mid \mathfrak{T}(w\$) = \text{yes}\}$ .
- AHAT /  $\tau$ -SMAT / SMAT: all languages accepted by an AHAT /  $\tau$ -SMAT / 1-SMAT
- NoPE-C: all languages accepted by a C-transformer without positional encoding (i.e.,  $\text{PE}(i, n) = \vec{0}$ )
- C[U]: all languages accepted by a C-transformer such that each layer is uniform (i.e., if the key and query maps  $K$  and  $Q$  are constant).
- C[ $\leq L$ ]: all languages accepted by a C-transformer with at most  $L$  attention layers.



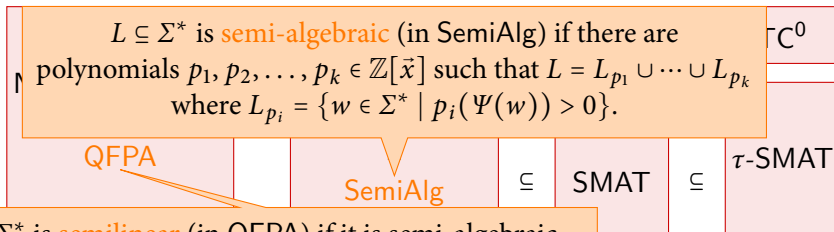
## Definition

Let  $\Sigma = \{a_1, a_2, \dots, a_d\}$  be an alphabet. The **Parikh map** is defined as

$$\Psi: \Sigma^* \rightarrow \mathbb{N}^d: w \mapsto (|w|_{a_1}, |w|_{a_2}, \dots, |w|_{a_d})$$

where  $|w|_{a_i}$  is the number of occurrences of  $a_i$  in the word  $w$ .

# Results



$L \subseteq \Sigma^*$  is **semilinear** (in QFPA) if it is semi-algebraic such that all polynomials are linear (i.e., have degree  $\leq 1$ ).



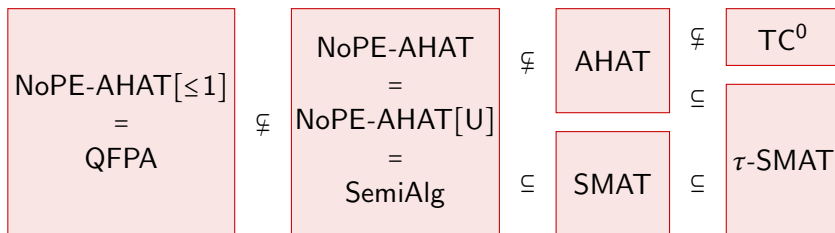
$L$  is described by a Presburger formula.

defined as

$$\Psi: \Sigma^* \rightarrow \mathbb{N}^d: w \mapsto (|w|_{a_1}, |w|_{a_2}, \dots, |w|_{a_d})$$

where  $|w|_{a_i}$  is the number of occurrences of  $a_i$  in the word  $w$ .

# Results



## Definition

Let  $\Sigma = \{a_1, a_2, \dots\}$

where  $|w|_{a_i}$  is the number of occurrences of  $a_i$  in  $w$

## Corollary

$\text{Proj}(\text{NoPE-AHAT}) = \text{RE} \cap \text{PI}$

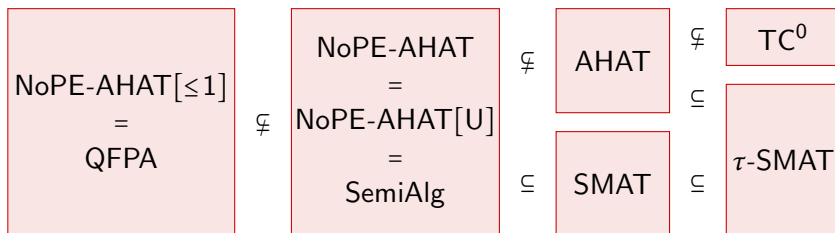
## Corollary

*The emptiness problem is undecidable for (NoPE-)AHAT with two (uniform) layers.*

defined as

$|a_d)$





## Definition

Let  $\Sigma = \{a_1, a_2, \dots\}$

## Corol

recursively enumerable languages

$$\text{Proj}(\text{NoPE-AHAT}) = \text{RE} \cap \text{PI}$$

permutation-invariant languages

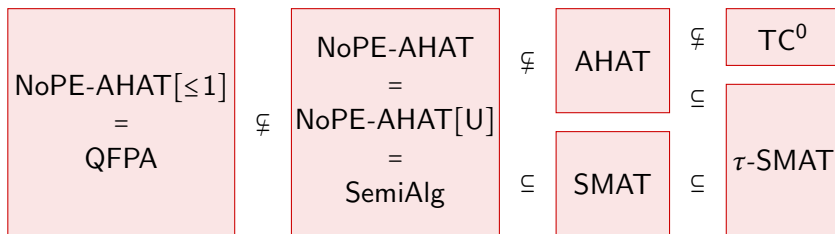
languages  $\pi(L)$  for  $L \in \text{NoPE-AHAT}$   
and projections  $\pi: \Sigma \rightarrow \Gamma \cup \{\varepsilon\}$

wh

is undecidable for  
(NoPE-)AHAT with two (uniform) layers.

$|a_d)$

# Results



## Definition

Let  $\Sigma = \{a_1, a_2, \dots\}$

where  $|w|_{a_i}$  is the number of occurrences of  $a_i$  in  $w$

## Corollary

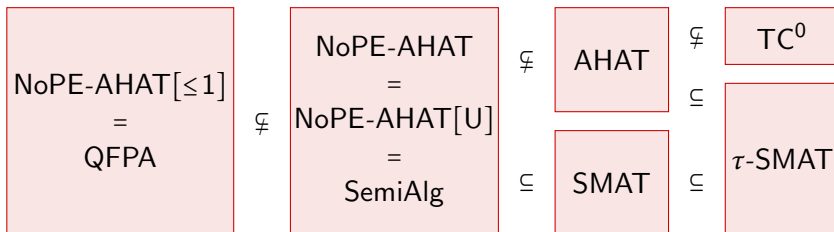
$\text{Proj}(\text{NoPE-AHAT}[U, \leq 2]) = \text{RE} \cap \text{PI}$

## Corollary

*The emptiness problem is undecidable for (NoPE-)AHAT with two (uniform) layers.*

defined as

$|a_d|$



## ■ Open Problems:

- Are  $\text{NoPE-AHAT} \subseteq \text{SMAT}$  and  $\text{AHAT}, \text{SMAT} \subseteq \tau\text{-SMAT}$  proper inclusions?
- Is  $\text{NoPE-AHAT}[\leq L, U]$  equal to semi-algebraic sets with polynomials of degree  $\leq L$ ?

Thank you!

