

*Theorie des Algorithmischen Lernens*  
*Sommersemester 2006*

Teil 2.2: Lernen formaler Sprachen:  
Hypothesenräume

Version 1.1

# Gliederung der LV

## Teil 1: Motivation

1. Was ist Lernen
2. Das Szenario der Induktiven Inferenz
3. Natürlichkeitsanforderungen

## Teil 2: Lernen formaler Sprachen

1. Grundlegende Begriffe und Erkennungstypen
2. Die Rolle des Hypothesenraums
3. Lernen von Patternsprachen
4. Inkrementelles Lernen

## Teil 3: Lernen endlicher Automaten

## Teil 4: Lernen berechenbarer Funktionen

1. Grundlegende Begriffe und Erkennungstypen
2. Reflexion

## Teil 5: Informationsextraktion

1. Island Wrappers
2. Query Scenarios

# Different Approaches

When we have to learn an indexable class  $\mathcal{L} = (L_j)_{j \in \mathbb{N}}$ , we can choose the hypothesis space as follows:

1. use  $\mathcal{L} = (L_j)_{j \in \mathbb{N}}$  as hypothesis space: **exact** identification
2. use another enumeration of  $\mathcal{L} = (L_j)_{j \in \mathbb{N}}$  as hypothesis space: **class preserving** identification
3. use another indexable class  $\mathcal{L}' = (L'_j)_{j \in \mathbb{N}}$  as hypothesis space that contains each  $L_j$ : **class comprising** identification

One could also ask for learnability w.r.t. **all** hypothesis spaces (**absolute** learning)

→ until now, we considered class-comprising learning

- does it make a difference?

# Learning in the Limit

## **Theorem 2.2.1:**

Let  $\mathcal{L} \in \text{LimTxt}$  and let  $\mathcal{H}$  be *any class comprising* hypothesis space for  $\mathcal{L}$ . Then, there is an IIM  $M \text{ LimTxt}_{\mathcal{H}}$ -identifying  $\mathcal{L}$ .

*Proof.*

Let  $M'$  be an IIM  $\text{LimTxt}_{\mathcal{H}'}$ -identifying  $\mathcal{L}$ .

$M(t_x)$ :

If  $M'(t_x) = ?$  then output “?”.

Otherwise, set  $j = M'(t_x)$  and test for  $k = 0, \dots, x$  whether or not

- $h_j(w) = h'_k(w)$  for all  $w \in \Sigma^*$  with  $|w| \leq x$ .

If such a  $k$  has been found, output the least one, otherwise output “?”.

Verification → **Exercise**

# Finite Learning

## Theorem 2.2.2:

Let  $\mathcal{L} \in \text{FinTxt}$  and let  $\mathcal{H}$  be *any class preserving* hypothesis space for  $\mathcal{L}$ . Then, there is an IIM  $M \text{ LimTxt}_{\mathcal{H}}$ -identifying  $\mathcal{L}$ .

*Proof.*

Let  $\mathcal{L} \in \text{FinTxt}$ . By theorem 2.1.9 there are an indexing  $\mathcal{L} = (L_j)_{j \in \mathbb{N}}$  and a recursively generable family  $(T_j)_{j \in \mathbb{N}}$  of finite sets such that

- for all  $j \in \mathbb{N}$ ,  $T_j \subseteq L_j$
- for all  $j, z \in \mathbb{N}$ , if  $T_j \subseteq L_z$  then  $L_j = L_z$

$M(t_x)$ :

If  $x = 0$  or  $M(t_{x-1}) = \text{"?"}$ , goto (\*). Otherwise output  $M(t_{x-1})$ .

(\*) For  $j = 0, 1, \dots, x$ , generate  $T_j$  and test whether  $T_j \subseteq t_x^+$ .

If no such  $j$  has been found, output "?". Otherwise, let  $\hat{j}$  be the minimal  $j$  and search for a  $j'$  such that  $T_{\hat{j}} \subseteq h_{j'}$ . Output  $j'$ .

Verification  $\rightarrow$  **Exercise**

# Finite Learning

## Theorem 2.2.3:

There is an  $\mathcal{L} \in \text{FinTxt}$  and a class comprising hypothesis space  $\mathcal{H}$  for  $\mathcal{L}$  such that no IIM  $M \text{ FinTxt}_{\mathcal{H}}$ -identifies  $\mathcal{L}$ .

*Proof.*

$\mathcal{L} = (L_j)_{j \in \mathbb{N}}$  with  $L_j = \{a^j\}$ . Clearly,  $\mathcal{L} \in \text{FinTxt}$ .

Define  $\mathcal{H}$  as follows:

$$h_{\langle k, x \rangle} = \begin{cases} \{a^k\} & : \phi_k(k) = x \\ \{a^k, b^{\phi_k(k)}\} & : \varphi_k(k) \downarrow \text{ and } \phi_k(k) \neq x \\ \{a^k\} & : \text{otherwise} \end{cases}$$

An IIM  $M \text{ FinTxt}_{\mathcal{H}}$ -identifying  $\mathcal{L}$  could be used to solve the halting problem:

On input  $k$  do:

Feed the text  $a^k, a^k, a^k, \dots$  to  $M$  until it outputs a hypothesis of form  $\langle k, x \rangle$ .

If  $\phi_k(k) = x$ , then output 1, otherwise output 0.

Verification  $\rightarrow$  **Exercise**

# Conservative Learning

## Theorem 2.2.4:

There is an  $\mathcal{L}$  which can be conservatively learned, but only if the hypothesis space used is *class comprising*.

## Theorem 2.2.5:

There is an  $\mathcal{L}$  for which

- there exists a class preserving hypothesis space  $\mathcal{H}$  and an IIM  $M$ , such that  $M$   $\text{ConsvTxt}_{\mathcal{H}}$ -identifies  $\mathcal{L}$
- there exists a class preserving hypothesis space  $\mathcal{H}'$  such that no IIM  $M$   $\text{ConsvTxt}_{\mathcal{H}'}$ -identifies  $\mathcal{L}$

proofs: see [2]

# Summary

For learning in the limit:

- *exact, class preserving, class comprising, absolute class preserving, and absolute class comprising learning are of the same power*

For conservative learning:

- *absolute class preserving learning  $\subset$  class preserving learning  $\subset$  class comprising learning*

For finite learning:

- *absolute class preserving, class preserving, and class comprising learning are of the same power*
- *absolute class comprising learning  $\subset$  class comprising learning*



# Changelog

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- V1.1:
  - Folie 5:  $\mathcal{L} \in \text{LimTxt} \rightarrow \mathcal{L} \in \text{FinTxt}$