# Polynomial-Time Algorithms for Learning Typed Pattern Languages

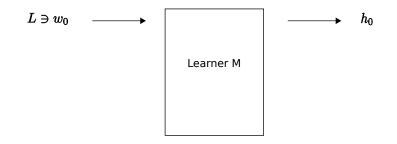
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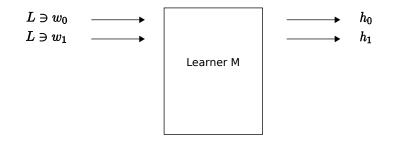
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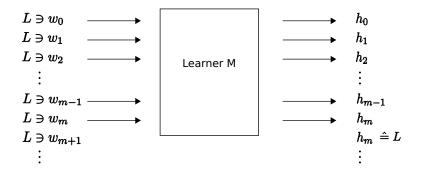
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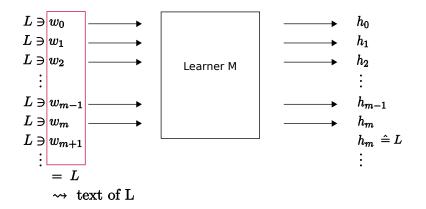
March 08, 2012











# Pattern Languages

 $\Sigma = \{a, b, ...\}$  be a finite set of **terminal symbols** with  $|\Sigma| \ge 2$  $X = \{x_1, x_2, ...\}$  be a countable set of **variables** such that  $\Sigma \cap X = \emptyset$ 

### Informal definition (Angluin)

A **pattern** is any finite string over terminal symbols and variables. The **language of a pattern** p is the set of all words that result from substituting all variables in p by strings of terminal symbols.

#### Example

$$\Sigma = \{a, b, c\}$$

$$p = (ab)^3 x_1 x_2 b^2 c^4 x_3 b^3$$

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$$\Sigma = \{a, b, c\}$$

$$heta(p)=(ab)^3$$
  $a^4$  ba  $b^2c^4$   $c^3$   $b^3$ 

# Typed Pattern Languages

Bibliographic data entry system:

Author:  $x_1$ , Title:  $x_2$ , Year:  $x_3$ 

#### Introduction of types

Each variable has exactly one type:  $\mathcal{T} := \{t_1, t_2\}$ 

$$L_{t_1} = \Sigma^+, \qquad X_{t_1} := \{x_1, x_2\}, \\ L_{t_2} = \{1900, \dots, 2100\} \cup \{\epsilon\} \qquad X_{t_2} := \{x_3\}$$

Learning Pattern Languages Efficiently - Problems

#### The membership problem

*Given:* pattern *p*, word *w Question:* does *p* generate *w*?

### Theorem (Angluin)

The membership problem for the class of untyped pattern languages is NP-complete.

 $\rightsquigarrow$  avoid membership tests during the learning process

# Learning Untyped Pattern Languages (1)

## Theorem (Lange and Wiehagen)

The class of untyped pattern languages as introduced by Angluin can be learned in polynomial time.

Idea: Only take words of shortest length to infer the pattern.

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# Learning Untyped Pattern Languages (2)

### Why is the set of shortest words sufficient?

There is a subset  $S_p$  of L(p) with  $|S_p| \le 2|p|$  such that  $S_p$  is a characteristic set with respect to the set of untyped pattern languages.

а	<i>x</i> <sub>1</sub>	<i>x</i> <sub>2</sub>	bb	<i>x</i> <sub>1</sub>	ab
а	а	а	bb	а	ab
а	b	а	bb	b	ab
а	а	а	bb	а	ab
а	а	b	bb	а	ab

→ de la Higuera's characteristic sets

# Learning Typed Pattern Languages (1)

For typed pattern languages this does no longer work in general!

#### Example

•  $\Sigma := \{a, b\}, t_1 := \{a, b\} \text{ and } t_2 := \{aa, ab, ba, bb, aaa, bbb\}$ 

• 
$$p := x_{(t_1,1)} x_{(t_2,1)}$$

• 
$$q := x_{(t_1,1)} x_{(t_1,2)} x_{(t_1,3)}$$

L(p) and L(q) have the same set of shortest words,  $S := \{aaa, aab, aba, abb, baa, bab, bba, bbb\}$ , but

$$L(p) = S \cup \{aaaa, baaa, abbb, bbbb\} \neq S = L(q).$$

# Learning Typed Pattern Languages (2)

### Type Witnesses

Let  $\mathcal T$  be a set of subsets of  $\Sigma^+$ .  $(\omega_1, \omega_2)$  is a **type witness** for  $\mathcal T$  if

• 
$$\omega_1, \omega_2 : \mathcal{T} \to \Sigma^+$$
 are mappings

- $\omega_1(t) \neq \omega_2(t)$  and  $\{\omega_1(t), \omega_2(t)\} \subseteq t \setminus \bigcup_{t' \in \mathcal{T} \setminus \{t\}} t'$  for all  $t \in \mathcal{T}$
- and some technical conditions are fulfilled

### Example (details omitted)

 $\mathcal{T} = \{t_1, t_2, t_3\}$  with

- *t*<sub>1</sub>: positive integers
- *t*<sub>2</sub>: floats

• *t*<sub>3</sub>: text

# Learning Typed Pattern Languages (3)

To infer the pattern, use words that result from substitutions that replace all variables by their type witnesses.

#### Terminal-free patterns

The properties of a type witness allow us to **decompose** words into type witnesses in **polynomial time**:

- words are processed from left to right
- a prefix of the remainder of the word is matched to a type witness



# Results (1)

#### Theorem

Let  $\mathcal{T}$  be a finite set of decidable subsets of  $\Sigma^+$  that has a type witness. Then the class of all non-erasing  $\mathcal{T}$ -typed pattern languages that are generated by terminal-free patterns is polynomially learnable from positive data.

### Sketch of Algorithm

- decompose words into type witnesses
- elect words with shortest decomposition
- use decompositions to infer the pattern

# Results (2)

#### Theorem

Let  $\mathcal{T}$  be a finite set of decidable subsets of  $\Sigma^+$  that has a short type witness. Then the class of all non-erasing  $\mathcal{T}$ -typed pattern languages is polynomially learnable from positive data.

#### More results

- some classes with **infinite** sets of types are polynomially learnable from positive data
- some classes of typed pattern languages are also polynomially learnable from positive data by a **consistent** learning algorithm