# Symbolic Regression with Interaction-Transformation

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# **Topics**

- 1. Goals
- 2. Interaction-Transformation
- 3. Experiments
- 4. Conclusions

# **Goals**

### Find the function

Find a function f that minimizes the approximation error:

$$\begin{aligned} & \underset{\hat{f}(\mathbf{x})}{\text{minimize}} & & \|\epsilon\|^2 \\ & \text{subject to} & & \hat{f}(\mathbf{x}) = f(\mathbf{x}) + \epsilon. \end{aligned}$$

# Simple is better

- Ideally this function should be as simple as possible.
- Conflict of interests:
  - minimize approximation (use universal approximators)
  - maximize simplicity (walk away from generic approximators)

# The good...

The Linear Regression:

$$\hat{f}(\mathbf{x}, \mathbf{w}) = \mathbf{w} \cdot \mathbf{x}.$$

- Very simple (and yet useful) model.
- Clear interpretation
- The variables may be non-linear transformation of the original variables.

...the bad...

The mean:

$$\hat{f} = \overline{f}(x)$$

- The average can lie!
- It's just for the sake of the pun

# ... the Deep Learning

A deep chaining of non-linear transformations that just works! fig

- Universal approximation
- Alien mathematical expression

# The best?

Despite its success with error minimization, it raises some questions:

- What does the answer mean?
- What if the data is wrong?

# **Symbolic Regression**

Searches for a function form and the correct parameters.

Hopefully a simple function

# Symbolic Regression

**Disclaimer:** I have large experience with evolutionary algorithms, but limited with Symbolic Regression. I have start studying that last year.

# Symbolic Regression

This was my first experience with GP:

$$+ -0.00* id(x_1^-4.0*x_2^3.0*x_3^2.0) - 0.01* id(x_1^-4.0*x_2^3.0*x_3^3.0)$$

$$-0.02* id(x_1^-4.0*x_2^3.0*x_3^4.0) + 0.01* cos(x_1^-3.0*x_2^-1.0) +$$

$$0.01* cos(x_1^-3.0) + 0.01* cos(x_1^-3.0*x_3^1.0) + 0.01* cos(x_1^-3.0*x_2^1.0)$$

$$+ 0.01* cos(x_1^-2.0*x_2^-2.0) - 0.06* log(x_1^-2.0*x_2^-2.0)$$

$$+ 0.01* cos(x_1^-2.0*x_2^-1.0) + 0.01* cos(x_1^-2.0*x_2^-1.0*x_3^1.0)$$

$$+ 0.01* cos(x_1^-2.0*x_3^-.0) + 0.01* cos(x_1^-2.0*x_3^1.0)$$

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$$+ 0.01* cos(x_1^-2.0*x_3^1.0) + 0.01* cos(x_1^-2.0*x_2^1.0)$$

 $6.379515826309025e - 3 + -0.00 * id(x_1^-4.0 * x_2^3.0 * x_3^1.0)$ 

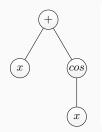
# Why?

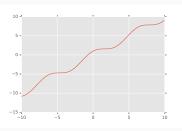
- Infinite search space
- Redundancy
- Rugged

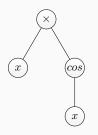
# Redundancy

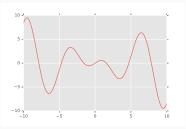
$$f(x) = \frac{x^3}{6} + \frac{x^5}{120} + \frac{x^7}{5040}$$
$$f(x) = \frac{16x(\pi - x)}{5\pi^2 - 4x(\pi - x)}$$
$$f(x) = \sin(x).$$

# Rugged space









#### What I wanted

- A few additive terms (linear regression of transformed variables)
- Each term with as an interaction of a couple of variables
- Maximum of one non-linear function applied to every interaction (no chainings)

Constrains the search space to what I want: a linear combination of the application of different transformation functions on interactions of the original variables.

Essentially, this pattern:

$$\hat{f}(x) = \sum_{i} w_{i} \cdot t_{i}(p_{i}(x))$$

$$p_{i}(x) = \prod_{j=1}^{d} x_{j}^{k_{j}}$$

$$t_{i} = \{id, \sin, \cos, \tan, \sqrt{\log}, \dots\}$$

# Valid expressions:

■ 
$$5.1 \cdot x_1 + 0.2 \cdot x_2$$

$$3.5 \sin \left(x_1^2 \cdot x_2\right) + 5 \log \left(x_2^3 / x_1\right)$$

### Invalid expressions:

- $tanh(tanh(tanh(w \cdot x)))$
- $\sin(x_1^2 + x_2)/x_3$

We can control the complexity of the expression by limiting the number of additive terms and the number of interactions:

$$\hat{f}(x) = \sum_{i=1}^{k} w_i \cdot t_i(p_i(x))$$

$$p_i(x) = \prod_{j=1}^{d} x_j^{k_j}$$

$$s.t.|\{k_j \mid k_j \neq 0\}| \leq n$$

Describing as an Algebraic Data Type can help us generalize to other tasks:

```
IT x = 0 | Weight (Term x) `add` (IT x)

Term x = Trans (Inter x)

Trans = a -> a

Inter x:xs = 1 | x s `mul` Inter xs
```

The meaning of add and mul can lead us to boolean expressions, decision trees, program synthesis.

# **SymTree**

```
Simple search heuristic:
```

# **SymTree**

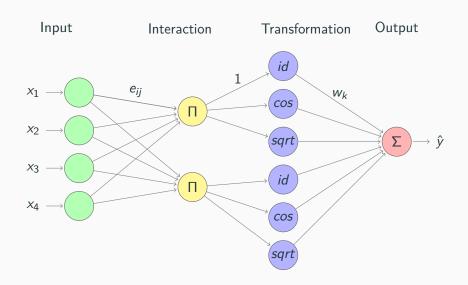
```
expand leaf = expand' leaf terms
  where terms = interaction leaf U transformation leaf

expand' leaf terms = node : expand' leaf leftover
  where (node, leftover) = greedySearch leaf terms
```

#### **IT-ELM**

Interaction-Transformation Extreme Learning Machine, it generates lots of random interactions, enumerates the transformations for each interaction and then adjust the weight of the terms using  $l_0$  or  $l_1$  regularization.

# **IT-ELM**



# **Experiments**

### Data sets

Features	5-Fold / Train-Test
5	5-Fold
8	5-Fold
7	5-Fold
8	5-Fold
8	5-Fold
25	5-Fold
11	5-Fold
11	5-Fold
6	5-Fold
57	Train-Test
3	Train-Test
3	Train-Test
25	Train-Test
	5 8 7 8 8 25 11 11 6 57 3

#### Methods

#### For the sets with folds:

- Each algorithm was run 6 times per fold and the median of the RMSE of the test set is reported
- SymTree was run 1 time per fold (deterministic)

#### For the sets with train-test split:

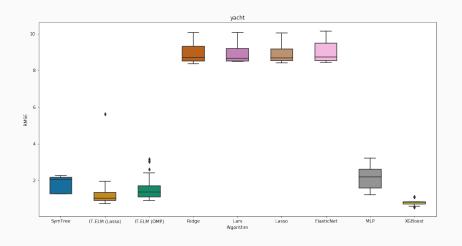
- Each algorithm was run 10 times and the median of the RMSE for the test set is reported
- SymTree was run 1 time per data set

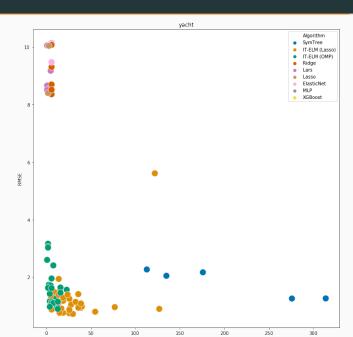
For a complete table:

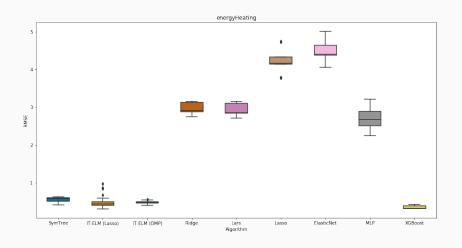
Binder

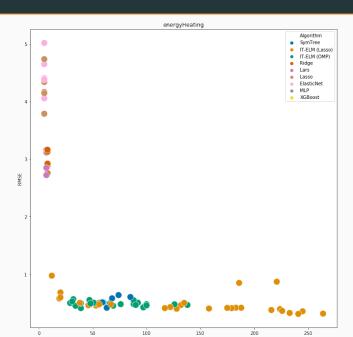
 $\mathsf{Cell} \to \mathsf{Run} \; \mathsf{All}$ 

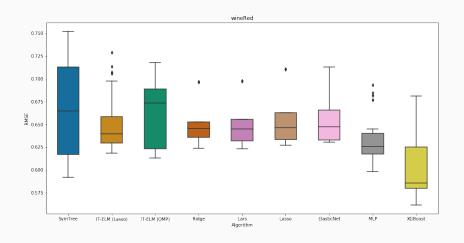
	Rank
Algorithm	
XGBoost	2
SymTree	3.3
IT-ELM (Lasso)	3.53846
IT-ELM (OMP)	3.84615
MLP	5
Ridge	5.69231
Lars	5.76923
Lasso	6.76923
<b>ElasticNet</b>	7.76923

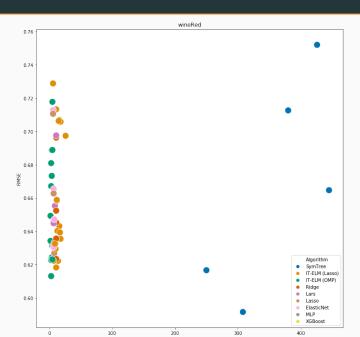












# Sample equations

F05128-f1:

$$-8.3 \times 10^{-3} \log \left( p^{2} t^{2} v + 1 \right)$$

$$+1 \times 10^{-3} \sin \left( t v^{2} \right)$$

$$+3.1 \times 10^{-3} \cos \left( p \right)$$

$$+8.7 \times 10^{-3} \cos \left( p^{2} t v^{2} \right)$$

$$-4.06 \cdot 10^{-5} \tan \left( p^{2} t^{2} v^{2} \right)$$

# Sample equations

CPU:

$$4.9\times10^{-4}\cdot\textit{maxMem}\sqrt{\textit{repPerf}}$$

# **Conclusions**

#### Resumo

A representação Interação-Transformação permite definir um espaço de busca de expressões matemáticas simples mas capaz de aproximar diversas bases de dados, sendo competitivo com algoritmos do estado-da-arte de regressão.

Além disso, o algoritmo SymTree é capaz de encontrar uma boa expressão IT com poucas iterações, sendo um algoritmo simples e computacionalmente leve.

#### **Future research**

#### Muitas possibilidades de estudos futuros:

- Generalizar a representação como uma tipo de dado algébrico
- Utilizar essa representação em outros contextos
- Aumentar o espaço de busca permitindo outras expressões simples ainda não compreendidas
- Criar novos algoritmos de busca para esse espaço de busca
- Muitos outros. . .

# Try it!

You can try a lightweight version of  $\ensuremath{\mathsf{SymTree}}$  at:

https://galdeia.github.io/

It works even on midrange Smartphones!