# When GNNs Met a Word Equations Solver: Learning to Rank Equations

FroCoS 2025

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## THE CONFERENCE MORNING SESSION



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#### **Flements**

- ullet Empty string:  $\epsilon$
- Letters: *a*, *b*, ...
- Variables: *X*, *Y*, *Z*, ...

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,  $Y = a$ ,  $Z = bb$ 

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## Easy, right...?

## $AaAkjhhohhhkokookkkookkhkoohohCAYZjhBkjhkkkokhkkEkkoWQB \\ = aABBBncYECaa$

• Variables: *A*, *C*, *Y*, *Z*, *B*, *E*, *W*, *Q* 

• Letters: *a, k, j, h, o, n, c* 

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Z3, cvc5, Ostrich, Z3-noodler, and Woorpje cannot solve it in 300 seconds

## Challenges

- Difficult to solve
  - ▶ Decidable
  - ▷ Complexity: NP-Hard and in PSPACE
  - ▶ Practical implementation are incomplete
- Important in modeling string constraints in verification tasks
  - ▷ User inputs, string manipulations, ...

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Can machine learning help to solve conjunctive word equation problem?

• Is it another "we used ML for [insert problem] and..."?

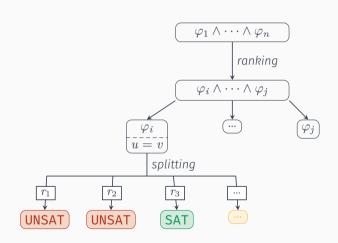
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  - ▶ Yes, but not only! (We also used our brains!)

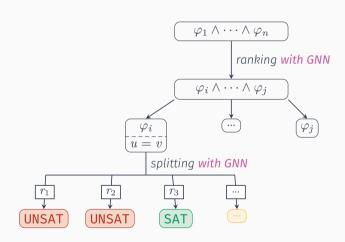
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  - ▷ It depends on the human
  - ▷ Obviously not!



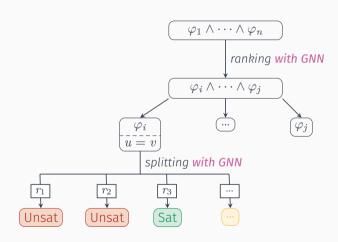


## Magichine Learning

#### Graph Neural Networks (GNN)

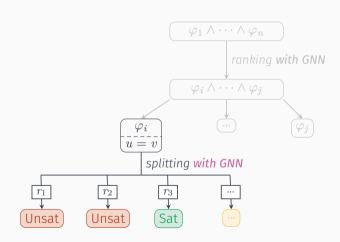
- Designed to work directly with graph-structured data
- Take graph as input, output node and graph representations
- Capture the structural features of graph

## **Big Picture**



g

## **Big Picture**



g

#### Calculus

- Nielsen transformation
- Splitting rules, iterative deepening & backtracking
- Continuously simplify the first terms of both sides
- DragonLi



$$Xabb = YaZ$$

<sup>&</sup>lt;sup>1</sup>Abdulla, Atig, Cailler, Liang and Rüemmer. *Guiding word equation solving using graph neural networks*. International Symposium on Automated Technology for Verification and Analysis (ATVA 2024).

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$$\downarrow r_1 \qquad \downarrow r_2 \qquad \downarrow r_3$$

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$$l \mid u \mid = \mid l' \mid v \mid$$

where l, l', a and b are a letters, u and v are terms, X and Y are variables, and X' and Y' are fresh variables.

$$u = v$$

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$$u = v$$

$$\begin{vmatrix} l & u & = & |l'| & v \\ \downarrow l & = & l' & = & |l'| & v \\ \downarrow l & = & |l'| & = & |l'| &$$

$$u = v$$

$$v = v$$

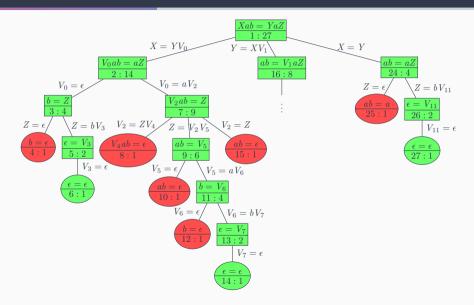
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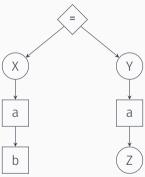
$$x =$$

### Here Is a Proof...

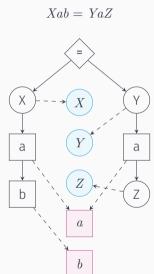


## ...and Here Is an Equation!

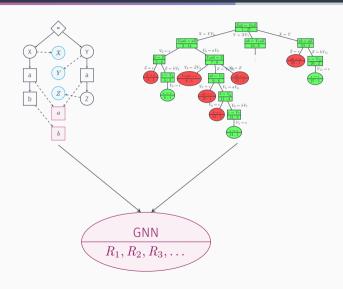




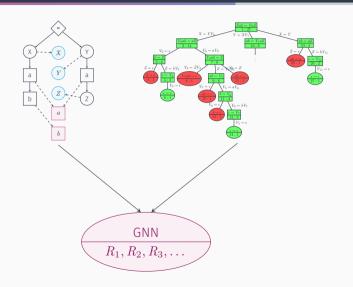
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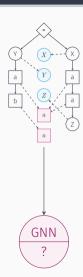


# Better Together!



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# DragonLi, I Choose You!

- Python
- Focus on SAT
- Results comparable to state-of-the-art solvers



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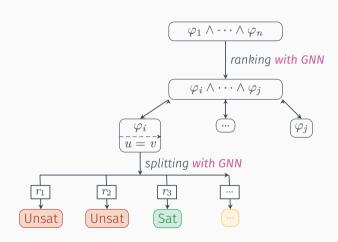
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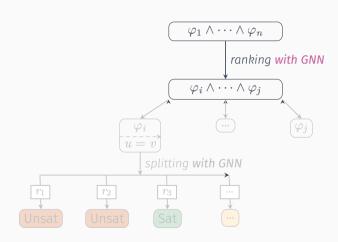
How to deal with more than one equation?

How to improve on **UNSAT** problems?

## **Big Picture**



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 $AaAkjhhooohohCA\,YZjhBkjhkkkokhkkEk = aABBBnc\,YECaa \wedge a = b$ 

 $AaAkjhhooohohCAYZjhBkjhkkkokhkkEk = aABBBncYECaa \land a = b$ 

- Left first: Timeout
- Right first: UNSAT

#### $AaAkjhhooohohCAYZjhBkjhkkkokhkkEk = aABBBncYECaa \land a = b$

- Left first: Timeout
- Right first: UNSAT
- Basic heuristic (+ length):
  - 1.  $\epsilon = \epsilon$
  - 2.  $\epsilon = u \cdot v \text{ or } u \cdot v = \epsilon$
  - 3.  $a \cdot u = b \cdot v \text{ or } u \cdot a = v \cdot b$
  - 4.  $a \cdot u = a \cdot v$
  - 5. Otherwise

#### $AaAkjhhooohohCAYZjhBkjhkkkokhkkEk = aABBBncYECaa \land a = b$

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Can a GNN do better?

## **Cherry Picking**

#### Minimal Unsatisfiable Set (MUS)

A set of equations U of a conjunctive word equation is a MUS iff the conjunction of U is unsatisfiable, and for all equations  $e \in U$ , the conjunction of  $U \setminus \{e\}$  is satisfiable.

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$$a = X \land b = Y \land aa = Xa \land aab = XaY \land X = b$$
 
$$\{a = X, X = b\}$$

### How to (m)Use Them?

### **Training Data**

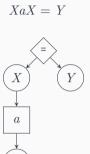
- Problems not solved by DragonLi are given to Z3, Z3-Noodler, cvc5, Ostrich and Woorpje
- Extraction of MUS

#### Labelling

For a conjunctive word equation  $\phi = e_1 \wedge \cdots \wedge e_n$  and U its MUS:

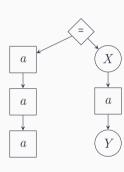
$$\mathsf{label}(e_i) = \begin{cases} 1 & \text{if } e_i \in U \text{ and } |e_i| = \min \big(\{|e| \mid e \in U\}\big) \\ 0 & \text{otherwise}. \end{cases}$$

## I Wouldn't Count on It...

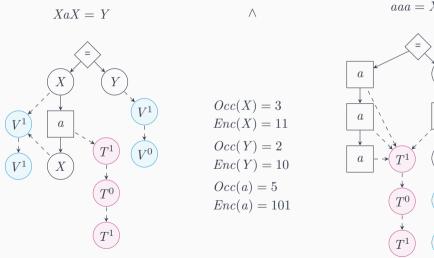








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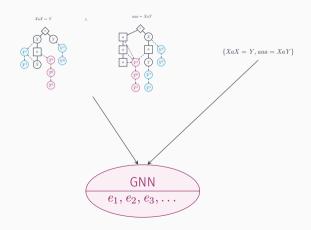
a

 $V^1$ 

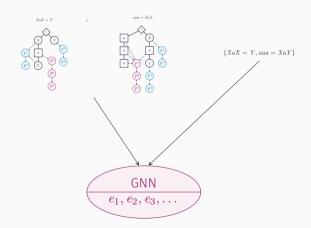
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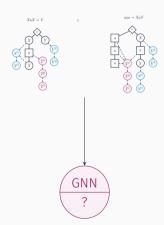
 $V^1$ 

# Better Together (bis repetita)



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• Why a second GNN?

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  - ▷ Different task
  - ▷ Varying size of the input

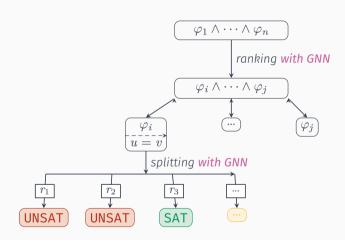
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- When do you call the GNN?
  - ▷ Once at the beginning with 5 equation of priority level 5



### I Need Problems...

- Real-world benchmarks:
  - ⊳ SMT-LIB and Zaligvinder
  - ▶ Remove length constraint, boolean operator, and regex
  - ▶ Too easy

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- Real-world benchmarks:
  - ▷ SMT-LIB and Zaligvinder
  - ▷ Remove length constraint, boolean operator, and regex
  - ▶ Too easy
- Artificial benchmarks:
  - $\triangleright s = s$
  - $\triangleright$  At most 60 letters (among a set T)
  - $\triangleright$  Replace n times substrings on both side by m variables
  - ▶ Between 0 and 100 equations

## Linearity and Benchmarks

#### Linearity

A conjunctive word equation is called *linear* iff each variable appear at most once in each equation.

$$X = a \wedge Y = b$$
  $\checkmark$   $X = a \wedge X = b$ 



$$X = a \wedge X = b$$

$$XaX = aaa \land YaX = YbZ$$



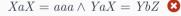
# Linearity and Benchmarks

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A conjunctive word equation is called *linear* iff each variable appear at most once in each equation.

$$X = a \land Y = b$$

$$X = a \land Y = b \quad \checkmark \qquad \qquad X = a \land X = b \quad \checkmark$$





- A1: fresh variable, T = 6,  $n \in [0, 5]$ ,  $m \in [0, 5]$  (linear)
- A2 : A1 + T = 26,  $n \in [0, 16]$ , and m = 1 (linear)
- B : A1 + No fresh variable (non-linear)
- C:  $X_n a X_n b X_{n-1} \cdots b X_1 = a X_n X_{n-1} X_{n-1} b \cdots X_1 X_1 b a a + b \mapsto A_1$  (highly non-linear)

### **Experimental Setup**

- DragonLi: basic, random and GNN-based
- Z3, Z3-Noodler, cvc5, Ostrich, and Woopje
- 300 seconds timeout
- 2 Intel Xeon E5 2630 v4 at 2.20 GHz/core and 128GB memory
- GNNs trained on NVIDIA A100 GPUs
- 60 000 problems per benchmark for training
- 1000 problems per benchmark for experiments

## Results — A1

Solver	Nu	mber of s	olved	proble	ems	Average solving time (split number)				
	SAT	UNSAT	UNI	CS	CU	SAT	UNSAT	CS	CU	
DragonLi	24	955	0		678	5.6 (244.8)	6.5 (1085.3)	5.0 (94.4)	5.7 (126.3)	
Random-DragonLi	22	944	0			5.6 (198.8)	6.3 (932.6)	5.6 (137.6)	5.7 (180.5)	
GNN-DragonLi	24	961	0			6.1 (164.7)	7.5 (1974.8)	6.1 (96.4)	6.3 <b>(60.5)</b>	
cvc5	24	952	1	13		0.5	0.6	0.1	0.3	
Z3	17	960	0			8.7		1.1	0.1	
Z3-Noodler	22	939	2			5.7	0.3	4.8	0.1	
Ostrich	17	931	0			15.0	5.5	8.0	4.7	
Woorpje	23	744	0			3.0	12.5	0.1	12.2	

## Results — A2

Solver	Nun	nber of so	olved p	roble	ms	Average solving time (split number)					
	SAT	UNSAT	UNI	CS	CU	SAT	UNSAT	CS	CU		
DragonLi	59	824	0			8.5 (4233.4)	11.8 (1231.3)	4.7 (27.3)	-		
Random-DragonLi	44	806	1	3 0		24.7 (29779.6)	6.2 (210.9)	4.6 (27.3)	-		
GNN-DragonLi	59	836	4			8.4 (1330.6)	11.6 (1074.1)	5.9 (27.3)	-		
cvc5	67	142	15			0.6	56.0	0.1	-		
Z3	8	870	10			1.1	0.6	0.1	-		
Z3-Noodler	22	7	1			15.4	3.8	0.4	-		
Ostrich	13	18	2			24.8	38.8	8.6	-		
Woorpje	0	0	0	-	-	-	-	-	-		

## Results — B

Solver	Nui	mber of s	olved p	oroble	ems	Average solving time (split number)				
	SAT	UNSAT	UNI	CS	CU	SAT	UNSAT	CS	CU	
DragonLi	11	805	0			4.9 (62.5)	5.2 (81.5)	4.9 (29.2)	5.3 (82.4)	
Random-DragonLi	10	894	0			5.0 (58.7)	5.8 (295.2)	5.0 (27.25)	5.2 (73.1)	
GNN-DragonLi	11	821	0	4	294	6.5 (65.1)	6.8 (70.0)	6.5 (28.25)	6.8 (60.2)	
cvc5	12	915	0			0.1	0.6	0.1	0.7	
Z3	11	859	3			0.1	0.1 0.2 0.1	0.1	0.1	
Z3-Noodler	24	911	1			4.9		1.3	0.4	
Ostrich	12	917	2			6.9	3.7	3.3	4.2	
Woorpje	19	330	1			29.5	6.0	0.2	5.0	
	3	62	0	1		65.0	28.4	0.2	223.1	

## Results — C

Solver	Nun	nber of so	olved p	roble	ms	Average solving time (split number)			
	SAT	UNSAT	UNI	CS	CU	SAT	UNSAT	CS	CU
DragonLi	2	0	0	-	-	5.1 (85.5)	-	-	-
Random-DragonLi	2	0	0	-	-	5.0 (85.5)	-	-	-
GNN-DragonLi	-	-	-	-	-	-	-	-	-
cvc5	0	909	17	-		-	46.9	-	17.3
Z3	1	821	12	1		0.8	1.7	0.8	0.1
Z3-Noodler	7	657	4	1	1	0.2	94.1	0.1	1.0
Ostrich	0	61	0	-		-	77.2	-	27.1
Woorpje	3	62	0	1		65.0	28.4	0.2	223.1

### "Why Is Your Work Good?"

- DragonLi performs well on linear problems
- Fewer splits
- No improvement on non-linear problems
- Need for dedicated strategies (equation length, letter count, ...)
- Re-implementation can offer improvements
- Increasing training dataset increases the number of problems solved

### This is the End!

#### Contributions

- Extension of **DragonLi** to handle conjunctive word equations
- Multi-classification task to handle variable number of inputs
- MUS & new graph representation
- Good results on linear problems

#### What's Next?

- Optimize GNN overhead
- Call timing
- Length constraint and regular expressions

### Thank you! 😉

https://github.com/ChenchengLiang/DragonLi

