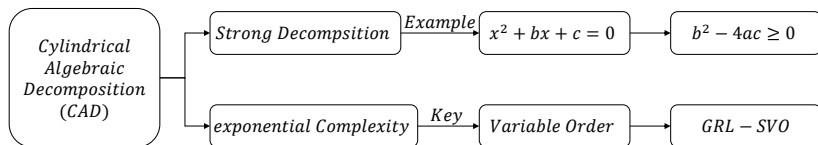


Suggesting Variable Order for Cylindrical Algebraic Decomposition via Reinforcement Learning

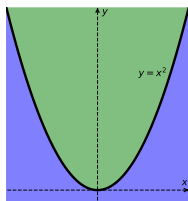
Fuqi Jia*, Yuhang Dong*, Minghao Liu, Pei Huang, Feifei Ma,
Jian Zhang

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Cylindrical Algebraic Decomposition (CAD)



(a) 3 cells

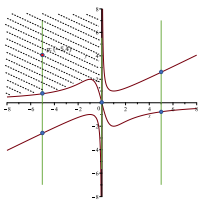
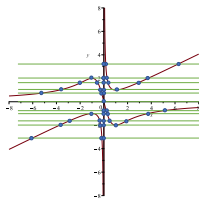
(b) $x < y$: 13 cells(c) $y < x$: 89 cells

Figure 1: Examples of CAD. Figure 1a shows cells of $\{y - x^2\}$. Figure 1b and 1c are different variable orders on $\{x^3y + 4x^2 + xy, -x^2 + 2xy - 1\}$.

Existing Heuristics

Table 1: Classification on Existing Heuristics

	EB (Expert-Based)	LB (Learning-Based)
UP (Utilizing <i>project</i>)	<i>sotd</i> [7], <i>ndrr</i> [8], <i>gmods</i> [9]	<i>GRL-SVO(UP)</i>
NUP (Not Utilizing <i>project</i>)	<i>brown</i> [10], <i>triangular</i> [11], <i>chord</i> [12]	<i>EMLP</i> [13], <i>PVO</i> [14], <i>GRL-SVO(NUP)</i>

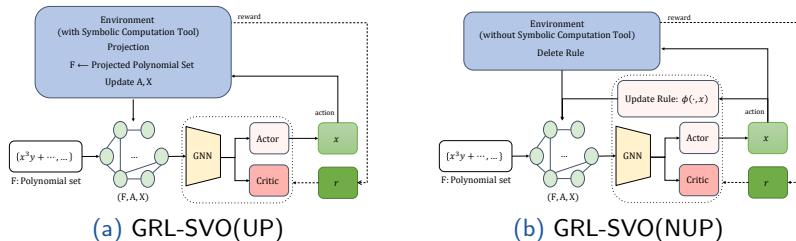
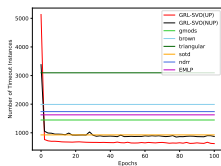
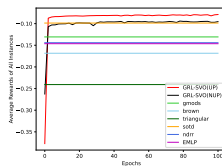


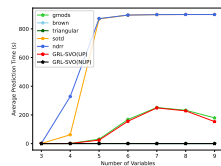
Figure 2: The architecture of GRL-SVO(UP) and GRL-SVO(NUP) where $\phi(\cdot, x) = \text{MLP}(\text{CONCAT}(\cdot, x))$ for updating the embedding for the neighbours of x . The dashed lines represent that it will be only utilized in training mode.



(a)



(b)



(c)

Figure 3: The performance of all heuristics. Figure 3a and 3b correspond to the training phase, and the horizontal lines represent the timeout instances of corresponding heuristics on the training set. Figure 3c is the *PT* graph over number of variables.

Table 2: The performance of all heuristics. The dash “-” indicates that the method does not support the category.

Categories		NUP						UP			
		EB		LB				EB		LB	
		<i>brown</i>	<i>triangular</i>	<i>EMLP</i>	<i>PVO(brown)</i>	<i>PVO(triangular)</i>	<i>GRL-SVO(NUP)</i>	<i>sotd</i>	<i>ndrr</i>	<i>gmods</i>	<i>GRL-SVO(UP)</i>
3-var(test)	<i>#SI</i>	1620	1504	1686	-	-	1772	1784	1663	1693	1798
	<i>AVG.T</i>	171.41	228.32	140.87	-	-	94.87	91.47	148.92	124.06	78.06
	<i>AVG.N</i>	2427.74	2669.67	2390.68	-	-	2166.67	2149.07	2007.98	2195.80	2089.68
4-var	<i>#SI</i>	415	376	-	408	392	443	625	488	513	533
	<i>AVG.T</i>	352.87	394.71	-	360.33	376.71	314.57	85.12	292.06	215.48	191.45
	<i>AVG.N</i>	5241.95	5585.90	-	5323.83	5582.46	5131.40	3925.28	4248.45	4849.18	4764.50
5-var	<i>#SI</i>	236	202	-	242	218	238	43	27	329	346
	<i>AVG.T</i>	434.52	494.37	-	418.34	465.43	420.51	827.47	853.75	238.01	207.79
	<i>AVG.N</i>	12310.70	13224.18	-	11795.90	12555.82	12090.49	14538.69	14845.67	10466.79	9744.58
6-var	<i>#SI</i>	175	149	-	180	160	202	5	5	273	306
	<i>AVG.T</i>	501.75	552.14	-	490.73	527.86	439.62	889.16	889.97	284.40	214.55
	<i>AVG.N</i>	20639.07	20440.23	-	20181.98	19290.37	19302.97	23298.50	23329.10	17561.67	16715.20
7-var	<i>#SI</i>	163	118	-	-	-	153	1	1	270	297
	<i>AVG.T</i>	548.15	631.85	-	-	-	552.47	897.75	897.79	313.73	245.57
	<i>AVG.N</i>	27790.31	27795.79	-	-	-	27302.28	30452.64	30456.31	24465.89	22432.30
8-var	<i>#SI</i>	173	138	-	-	-	172	0	0	310	345
	<i>AVG.T</i>	601.90	654.20	-	-	-	597.09	900.00	900.00	372.34	322.80
	<i>AVG.N</i>	39382.26	40679.57	-	-	-	38815.98	43112.02	43112.02	34016.98	33505.21
9-var	<i>#SI</i>	151	125	-	-	-	158	0	0	286	325
	<i>AVG.T</i>	649.41	690.29	-	-	-	625.69	900.00	900.00	431.11	374.78
	<i>AVG.N</i>	48273.67	49832.78	-	-	-	46946.09	52173.03	52173.03	42594.25	42270.91
SMT-LIB (3-var)	<i>#SI</i>	1770	1763	1675	-	-	1766	1750	1694	1772	1772
	<i>AVG.T</i>	20.33	23.68	83.09	-	-	22.38	34.38	65.10	18.32	18.53
	<i>AVG.N</i>	4449.79	5070.46	7661.07	-	-	4104.43	3672.21	4140.72	3873.22	3946.84
SMT-LIB (4-var to 6-var)	<i>#SI</i>	374	372	-	372	372	364	356	339	379	379
	<i>AVG.T</i>	86.03	89.95	-	88.32	88.98	91.17	105.18	142.32	59.96	67.51
	<i>AVG.N</i>	24596.20	24260.88	-	23039.49	22730.34	21040.09	16896.16	21013.25	17388.52	18894.51
SMT-LIB (7-var to 9-var)	<i>#SI</i>	13	12	-	-	-	12	11	11	16	14
	<i>AVG.T</i>	308.14	377.32	-	-	-	339.90	541.53	588.33	260.53	329.91
	<i>AVG.N</i>	53971.24	58675.94	-	-	-	51570.88	51470.41	62185.82	50381.12	56312.41

Thanks!

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