

# Bipartite Encoding: A New Binary Encoding for Solving Non-Binary CSPs

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## Main Results

1. We introduce a new binary encoding called **Bipartite Encoding (BE)**;
2. An algorithm and heuristic are used to generate *binary* BE instances from non-binary CSPs;
3. A special Arc Consistency (AC) propagator on BE instances;
4. Experiments show that BE with AC can outperform state-of-the-art generalized AC (GAC) propagators and binary encodings.

## Why Binary CSPs with BE?

- Binary & Non-Binary CSPs are NP-complete. Is it better to solve a Non-Binary CSP in original form or by encoding into a Binary CSP?
- BE encoding shows binary encoding of non-binary CSP can be superior: higher consistency & faster propagation

## Bipartite Encoding (BE)

The main idea is that every constraint is partitioned into 2 sub-tables and represented as a conjunction between 2 sub-tables, and then

1. **sub-tables** are encoded as **factor variables**, where the values of factor variables are mapped to tuples in the sub-tables by using binary constraints called **mapping constraints**, and
2. **conjunctions** are encoded as **bipartite constraints**.

$x_1$	$x_2$	$x_3$	$x_4$	$fv_1$	$fv_2$	$x'_1$	$fv_1$	$x'_2$	$fv_1$	$x'_3$	$fv_2$
0	0	0	1	a	b	0	a	0	a	0	a
0	0	1	0	a	c	0	b	1	b	0	b
0	1	0	0	b	a					1	c
1	0	0	0								

$x_1$	$x_2$	$x_5$	$x_6$	$fv_1$	$fv_3$	$x'_5$	$fv_3$	$x'_6$	$fv_3$	$x'_4$	$fv_2$
0	0	0	1	a	b	0	b	1	b	0	a
0	0	1	1	a	d	1	c	0	c	1	b
0	1	1	0	b	c	1	d	1	d	0	c
1	1	0	0								

**Figure 1:** Bipartite encoding (BE) example: The left and right part is the original non-binary CSP and BE instance respectively

**Example 1.** Figure 1 shows a simple example to illustrate BE encoding. In the example, we encode 2 non-binary constraints into 2 bipartite constraints between the factor variables  $\{fv_1, fv_2, fv_3\}$ , and each factor variable is mapped to sub-tables by 2 mapping constraints. The factor variable  $fv_1$  is shared, thus, some tuples in the constraint relations can be removed.

## Consistency level of AC on BE instances

**Proposition 1.** The image  $\mathcal{P}^l$  of a binary encoding  $BE(\mathcal{P})$  is GAC if  $BE(\mathcal{P})$  is AC, where  $\mathcal{P}^l$  is the same as the original non-binary CSP  $\mathcal{P}$  except variable domains and constraint relations may be reduced.

## Generation of BE instances

The method used for constructing BE instances is as follows:

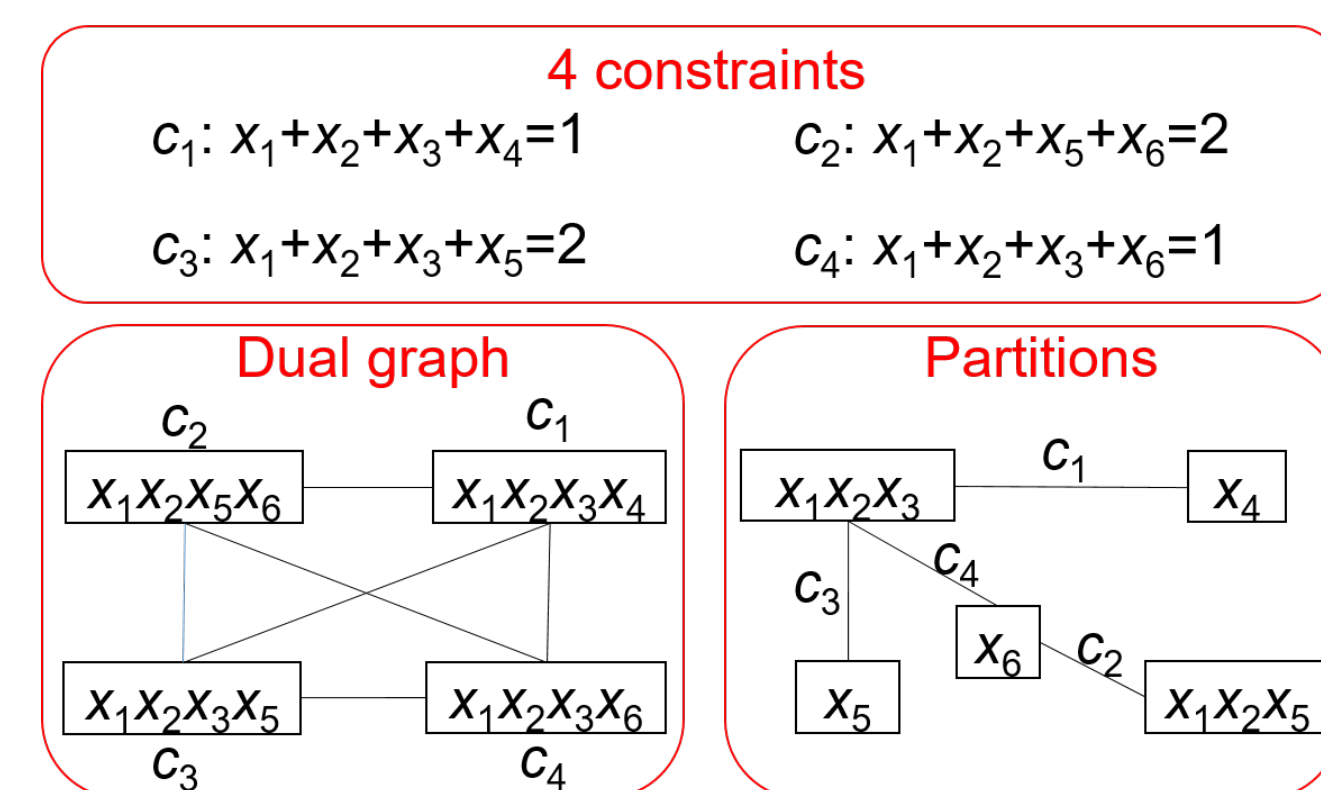
- (i) The scope of each constraint is split into 2 variable subsets.
- (ii) Generating factor variables for the sub-tables over variable subsets.
- (iii) Constructing BE instance based on the generated factor variables and partitions of constraint scopes.

## Constraint scope partition

Two kinds of methods used to partition a constraint scope  $scp(c_1)$ :

1. select an edge  $\{c_1, c_2\}$  in the dual graph such that the size  $|S|$  is greater than 1 where  $S = scp(c_1) \cap scp(c_2)$ , and partition  $scp(c_1)$  and  $scp(c_2)$  into subsets  $S$ ,  $scp(c_1) \setminus S$  and  $scp(c_2) \setminus S$ ;
2. select a variable  $x \in scp(c_1)$  and partition  $scp(c_1)$  into subsets  $\{x\}$ ,  $scp(c_1) \setminus \{x\}$  (this is a basic partition).

Every constraint scope can only be **partitioned once**, meanwhile, there are some partitions such that the BE instances based on the partitions are **not compact**. Therefore we only select some available edges to do partition, and the rest constraints use the basic partitions. **We select the edges with a larger size as soon as possible.**



**Figure 2:** Constraint partition example: The top, bottom left and bottom right part is the non-binary CSP, dual graph and partitions respectively

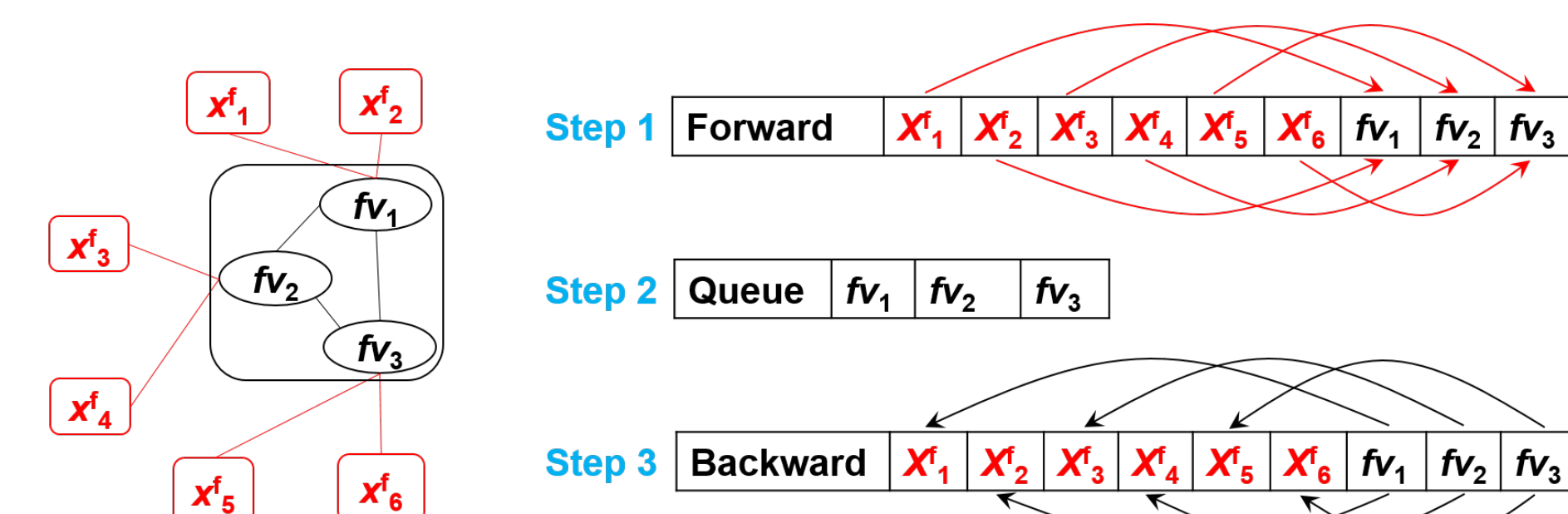
**Example 2.** For the non-binary CSP given in Figure 2, we first select 3 edges from its dual graph  $\{c_1, c_3\}$ ,  $\{c_1, c_4\}$ ,  $\{c_3, c_4\}$ , and then split the constraint scopes  $scp(c_1)$ ,  $scp(c_3)$ ,  $scp(c_4)$  into the subsets  $\{x_1, x_2, x_3\}$ ,  $\{x_4\}$ ,  $\{x_5\}$ ,  $\{x_6\}$ . There is not any available edge for the last constraint  $c_2$ , thus, we use the basic partition.

## AC propagator on BE instances

We follow the algorithm framework used by the HTAC propagator:

1. We first partition the binary constraints  $\mathcal{C}$  into a set of **connected components** in an undirected graph  $(\mathcal{C}, E)$  where  $E = \{\{c_i, c_j\} \subseteq \mathcal{C} \mid scp(c_i) \cap scp(c_j) \text{ includes compound factor variables}\}$ .
2. For each connected component  $com$ , we partition the variables in  $com$  into 2 subsets  $T$  and  $U$  such that the variables in  $T$  are not included by any cycle in the primal graph of  $com$ .
3. For variables in  $T$ , we update domains from leaves to roots (**forward propagation**), and then from roots to leaves (**backward propagation**).

4. For variables in  $U$ , we use a **normal propagation queue**.



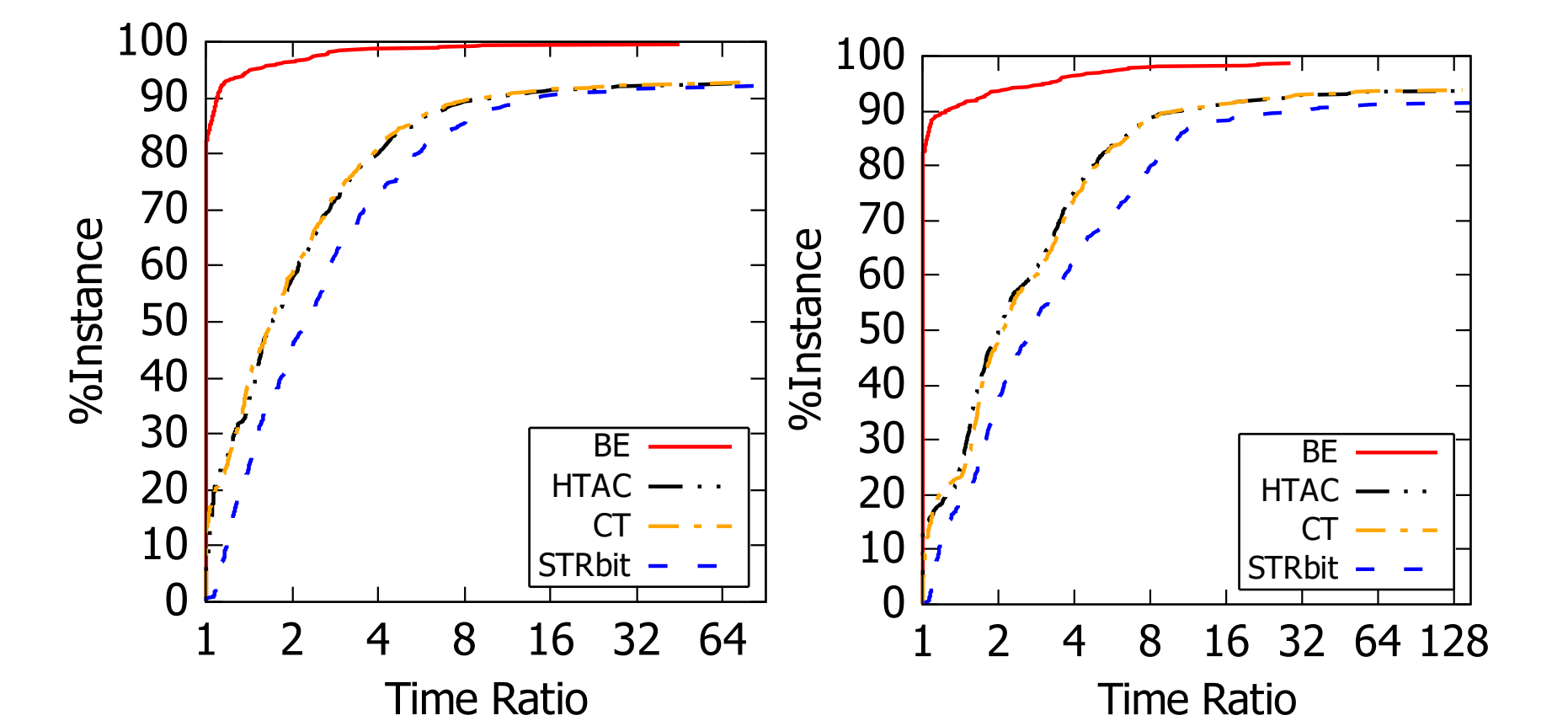
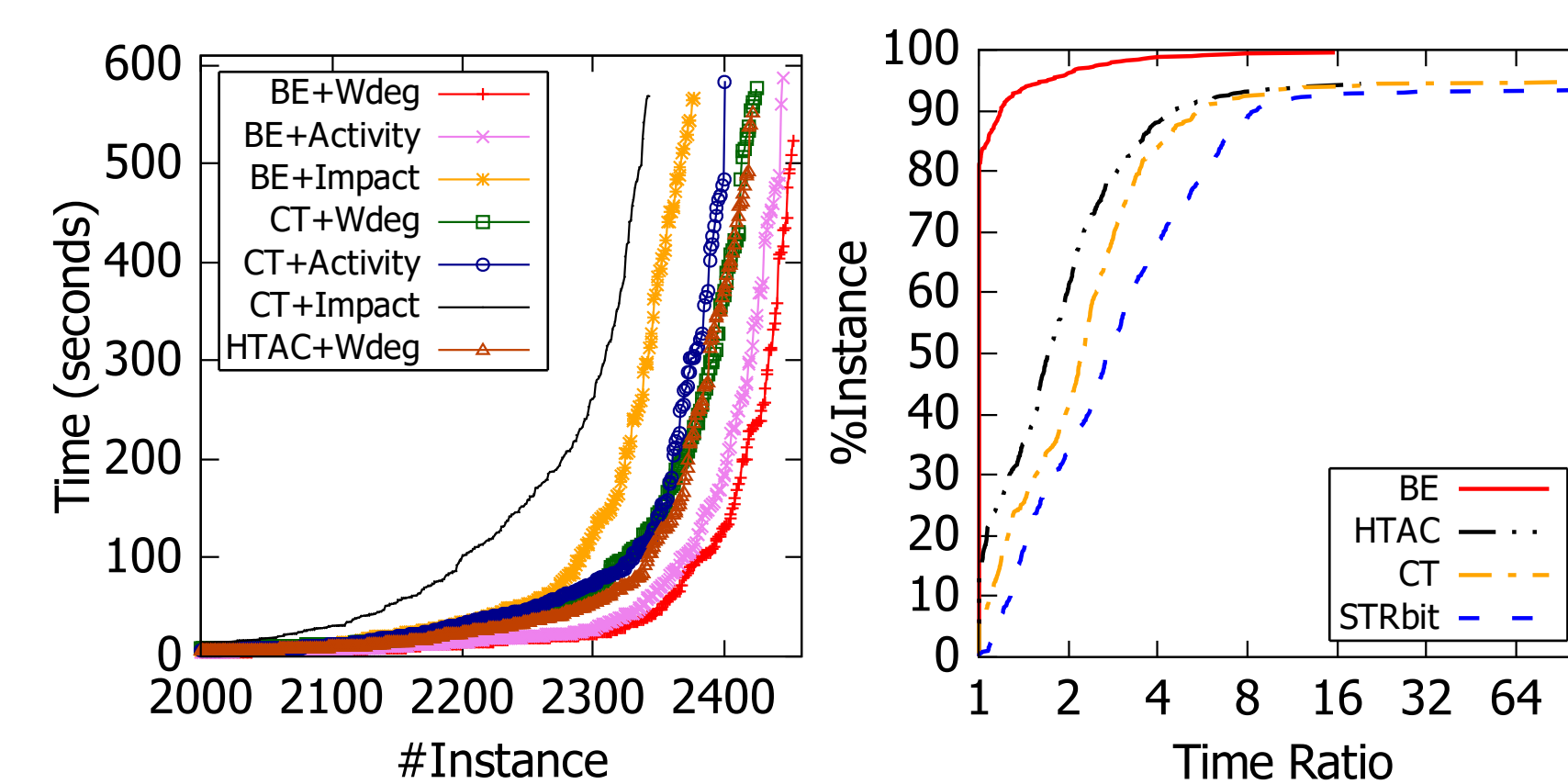
**Figure 3:** Propagation ordering for a component: The left and right part is the component and propagation ordering respectively

**Example 3.** Figure 3 shows the propagation ordering used for enforcing AC on 2 subsets  $T$  and  $U$  of a component, where  $T = \{x_1^f, \dots, x_6^f\}$  and  $U = \{fv_1, fv_2, fv_3\}$ . We first do forward propagation on  $T$ , and then enforce AC on  $U$ , finally do backward propagation on  $T$ .

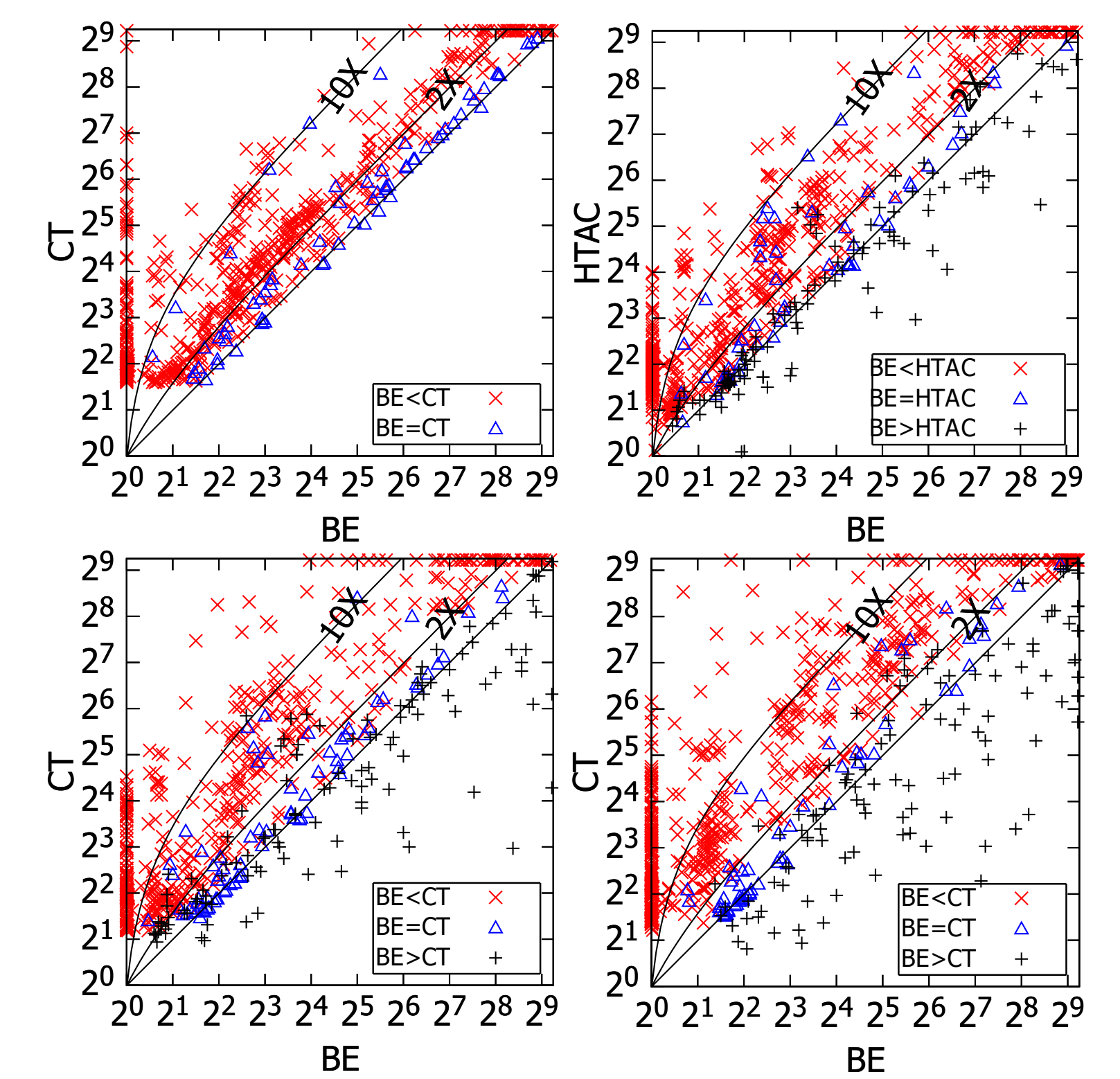
## Experiments

		STRbit	CT	HTAC	BE
Wdeg (682)	AvgR	3.50	2.68	1.99	-
	MaxR	91.46	80.98	19.2	-
	#F	3	37	89	<b>553</b>
	#TO	45	36	38	<b>3</b>
Activity (655)	AvgR	3.58	2.76	2.80	-
	MaxR	90.54	80.01	77.83	-
	#F	4	87	41	<b>539</b>
	#TO	51	47	48	<b>3</b>
Impact (691)	AvgR	4.73	3.52	3.64	-
	MaxR	149.78	129.44	138.61	-
	#F	0	89	33	<b>569</b>
	#TO	59	44	43	<b>9</b>
Initial Time (s)	#BF	0	0	0	<b>175</b>
		1.39	1.37	1.38	1.65

**Table 1:** Relative comparison with Total Times



**Figure 4:** Runtime distribution of total time (top left plot) and Performance profiles comparing algorithms with heuristics: The top right, bottom left and bottom right plot is for the heuristic Wdeg, Activity and Impact respectively



**Figure 5:** Solving time comparison: The top left, top right, bottom left and bottom right plot is for the heuristic Ddeg+O, Wdeg, Activity and Impact respectively. Each dot denotes an instance, the time on the x-axis and y-axis is (1 + solving time) to enable logarithmic scales, "A=B" ("A>B" and "A<B") means the number of search nodes of Algorithm A is within 2% (greater than 1.02X and less than 1.02X) than that of B. The 10X (and 2X) line means BE is 10X (and 2X) faster than CT or HTAC

- Fastest against CT (GAC) & HTAC (AC HVE encoding) across Wdeg, Activity, Impact heuristics
- Higher consistency than GAC on original CSP - many instances become Backtrack Free
- Encoding non-binary CSP to binary CSP with BE encoding and solving with AC-BE propagator gives state-of-art results

