

Improving Relational Consistency Algorithms Using Dynamic Relation Partitioning

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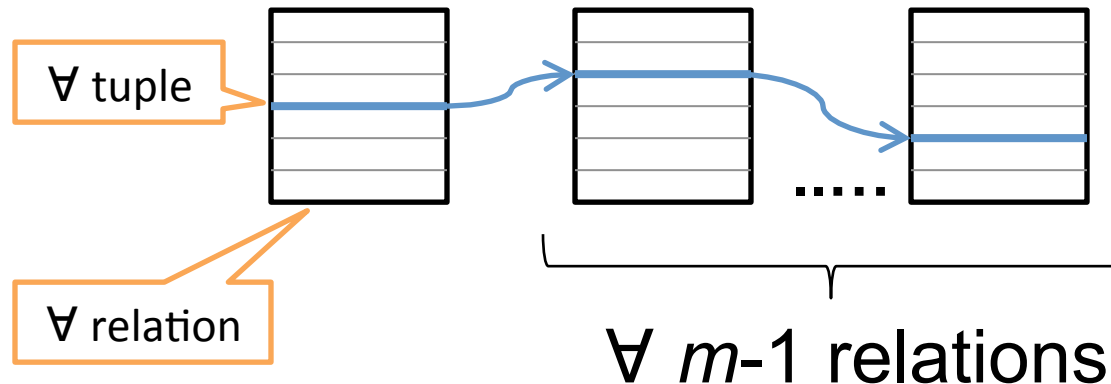
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Outline

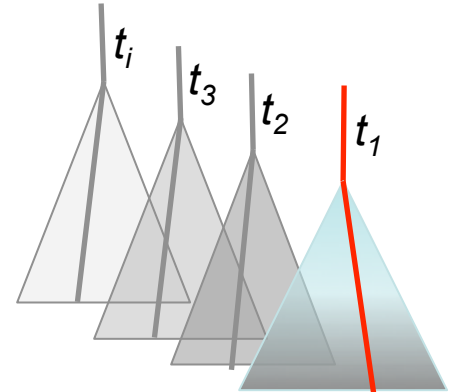
- Introduction
 - $R(*,m)C$ property and its algorithm PERTUPLE
- Partitioning a relation
 - Coarse, fine, intermediate blocks
- Improve PERTUPLE using partitions
 - PERTUPLE \rightarrow PERFB
- Experimental results
- Conclusion

$R(*,m)C$ (a.k.a. m -wise consistency)

- A CSP is $R(*,m)C$ iff
 - Every tuple in a relation can be extended
 - to the variables in the scope of any $(m-1)$ other relations
 - in an assignment satisfying all m relations simultaneously
- $R(*,m)C \equiv$ Every set of m relations is minimal



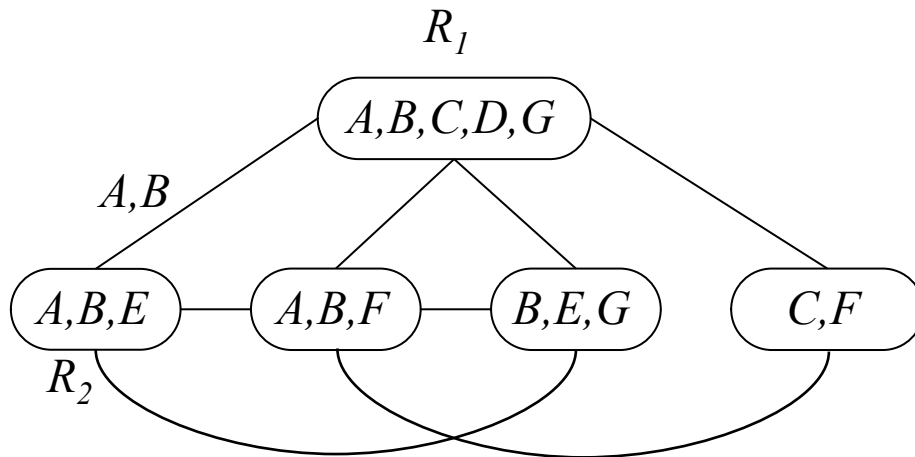
- PERTUPLE enforces $R(*, m)C$
- Store all connected combinations of m relations
- For each relation in a combination
 - For each tuple in the relation
 - **SEARCHSUPPORT**: Conduct backtrack search with FC over the dual CSP induced by the m relations
 - Remove the tuple if no solution is found
- Update propagation queue



Piecewise Functional Constraints

- PW-AC enforces $R(*,2)C$

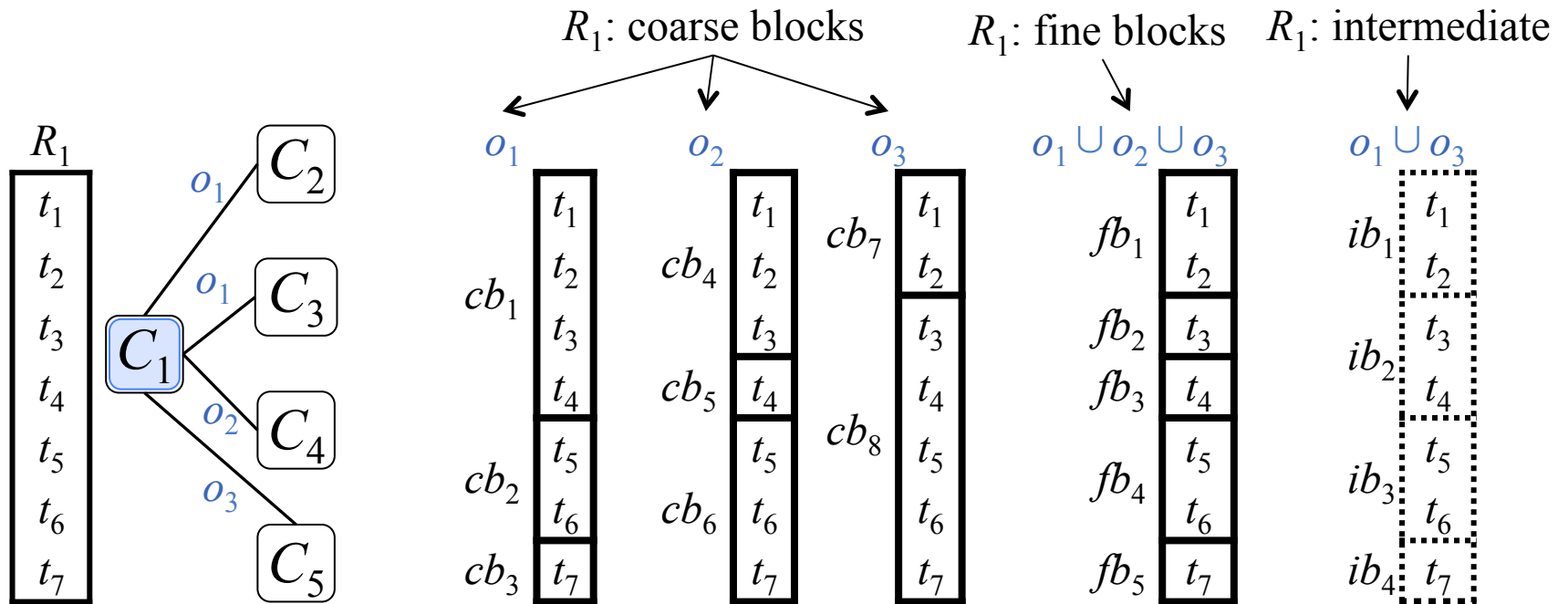
[Samaras & Stergiou JAIR 05]



	R_1					R_2		
	A	B	C	D	G	A	B	E
t_1	0	0	0	0	0	0	0	0
t_2	0	0	0	1	0	0	0	1
t_3	0	0	1	0	0	0	1	0
t_4	0	0	1	1	1	0	1	1
t_5	0	1	1	0	1	1	0	0
t_6	0	1	1	1	1	1	0	1
t_7	1	1	1	1	1	1	0	1

Arrows indicate that the columns for A and B in R_1 are mapped to the columns for A and B in R_2 . Dashed boxes and 'x' marks indicate that the remaining columns in R_1 and R_2 are not mapped.

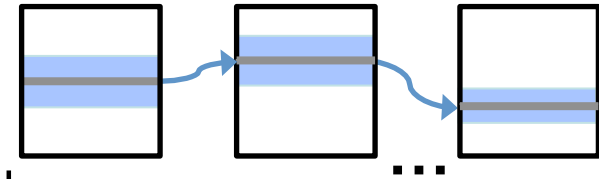
Types of Partitions



- For each relation, we store
 - A single partition of fine blocks
 - As many partitions of coarse blocks as shared subscopes

PERTUPLE → PERFB

- Tuple → Fine Block (FB)
- General mechanism is identical
 - Combinations, queuing, and propagation
- **SEARCHSUPPORT**
 - Search enumerates fine blocks, not tuples
 - Forward checking operates on coarse blocks
- Calls to SEARCHSUPPORT reduced
 - Skips tuples in the same fine blocks
 - Skips fine blocks in the same intermediate block
 - Using two local data structures
- Details in paper



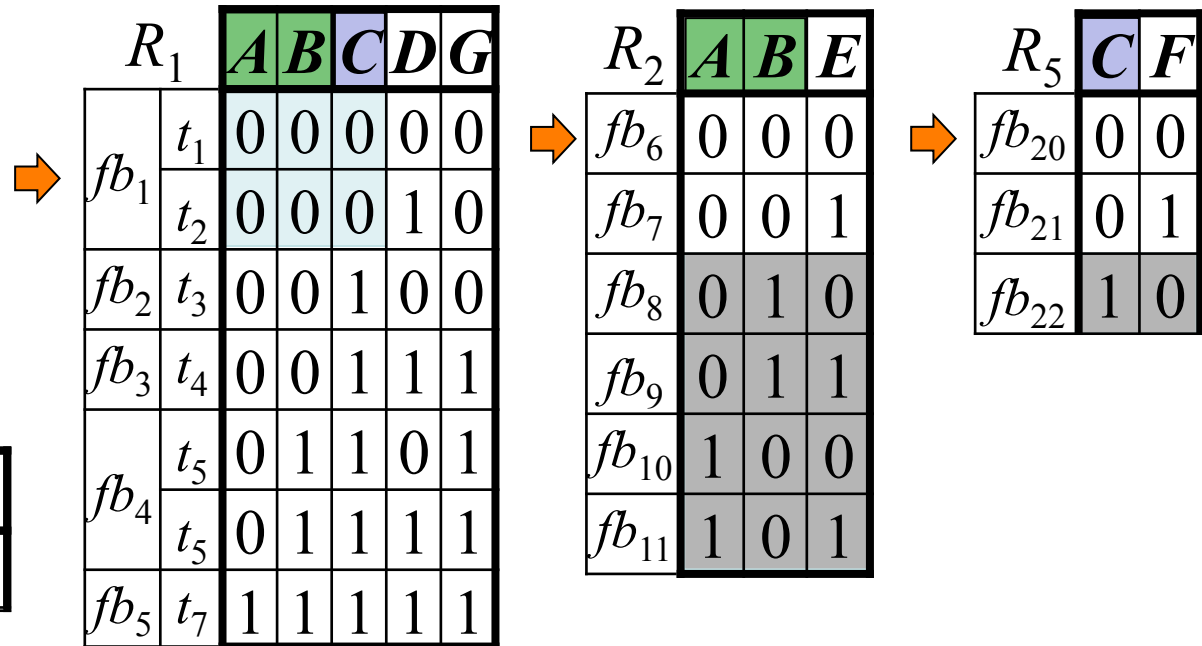
Enforcing $R(*,m)C$ with **PERFB**

Vars in subsopes

<i>Rel</i>	<i>Variables</i>

Equiv FBs for R_1

<i>A</i>	<i>B</i>	<i>C</i>	<i>Consistent</i>
0	0	0	True



Enforcing $R(*,m)C$ with **PERFB**

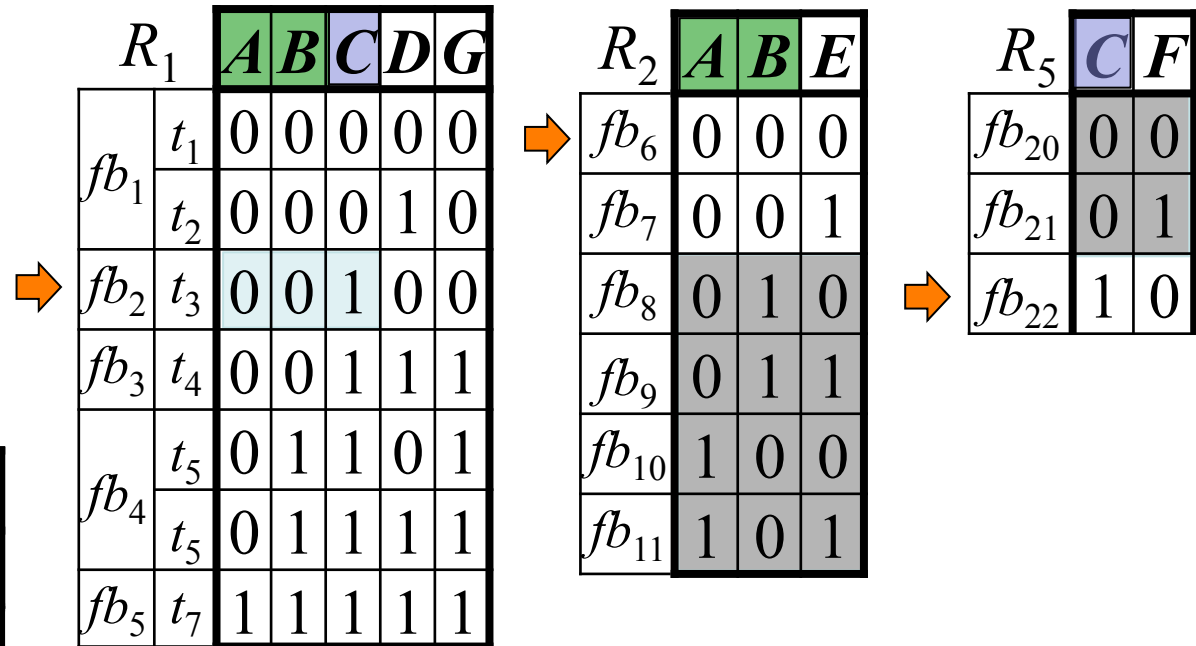
Vars in subsopes

<i>Rel</i>	<i>Variables</i>
R_1	ABC

?

Equiv FBs for R_1

A	B	C	<i>Consistent</i>
0	0	0	True
0	0	1	True



Enforcing $R(*,m)C$ with **PERFB**

Vars in subsopes

<i>Rel</i>	<i>Variables</i>
R_1	ABC

Yes!

Equiv FBs for R_1

<i>A</i>	<i>B</i>	<i>C</i>	<i>Consistent</i>
0	0	0	True
0	0	1	True



R_1		A	B	C	D	G
fb_1	t_1	0	0	0	0	0
	t_2	0	0	0	1	0
fb_2	t_3	0	0	1	0	0
fb_3	t_4	0	0	1	1	1
fb_4	t_5	0	1	1	0	1
	t_5	0	1	1	1	1
fb_5	t_7	1	1	1	1	1

R_2		A	B	E
fb_6		0	0	0
fb_7		0	0	1
fb_8		0	1	0
fb_9		0	1	1
fb_{10}		1	0	0
fb_{11}		1	0	1

R_5		C	F
fb_{20}		0	0
fb_{21}		0	1
fb_{22}		1	0

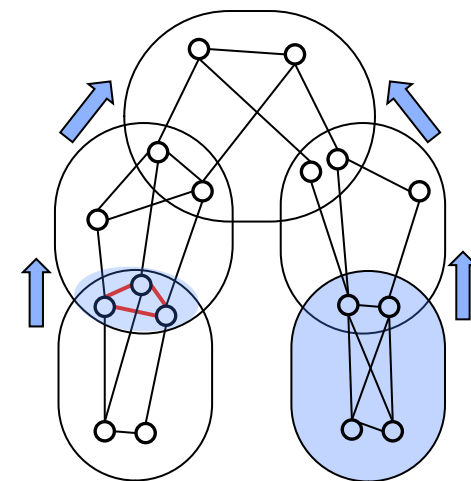
Experimental Setup

- Backtrack search, find first solution, dom/deg

- Maintaining $R(*, m)C$

[Karakashian+ AAI 13]

- Use a tree-decomposition of the CSP
- Enforce consistency on individual clusters
- Add projection of constraints to separators to bolster propagation between adjacent clusters
- Use a minimal dual graph to reduce number of combinations of m constraints
- Better performance than GAC, maxRPWC

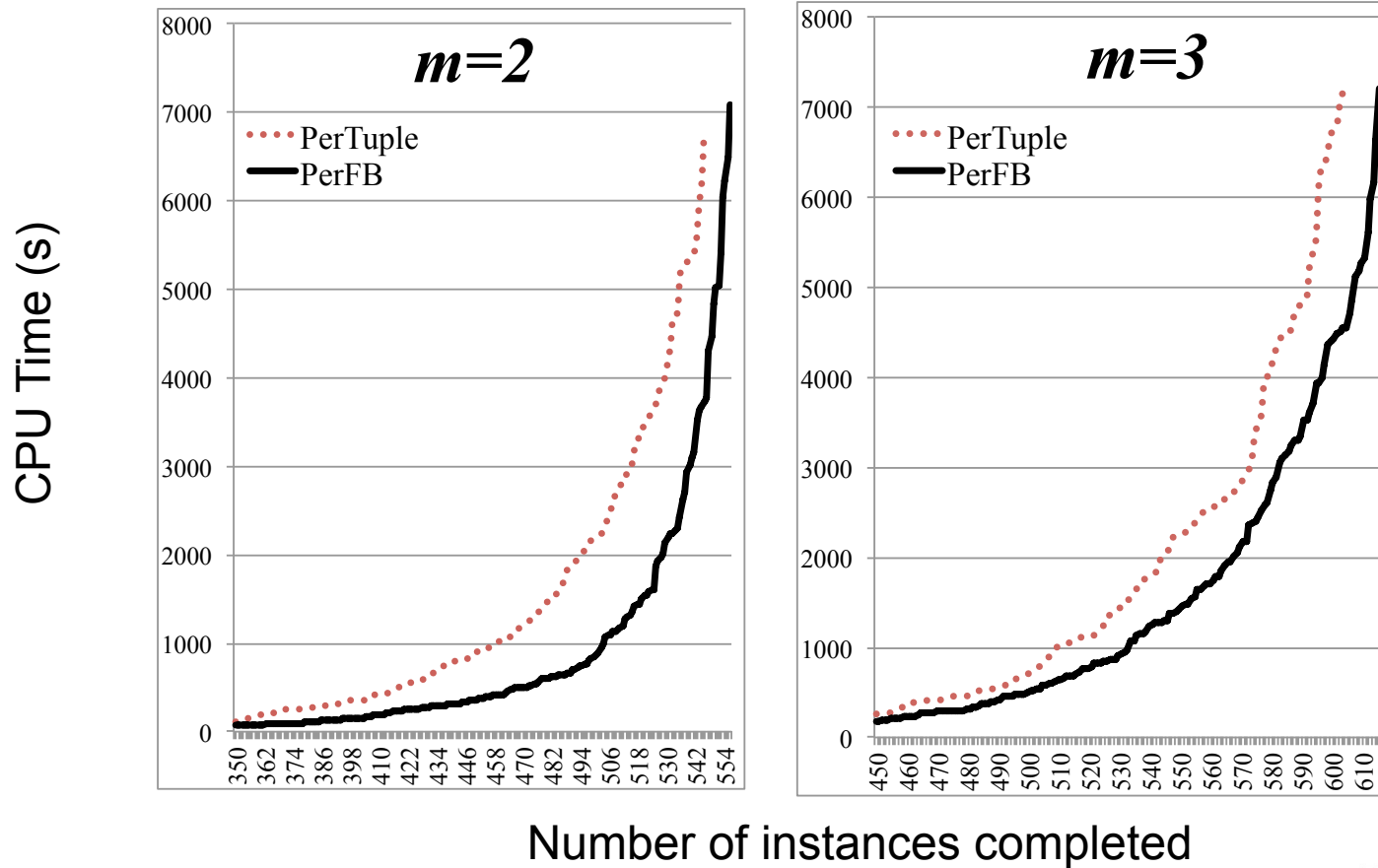


- $m = 2, 3, 4, |\psi(cl)|$

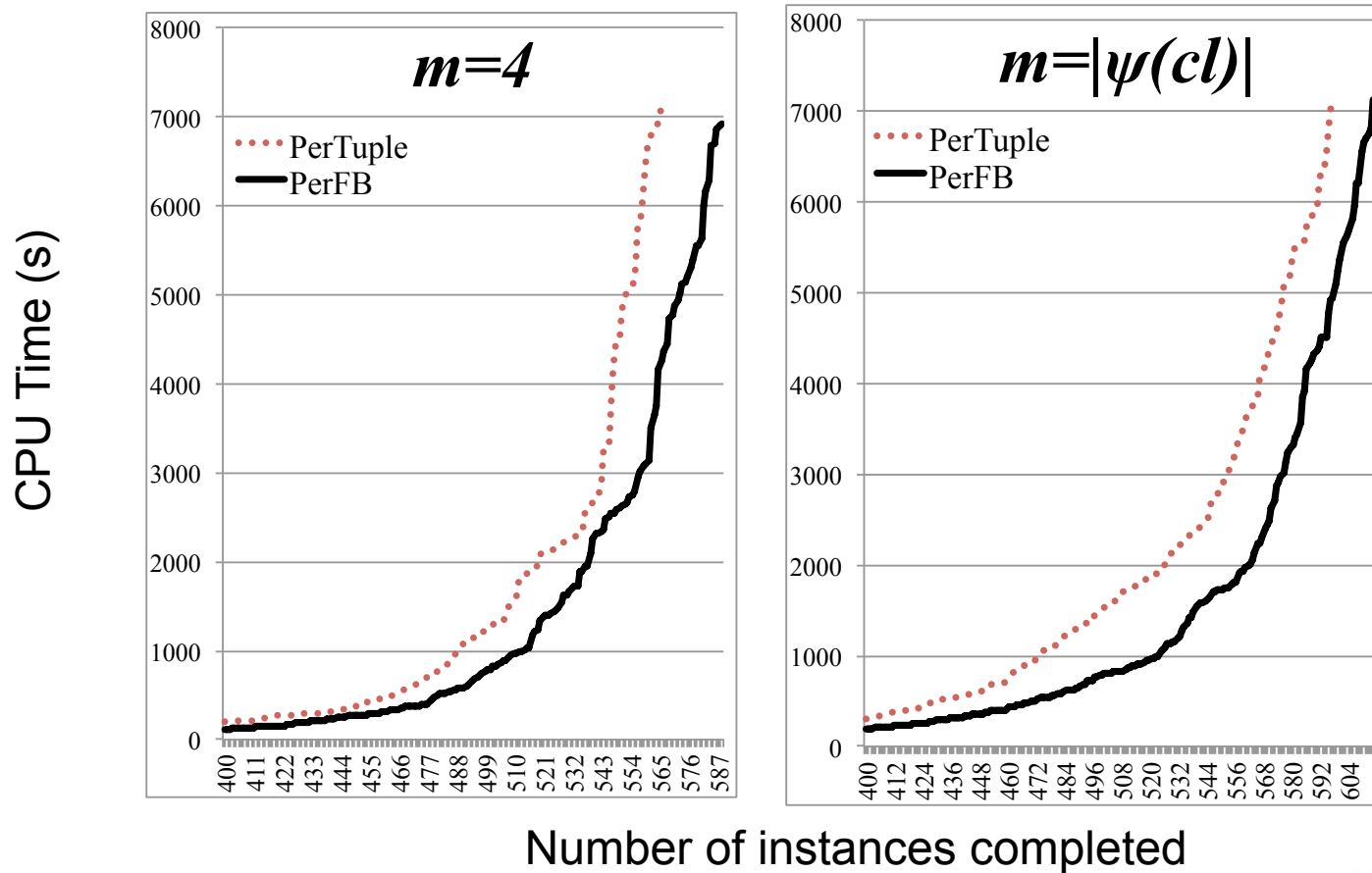
- XCSP benchmark of CP Solver Competition [Lecoutre+]

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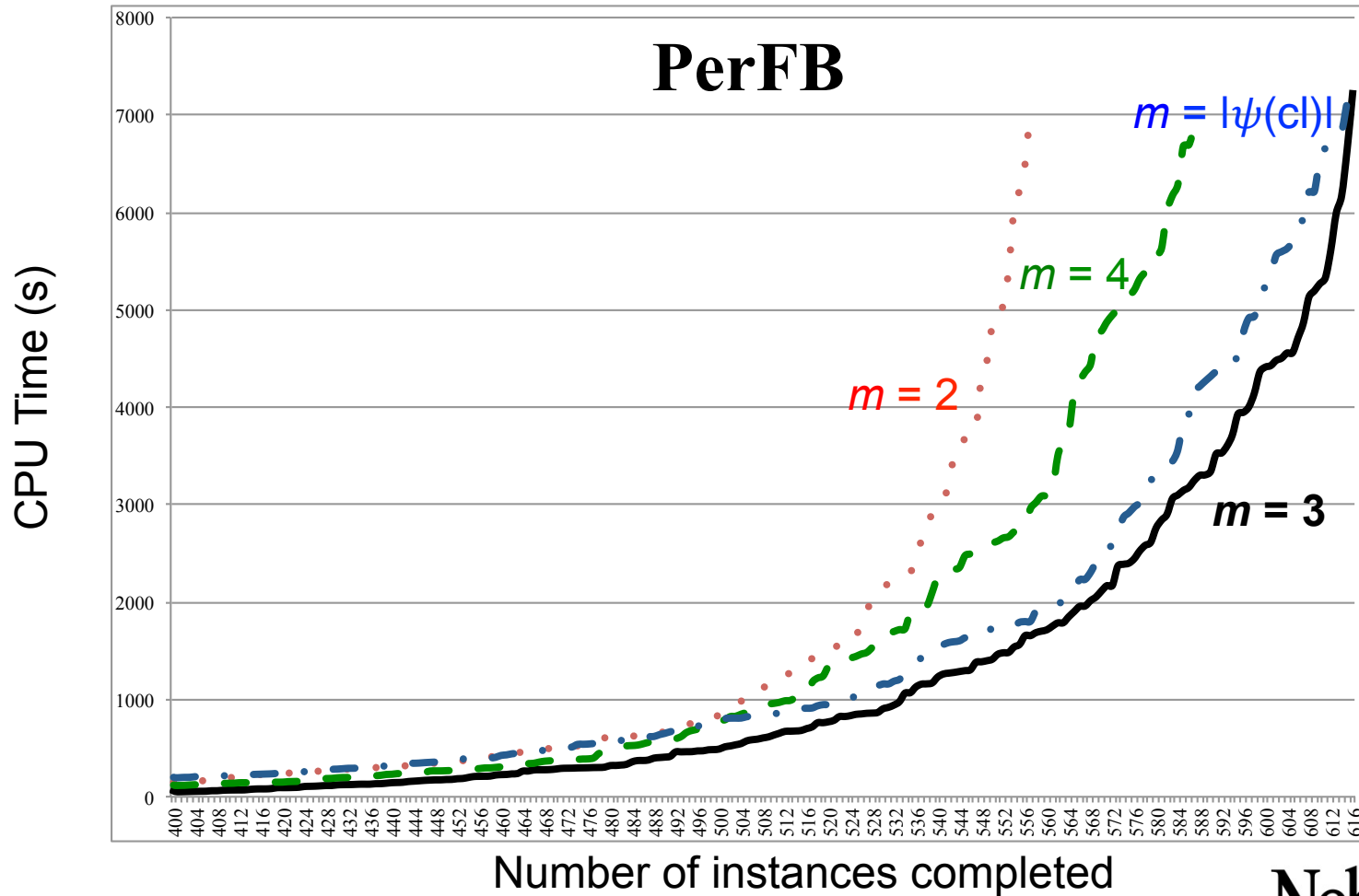
PERFB vs. PERTUPLE: completed instances



PERFB vs. PERTUPLE: completed instances



PERFB $m = 2,3,4, |\psi(c)|$



Detailed Results

Number of instances tested: 853

	$m = 2$		$m = 3$		$m = 4$		$m = \psi(c) $	
	PERTUPLE	PERFB	PERTUPLE	PERFB	PERTUPLE	PERFB	PERTUPLE	PERFB
#Completed	546	557	604	616	566	589	597	615
... only by	5	16	1	13	2	25	8	26
... by both	541		603		564		589	
Avg. CPU (sec)	538	227	521	362	472	314	669	458
SEARCHSUPPORT calls (10^9)	86.4	0.0	88.1	26.1	52.7	19.6	24.7	8.1
Call ratio	--		3.37		2.69		3.06	

Conclusions

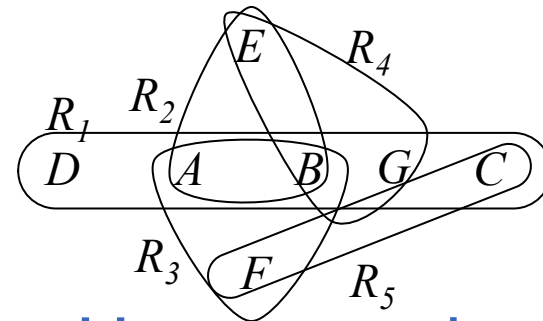
- Contributions
 - Designed PERFB
 - to replace PERTUPLE of [Karakashian+ AAAI 10]
 - by extending the work of [Samaras & Stergiou JAIR 05]
 - Empirically showed benefits of our approach
- Future work
 - Extend our approach to our other algorithm for enforcing $R(*, m)C$ [Karakashian PhD 13]

Thank you for listening

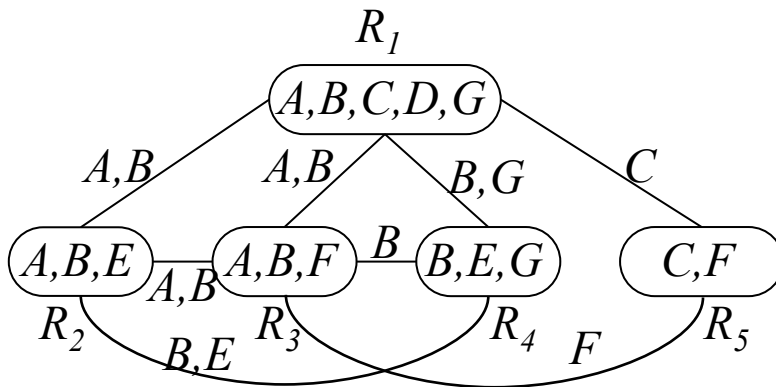
Wake up the Chair!

CSP: Graphical Representations

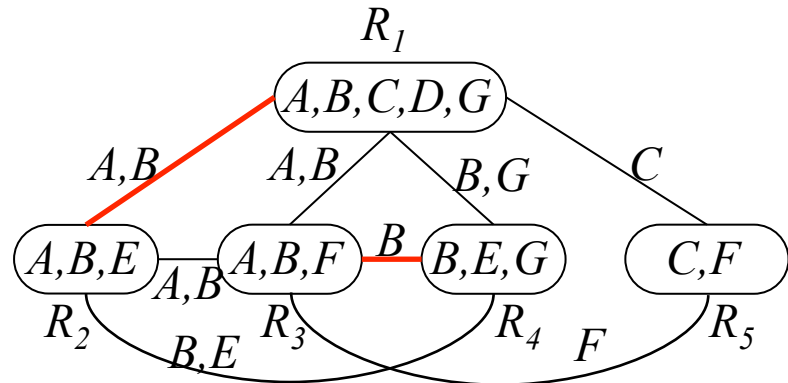
- Hyper graph
- Dual graph
- Minimal dual graph



Hyper graph



Dual graph



Minimal dual graph

[Janssen+, 1989]

Block Statistics

Benchmark	Absolute	Averages		
	Max	Min	Max	Mean
geom	17	1.0	1.2	1.0
graphColoring-hos	3	1.0	2.0	1.0
graphColoring-sgb-book	12	1.0	7.7	1.1
hanoi	2	1.0	2.0	1.0
modifiedRenault	260	1.0	25.6	1.0
rand-10-20-10	2	1.0	1.3	1.0
renault	4	1.0	4.0	1.0
ssa	8	1.0	3.1	1.1
tightness0.9	38	1.0	28.1	1.0
varDimacs	16	1.0	3.4	1.1

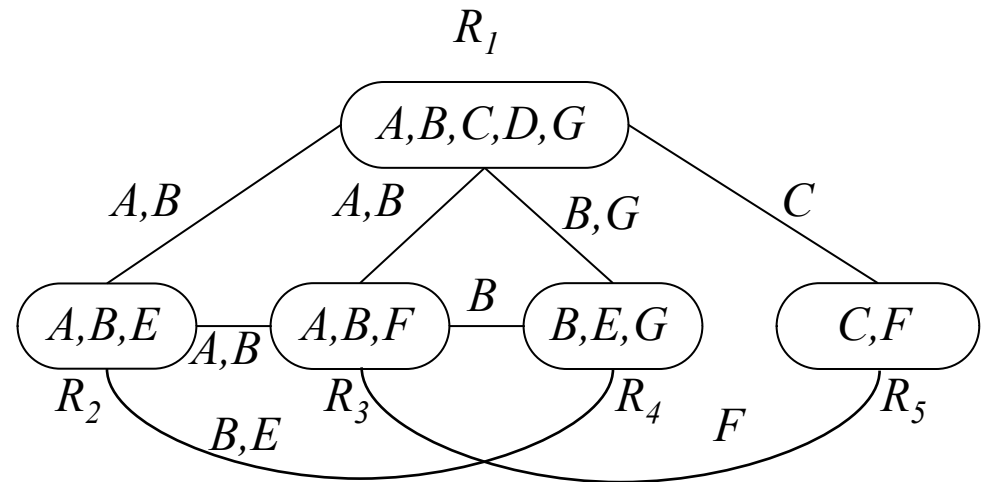
Block Statistics

Benchmark	Abs	Averages		
	Max	Min	Max	Mean
aim-50,100,200,pseudo	4	1.0	4.0	2.1
cmpsed-25-1-2,25,40,80	10	1.0	10.0	8.0-8.4
cmpsed-25-10-20	10	1.0	10.0	7.6
cmpsed-75-1-2,25,40,80	10	1.0	10.0	8.3-8.5
dag-rand	108	1.0	91.6	2.9
dubois,pret	2	1.0	2.0	1.5
geom	20	6.4	20.0	15.0
grCol-hos	6	1.0	3.3	3.3
grCol-mug	3	1.0	2.5	2.4
grCol-register-mulsol	48	23.2	23.2	23.2
grCol-sgb-book	12	1.0	7.7	7.5
grCol-sgb-games	8	1.0	6.3	6.1

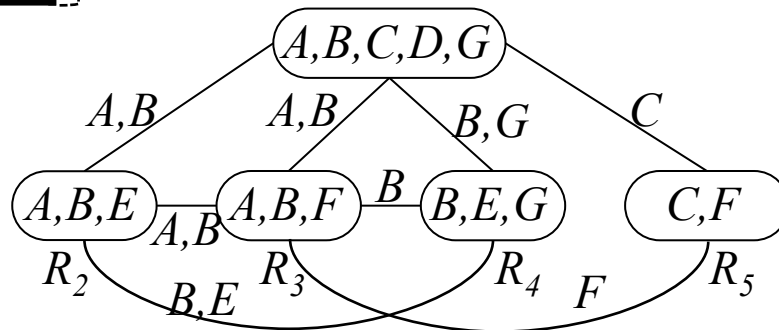
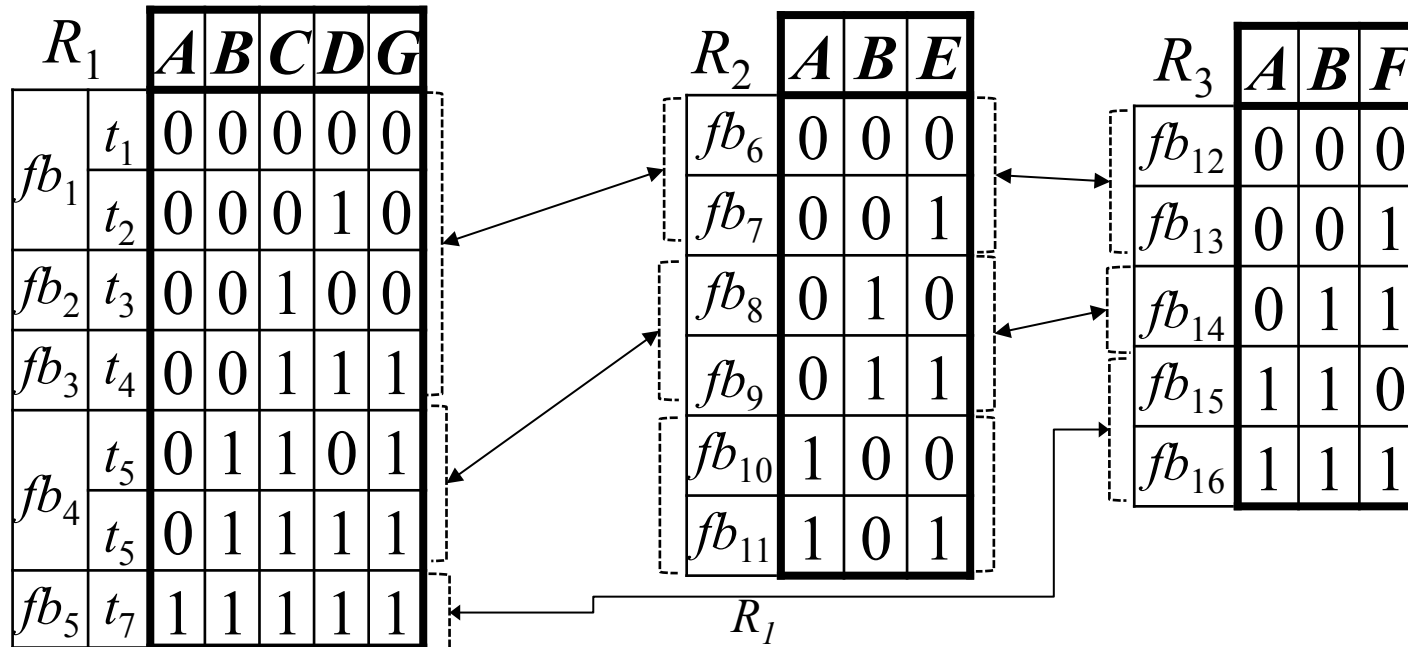
Benchmark	Abs	Averages		
	Max	Min	Max	Mean
grCol-sgb-queen	17	10.3	10.3	10.3
hanoi	3	1.0	3.0	2.9
lexVg	875	1.0	484.7	3.6
modifiedRenault	48,720	1.0	48,720.0	7.9
rand-10-20-10	1,046	1.0	119.2	1.3
rand-3-20-20-fcd	190	1.0	181.5	12.8
renault	48,720	1.0	48,720.0	7.7
rlfapGr/ScensMod	44,43	1.0	30.0,35.6	18.5,19.4
ssa	31	1.0	14.7	2.1
super-queens	49	15.6	17.6	16.4
tightness0.9	40	1.0	36.3	16.9
varDimacs	512	1.0	115.0	5.6

Partitioning Relations -- Definitions

- Scope
- Subscope
- Combination

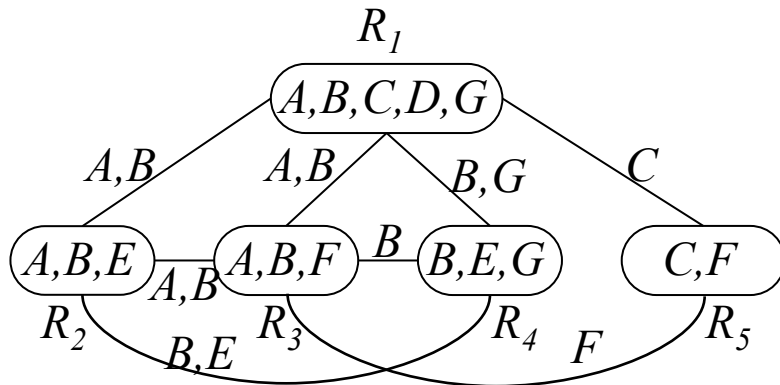


Coarse Partitions



Fine Partitions

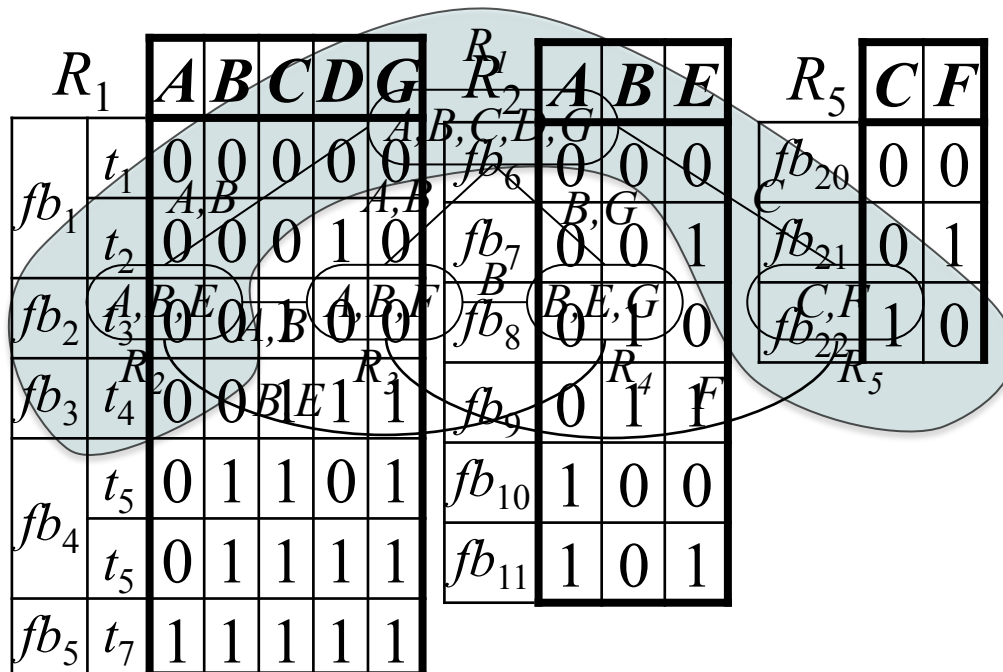
- Equivalence class that induces smallest set of tuples



R_1		A	B	C	D	G	R_2			A	B	E
fb_1	t_1	0	0	0	0	0	fb_6	0	0	0		
	t_2	0	0	0	1	0		fb_7	0	0	1	
fb_2	t_3	0	0	1	0	0	fb_8		0	1	0	
	t_4	0	0	1	1	1		fb_9	0	1	1	
fb_4	t_5	0	1	1	0	1	fb_{10}		1	0	0	
	t_5	0	1	1	1	1		fb_{11}	1	0	1	
fb_5	t_7	1	1	1	1	1						

Intermediate Partitions

- Consider a combination of size $m=3$...



Intermediate Partitions

- First, identify the subsopes that affect R_1

R_1		A	B	C	D	G	R_2			R_5	C	F	
fb_1	t_1	0	0	0	0	0	fb_6	0	0	0	fb_{20}	0	0
	t_2	0	0	0	1	0	fb_7	0	0	1	fb_{21}	0	1
fb_2	t_3	0	0	1	0	0	fb_8	0	1	0	fb_{22}	1	0
fb_3	t_4	0	0	1	1	1	fb_9	0	1	1			
fb_4	t_5	0	1	1	0	1	fb_{10}	1	0	0			
	t_5	0	1	1	1	1	fb_{11}	1	0	1			
fb_5	t_7	1	1	1	1	1							

Intermediate Partitions

- Project the union of those subscopes over the relation

R_1			A	B	C	D	G	R_2			A	B	E	R_5		C	F
ib_1	fb_1	t_1	0	0	0	0	0	fb_6	0	0	0	fb_{20}	0	0			
		t_2	0	0	0	1	0		fb_7	0	0		1	fb_{21}	0	1	
ib_2	fb_2	t_3	0	0	1	0	0	fb_8	0	1	0	fb_{22}	1	0			
	fb_3	t_4	0	0	1	1	1	fb_9	0	1	1						
ib_3	fb_4	t_5	0	1	1	0	1	fb_{10}	1	0	0						
		t_5	0	1	1	1	1	fb_{11}	1	0	1						
ib_4	fb_5	t_7	1	1	1	1	1										