in Privacy Preserving Planning

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Partial Disclosure of Private Dependencies in Privacy Preserving Planning: Privacy Preserving Planning

Plans can be found while revealing only a small portion of the private dependencies

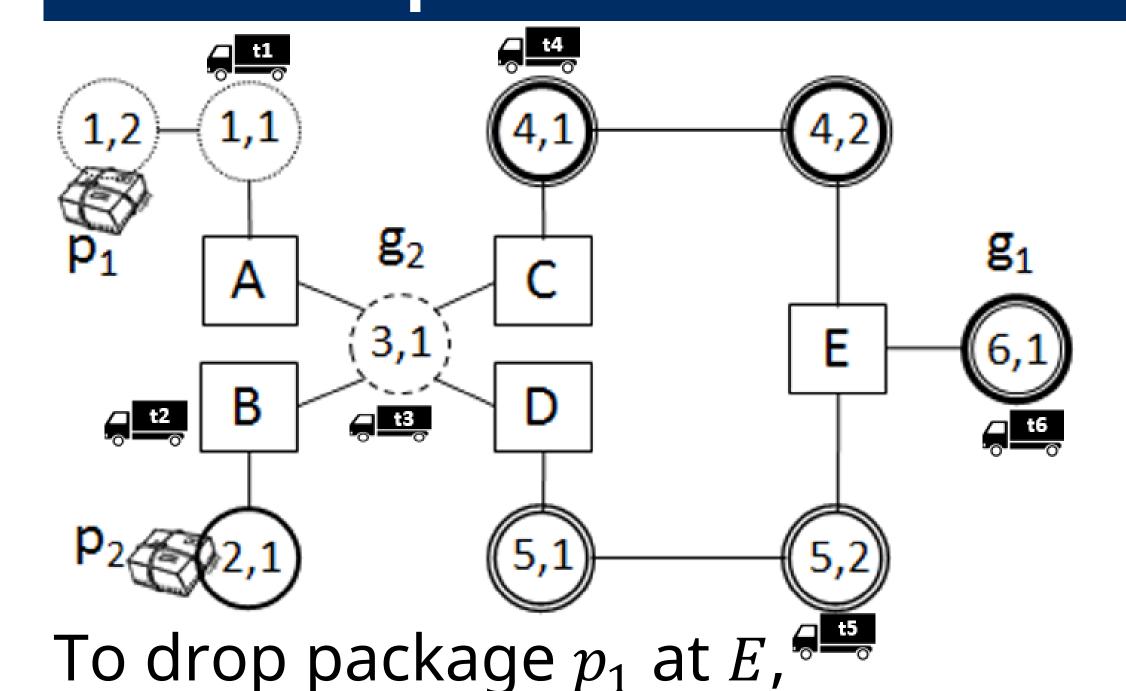
1. The Problem

Collaborative Privacy Preserving Planning (CPPP) algorithms reveal private dependencies. Private dependency: which public actions facilitate the execution of another public action. Can we find plans while revealing only a small portion of the dependencies?

2. Methods

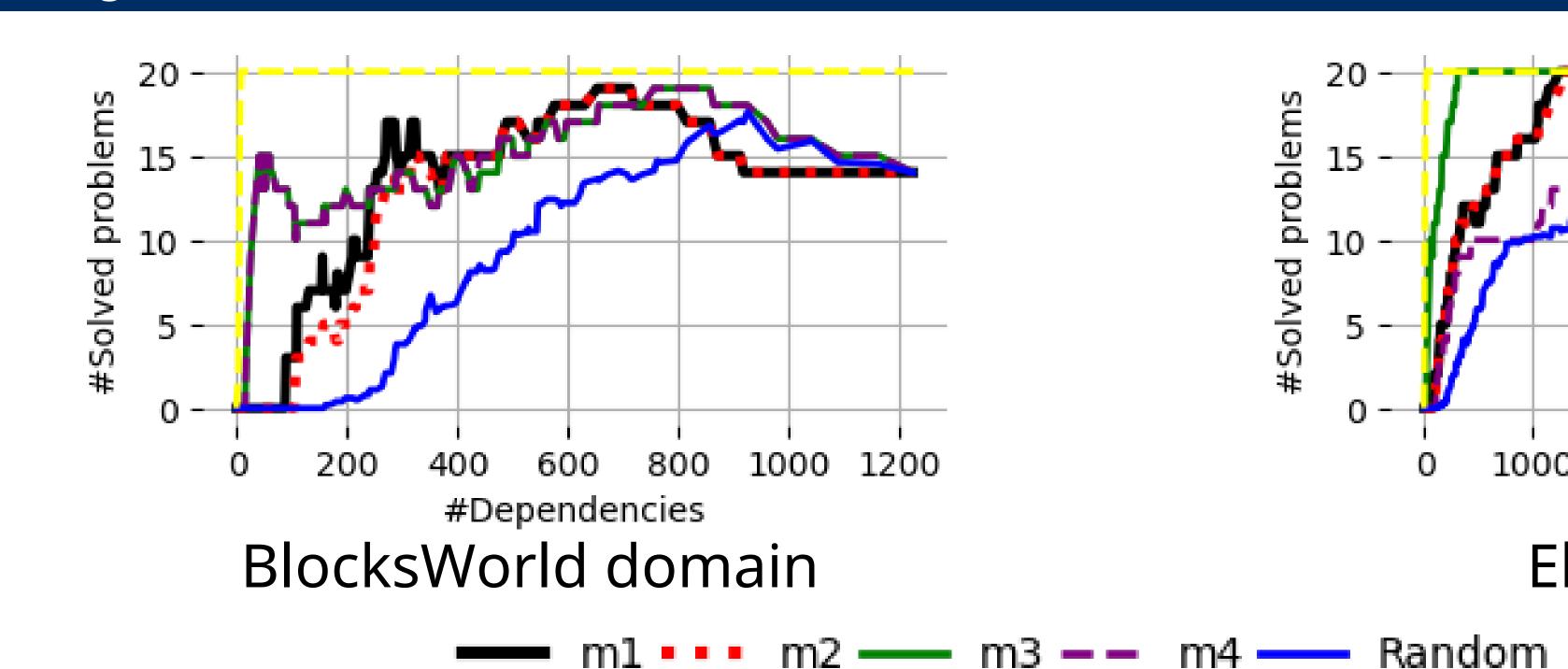
- 1. Creating the dependencies graph.
- 2. Computing 4 heuristic dependency revealing rankings over the graph.
- 3. Applying the heuristics to MAFS and the Joint Projection planners.

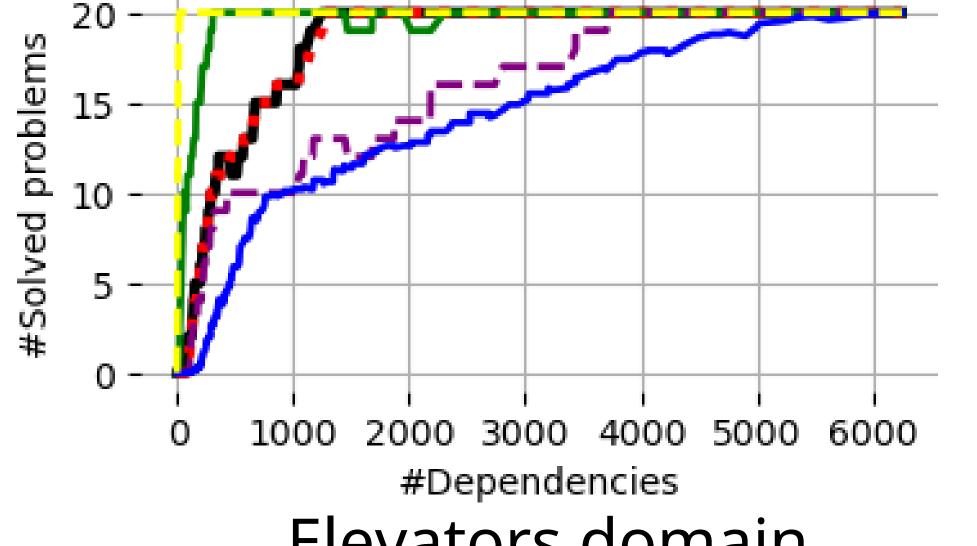
3. Example



truck t_4 must pick p_1 from (e.g.) C

4. Key Results

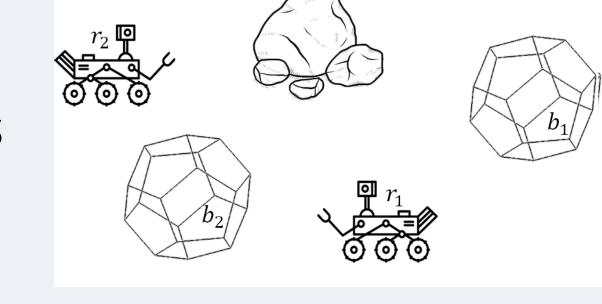


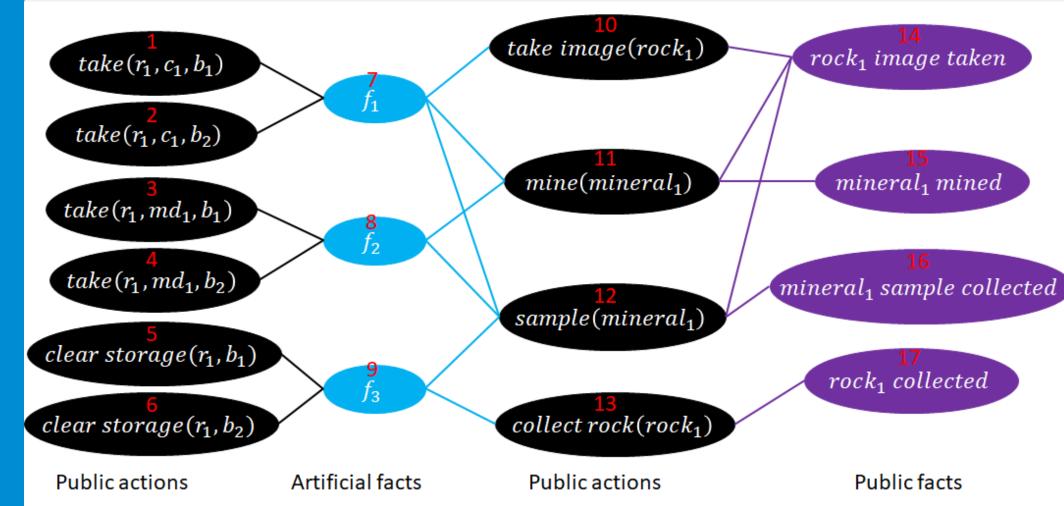


Elevators domain

Hindsight

Technical details

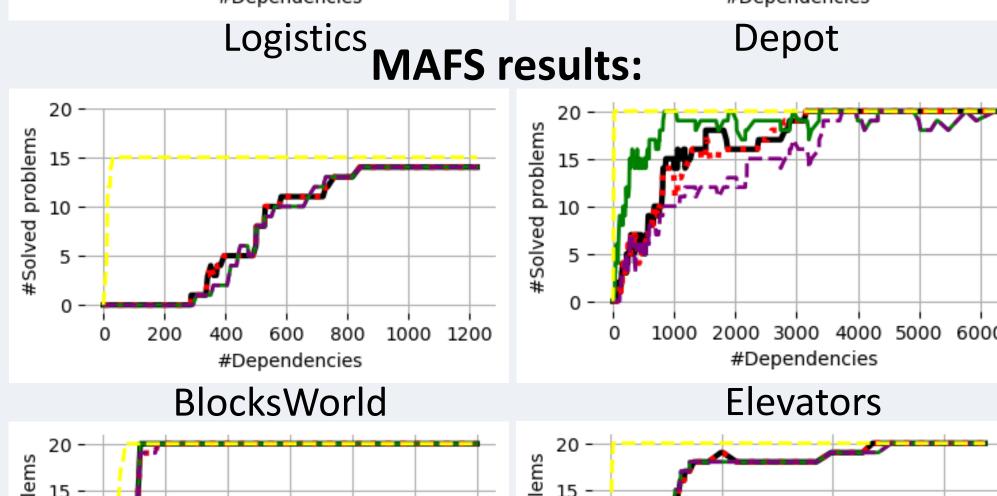


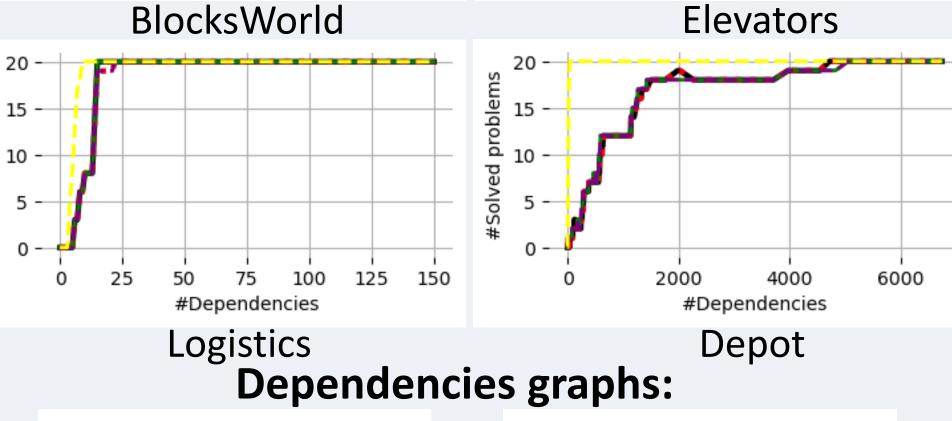


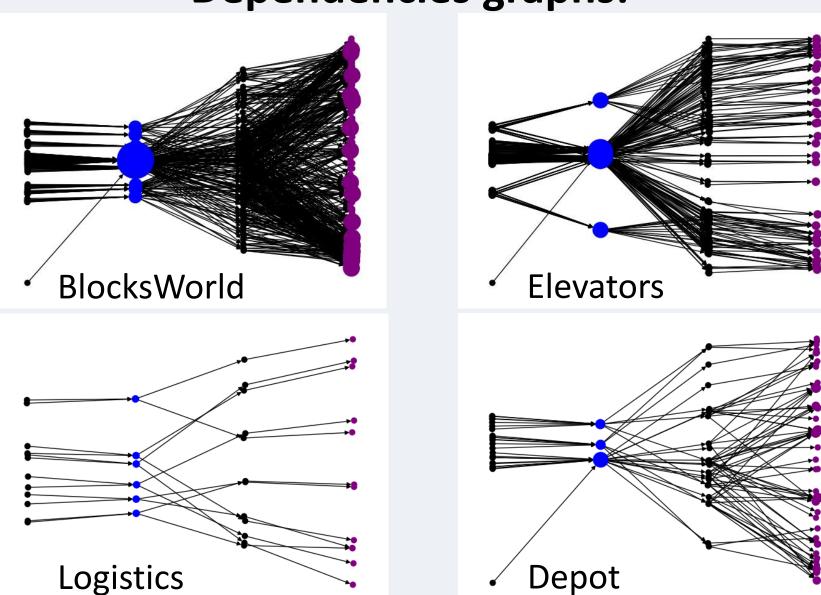
Heuristics:

- M1 Precondition to the largest amount of public actions
- M2 Precondition to the largest amount of public facts.
- M3 Achieves the largest amount of public actions.
- M4 Achieves the largest amount of public facts.

Joint Projection results:







Domain	Joint Projection			MAFS		
	Min	Min. dep	lmp.	Min	Min. dep	Imp.
locksWorld	48.78	84.10	41.83%	28.95	29.82	2.45%
epot	21.40	28.55	20.20%	21.20	21.26	0.82%
riverlog	28.19	51.08	28.80%	21.26	22.31	5.96%
evators	33.70	38.33	12.58%	36.94	39.89	8.03%
ogistics	25.02	27.02	5.97%	25.15	27.60	7.41%
overs	50.75	58.71	12.57%	58.50	70.32	16.09%
enoTravel	46.52	49.16	9.13%	48.52	51.17	9.02%
verage	36.34	48.13	18.73%	34.36	37.48	7.11%

Imp = the improvement from the first solution cost to the best solution cost