

Memory Efficient Max Flow for Multi-label Submodular MRFs

Introduction

Problem: Minimize a multi-label MRF with pairwise interactions

$$E(\mathbf{x}) = \sum_{i \in \mathcal{V}} \theta_i(x_i) + \sum_{(i,j) \in \mathcal{E}} \theta_{ij}(x_i, x_j),$$

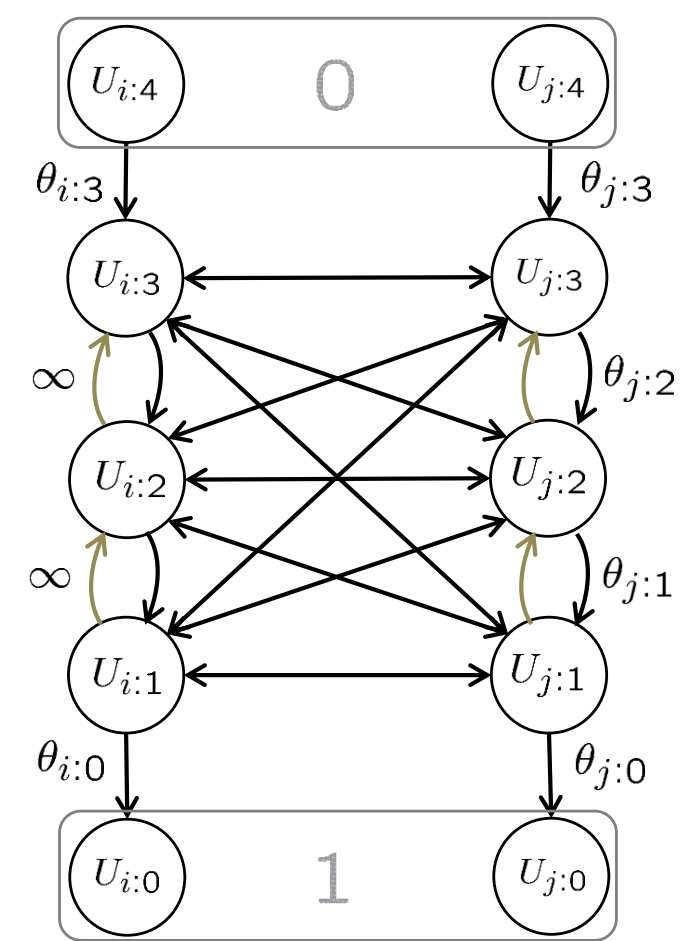
where $x_i \in \{0, 1, \dots, \ell - 1\}$.

Multi-label submodular:

$$\theta_{ij}(\lambda', \mu) + \theta_{ij}(\lambda, \mu') - \theta_{ij}(\lambda, \mu) - \theta_{ij}(\lambda', \mu') \geq 0,$$

for all $\lambda, \lambda', \mu, \mu'$ where $\lambda < \lambda'$ and $\mu < \mu'$ [4].

Current method: Ishikawa algorithm [3].



The Ishikawa graph

Memory complexity: $O(|\mathcal{E}|\ell^2)$

E.g. $|\mathcal{V}| = 10^6$, $\ell = 256$
 $|\mathcal{E}| \approx 2 \times 10^6$ (4-connected)
 Ishikawa edges $\approx 2 \times 10^6 \times 2 \times 256^2$
 Memory \approx **1000 GB**

Contribution: An algorithm with memory complexity $O(|\mathcal{E}|\ell)$.

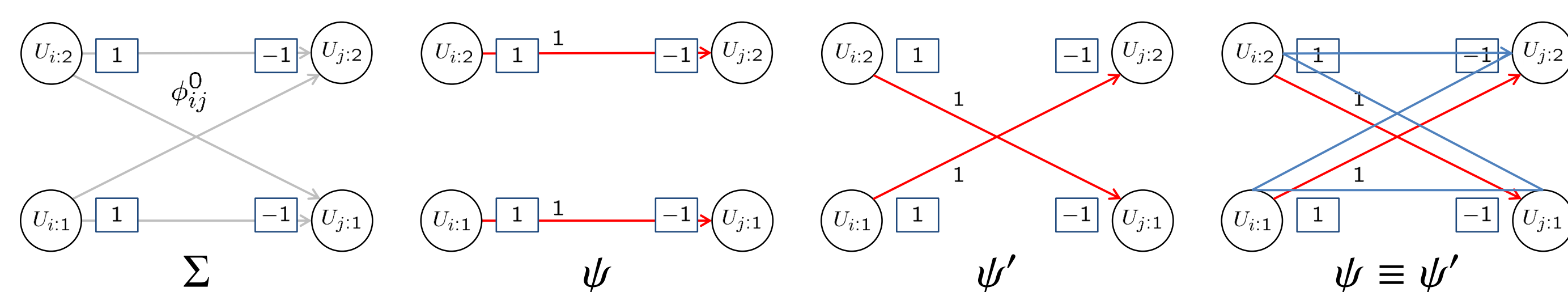
Memory Efficient Flow Encoding

Idea: Don't store the residual graph but **exit-flows** between each pair of neighbouring columns.

Exit-flow: Given flow ψ , an exit-flow is defined as

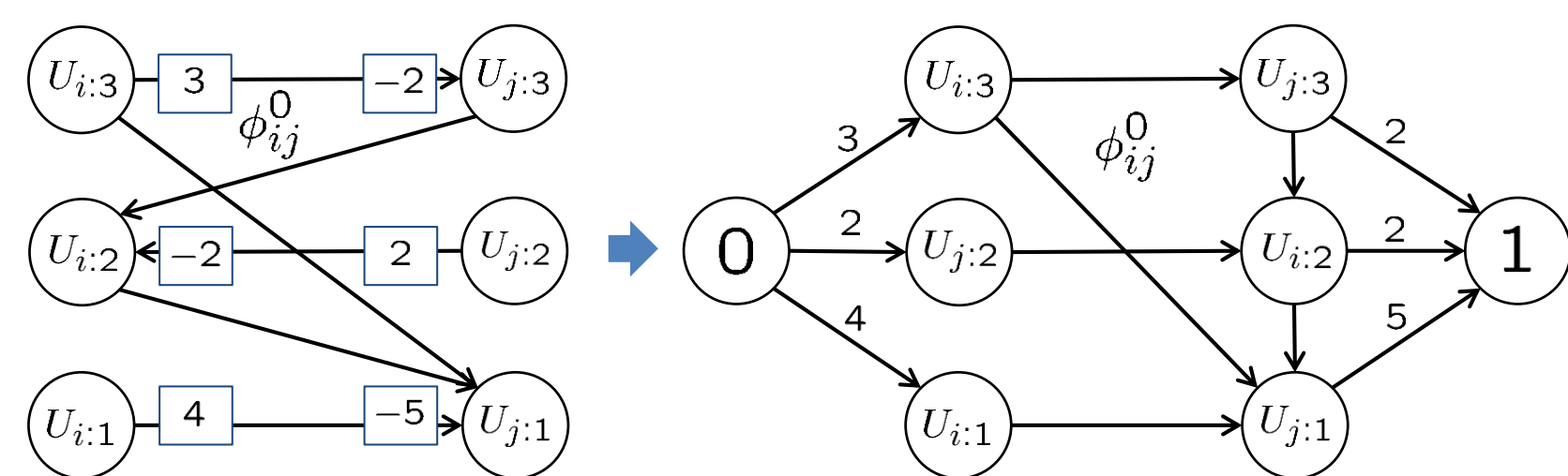
$$\Sigma_{ij:\lambda} = \sum_{\mu} \psi_{ij:\lambda\mu}.$$

The residual graph can be rapidly computed from the exit-flows.



Rapidly computing the residual graph:

Idea: Formulate a small max-flow problem.



Memory Efficient Max Flow

Algorithm

Require: ϕ^0 \triangleright Initial Ishikawa capacities
 $\Sigma \leftarrow 0$ \triangleright Initialize exit-flows
repeat
 $P \leftarrow \text{augmenting_path}(\phi^0, \Sigma)$
 $\Sigma \leftarrow \text{augment}(P, \phi^0, \Sigma)$
until no augmenting paths possible

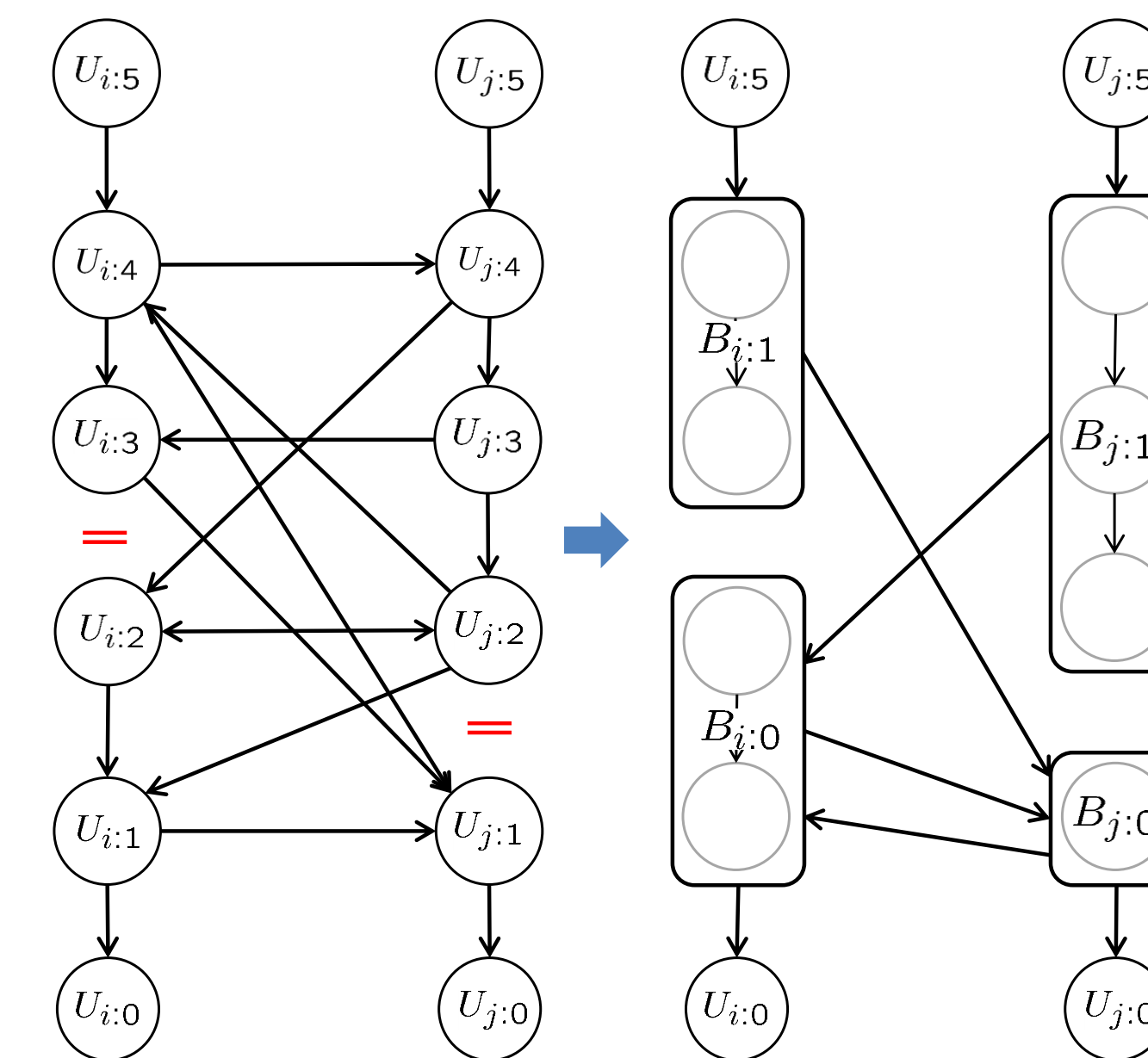
Assumption:
 ϕ^0 can be stored in an efficient manner.

Efficiently Finding an Augmenting Path

Idea: Search for augmenting paths in a **simplified graph**.

Simplified graph:

- Unweighted sparse graph.
- Fewer augmenting paths.



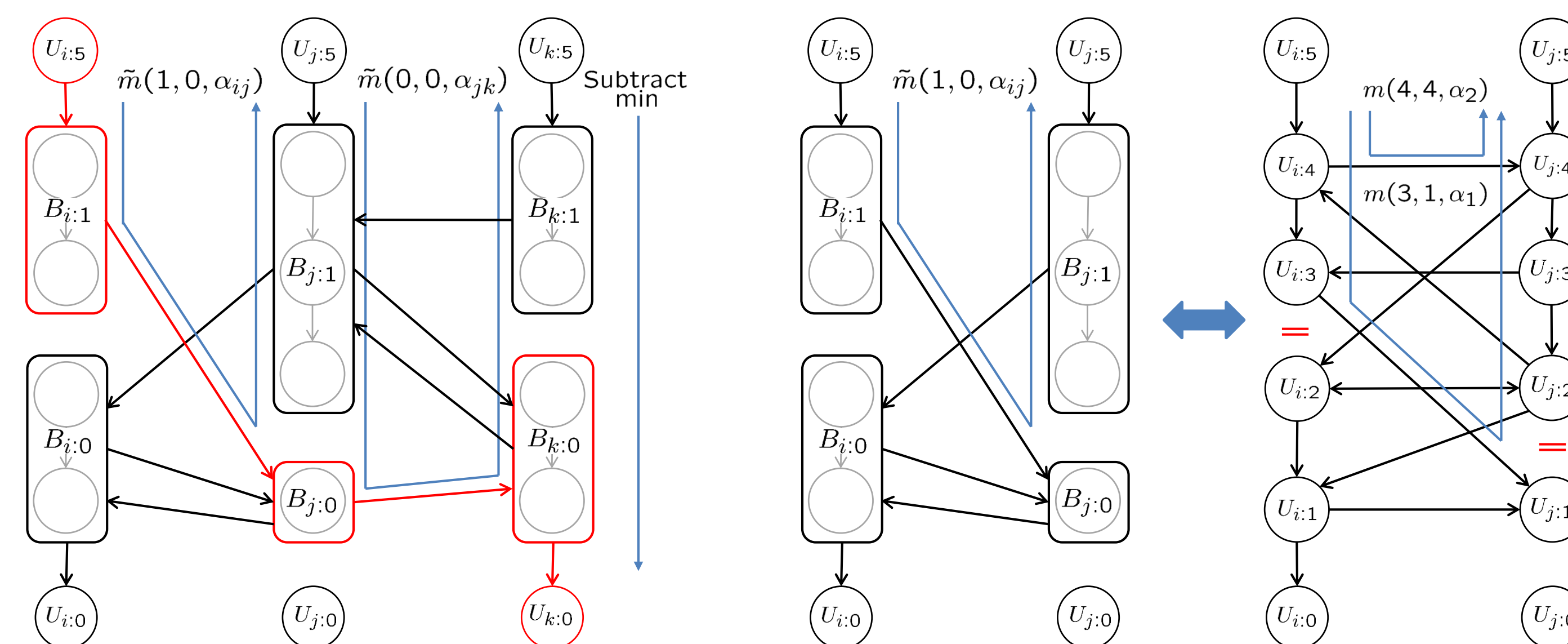
Search-tree-recycling:

- Good empirical performance.

Augmentation

Idea: Pass flow around loops.

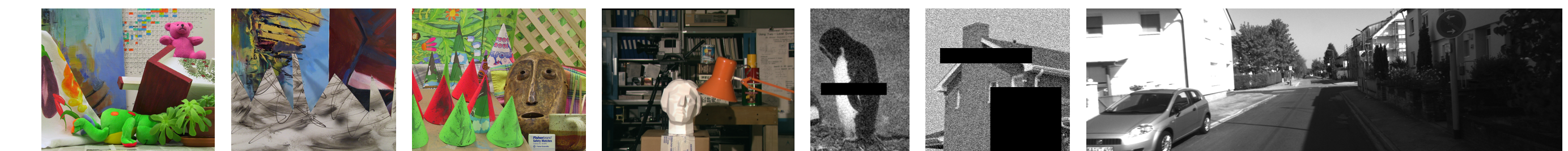
- Push the maximum permissible flow through each flow-loop.
- Applying flow-loops translates to updating the exit-flows.



Experiments

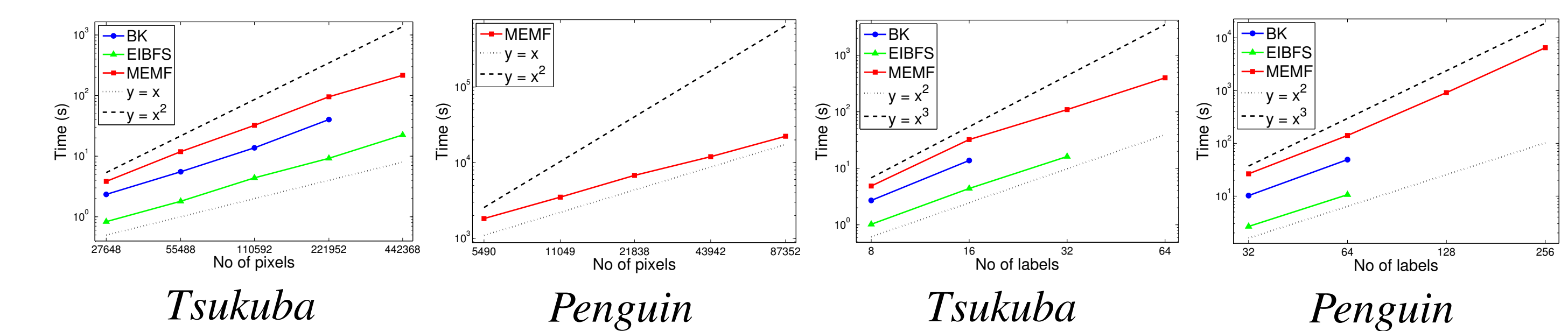
Dataset:

- Middlebury stereo and inpainting instances.
- KITTI stereo instance.



Problem		Memory [MB]			Time [s]		
Name	Labels	BK [1]	EIBFS [2]	MEMF	BK	EIBFS	MEMF
Tsukuba	16	3195	2495	211	14	4	30
Venus	20	7626	5907	396	35	9	60
Sawtooth	20	7566	5860	393	31	8	35
Map	30	6454	4946	219	57	9	36
Cones	60	*72303	*55063	1200	-	-	371
Teddy	60	*72303	*55063	1200	-	-	2118
KITTI	40	*88413	*67316	2215	-	-	19008
Penguin	256	*173893	*130728	663	-	-	6835
House	256	*521853	*392315	1986	-	-	9290

Comparison with other max-flow implementations



Empirical time complexity: $O(|\mathcal{V}|\ell^3)$

Code: <https://github.com/tajanathan/memf>



References

- Y Boykov and V Kolmogorov. An experimental comparison of min-cut/max-flow algorithms for energy minimization in vision. *PAMI*, 2004.
- A V Goldberg, S Hed, H Kaplan, P Kohli, R E Tarjan, and R F Werneck. Faster and more dynamic maximum flow by incremental breadth-first search. In *Algorithms-ESA*. 2015.
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