

Optimization for Machine Learning

Exercise sheet 3

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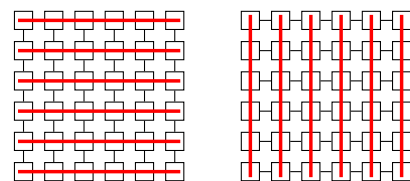
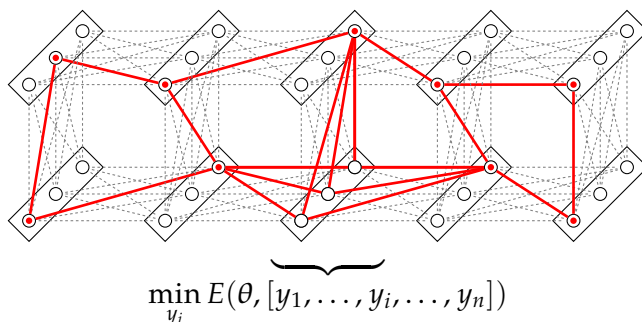
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General Notes The goal of this exercise is to implement the individual inference techniques that have been discussed in the lecture. In addition to the implementation you should answer the following questions for *each* of the following exercises:

- (i) What type of algorithm is it?¹
- (ii) Does the algorithm provide guarantees?²
- (iii) Give a rough estimate of the time complexity per iteration.
- (iv) Describe the quality of the output.

The implementation should be fully functional on the provided *tsukuba* model. You are allowed to assume that your input is a grid graph, as this makes the implementation more straight-forward. Output the bounds of the objective after each iteration (when applicable). Plot the progress of the dual/primal objective with respect to the number of iterations and elapsed runtime. Include the output and the results in your submission. We provide helper functions for some technical subtasks (parsing of input files, visualization of labeling, fast grid graph access).

Exercise 3.1 Implement the iterated conditional modes algorithm. Extend it to block-ICM where you use a row/column-decomposition.

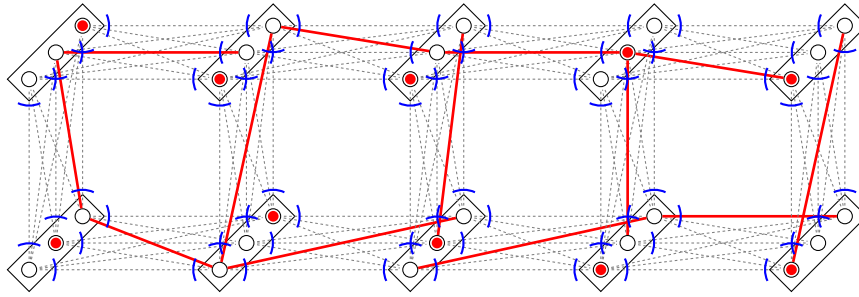


row / column decomposition

¹e.g. primal/dual, subgradient/block-coordinate based, etc.

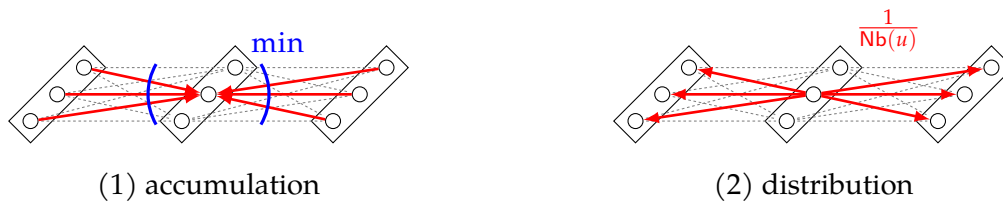
²e.g. convergence, approximation/exact optimization, etc.

Exercise 3.2 Implement the subgradient algorithm for graphical models. Use naïve rounding to obtain the label assignment.



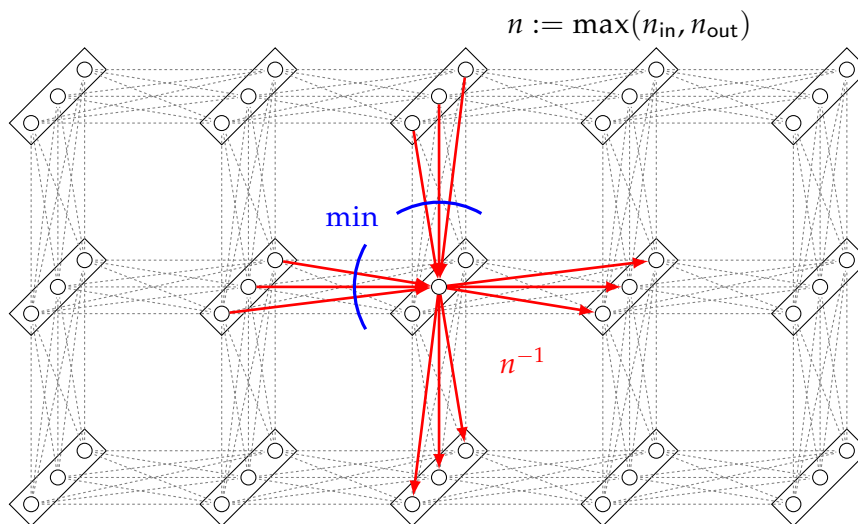
Bonus: Can you improve the output of the algorithm by adapting the step size (fine tune $\beta > 0$ and $0 > \gamma \geq -1$ on $\alpha^t = \beta(1+t)^\gamma$)?

Exercise 3.3 Implement the (isotropic) MinSum-Diffusion inference algorithm.



Use the rounding scheme that was discussed in the lecture.

Exercise 3.4 Implement the TRW-S (sequential tree-reweighted message-passing) inference algorithm.



Use the rounding scheme that was discussed in the lecture.

Note that this algorithm is still one of the best inference schemes for graphical models, even though the general idea of the algorithm is relatively easy.