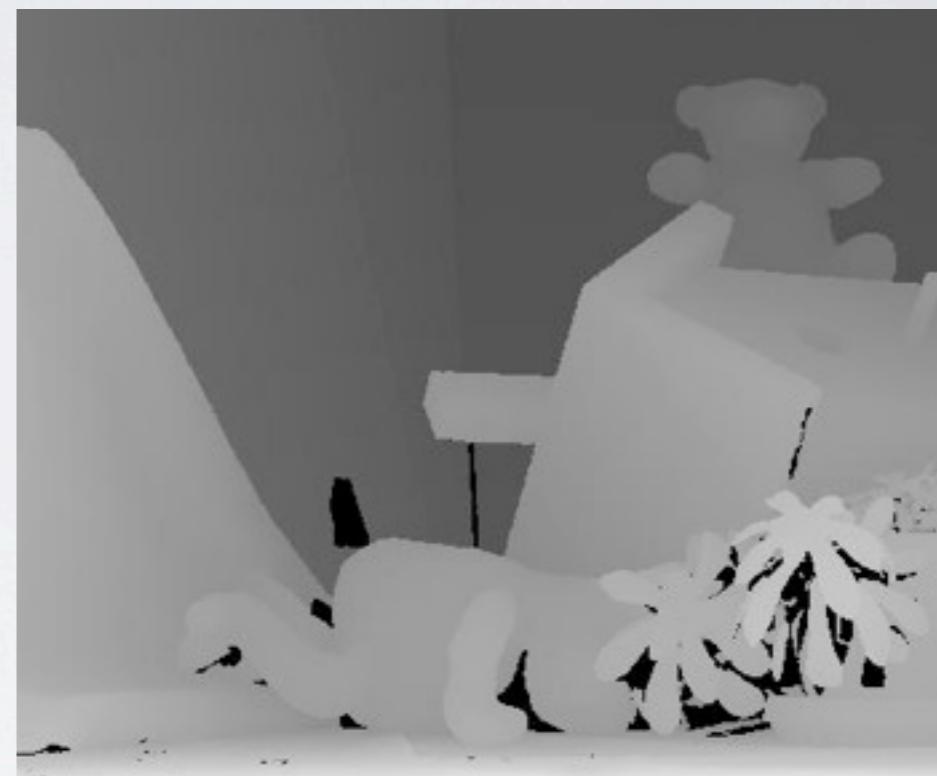


# ASSIGNMENT 3

Stereo Correspondence

Nov. 6, 2013

# STEREO CORRESPONDENCE



# STEPS FOR STEREO CORRESPONDENCE

- Camera calibration
- Dense pixel correspondence (epipolar constraint)
  - Image rectification
  - Disparity estimation
- Depth estimation

# STEPS FOR STEREO CORRESPONDENCE

- ~~Camera calibration~~ Assume calibrated camera
- Dense pixel correspondence (epipolar constraint)
  - Image rectification
  - Disparity estimation
- Depth estimation

# STEPS FOR STEREO CORRESPONDENCE

- ~~Camera calibration~~
- Dense pixel correspondence (epipolar constraint)
  - ~~Image rectification~~ Assume rectified images
  - Disparity estimation
- Depth estimation

# STEPS FOR STEREO CORRESPONDENCE

- ~~Camera calibration~~
- Dense pixel correspondence (epipolar constraint)
  - ~~Image rectification~~
  - Disparity estimation
- ~~Depth estimation~~ Trivial to obtain from disparity

# STEPS FOR STEREO CORRESPONDENCE

- ~~Camera calibration~~
- Dense pixel correspondence (epipolar constraint)
  - ~~Image rectification~~
  - Disparity estimation
- ~~Depth estimation~~

# DISPARITY ESTIMATION

- Estimate the optimal disparity assignment for each pixel by minimizing the following energy function

$$E(y, x, d) = \sum_{x,y}^{\text{Pixels}} \text{data}(y, x, d(y, x)) + \sum_{x,y,nx,ny}^{\text{Pixel neighbors}} \text{smoothness}(d(y, x), d(ny, nx))$$

where:

$\text{data}(y, x, d)$  = cost of assigning disparity  $d$  at pixel  $(y, x)$

$\text{smoothness}(d_1, d_2)$  = cost of assigning disparities  $d_1$  and  $d_2$  at neighboring pixels

# STEPS FOR DISPARITY ESTIMATION

$$E(y, x, d) = \sum_{x,y}^{\text{Pixels}} \text{data}(y, x, d(y, x)) + \sum_{x,y,nx,ny}^{\text{Pixel neighbors}} \text{smoothness}(d(y, x), d(ny, nx))$$

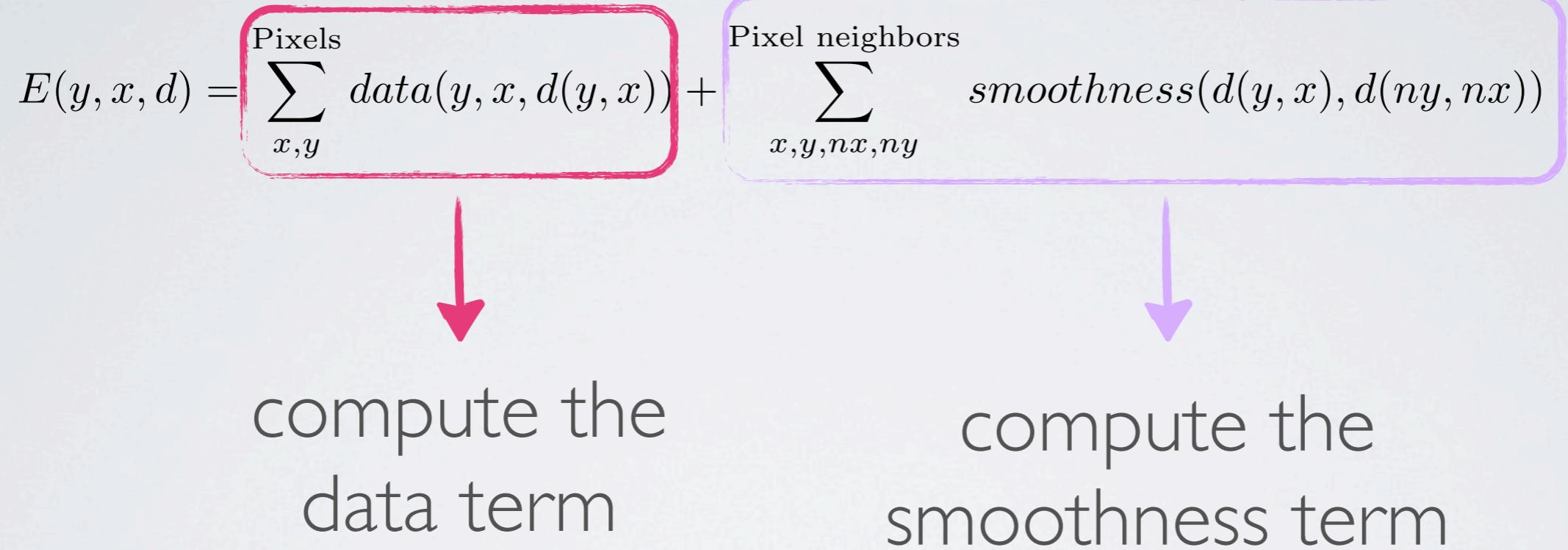
# STEPS FOR DISPARITY ESTIMATION

$$E(y, x, d) = \sum_{x,y}^{\text{Pixels}} \text{data}(y, x, d(y, x)) + \sum_{x,y,nx,ny}^{\text{Pixel neighbors}} \text{smoothness}(d(y, x), d(ny, nx))$$

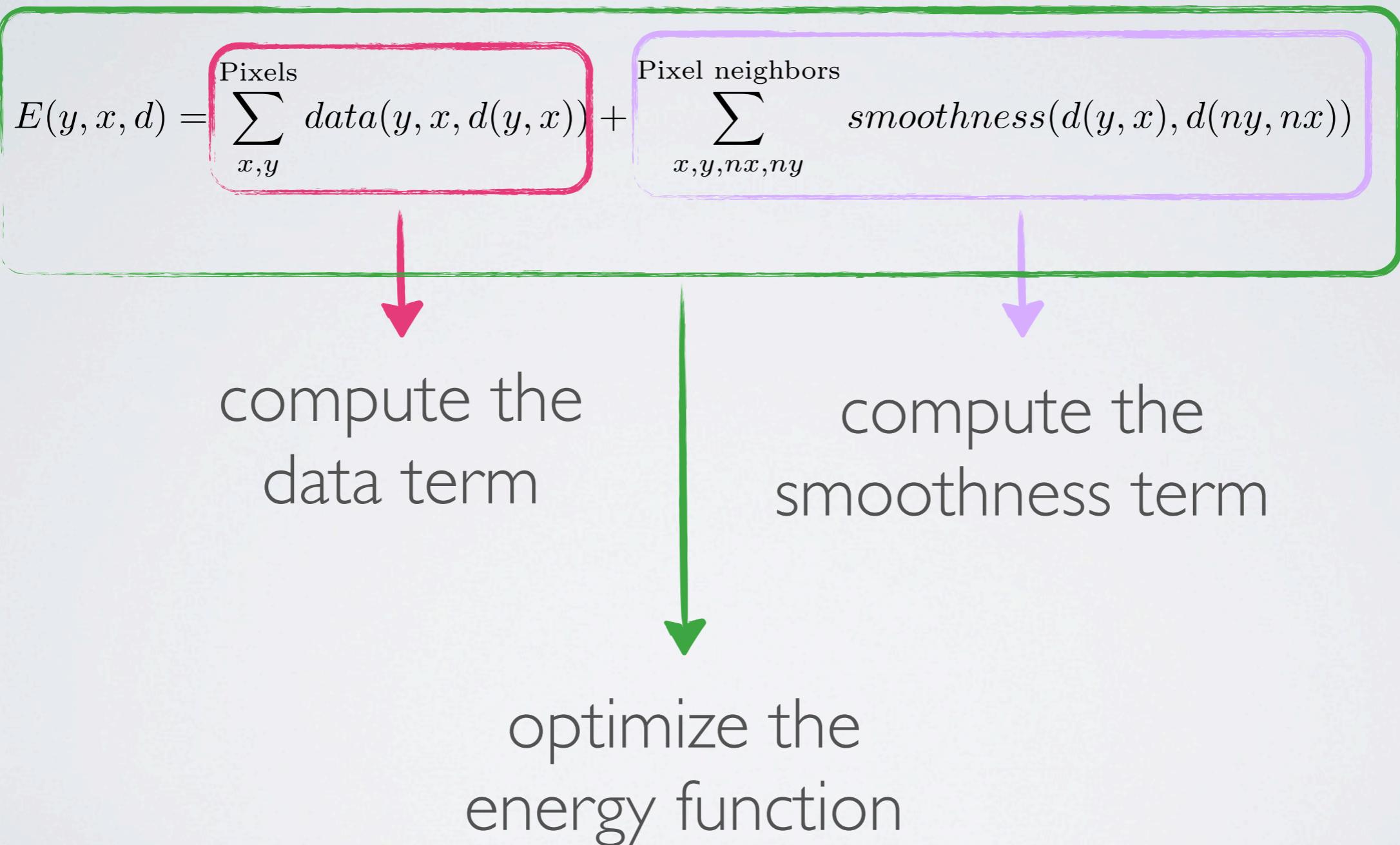


compute the  
data term

# STEPS FOR DISPARITY ESTIMATION



# STEPS FOR DISPARITY ESTIMATION



# ALGORITHMS TO IMPLEMENT

compute data term	compute smoothness term	optimize energy
LI	LI	dynamic programming graph cut
awesome	awesome	awesome

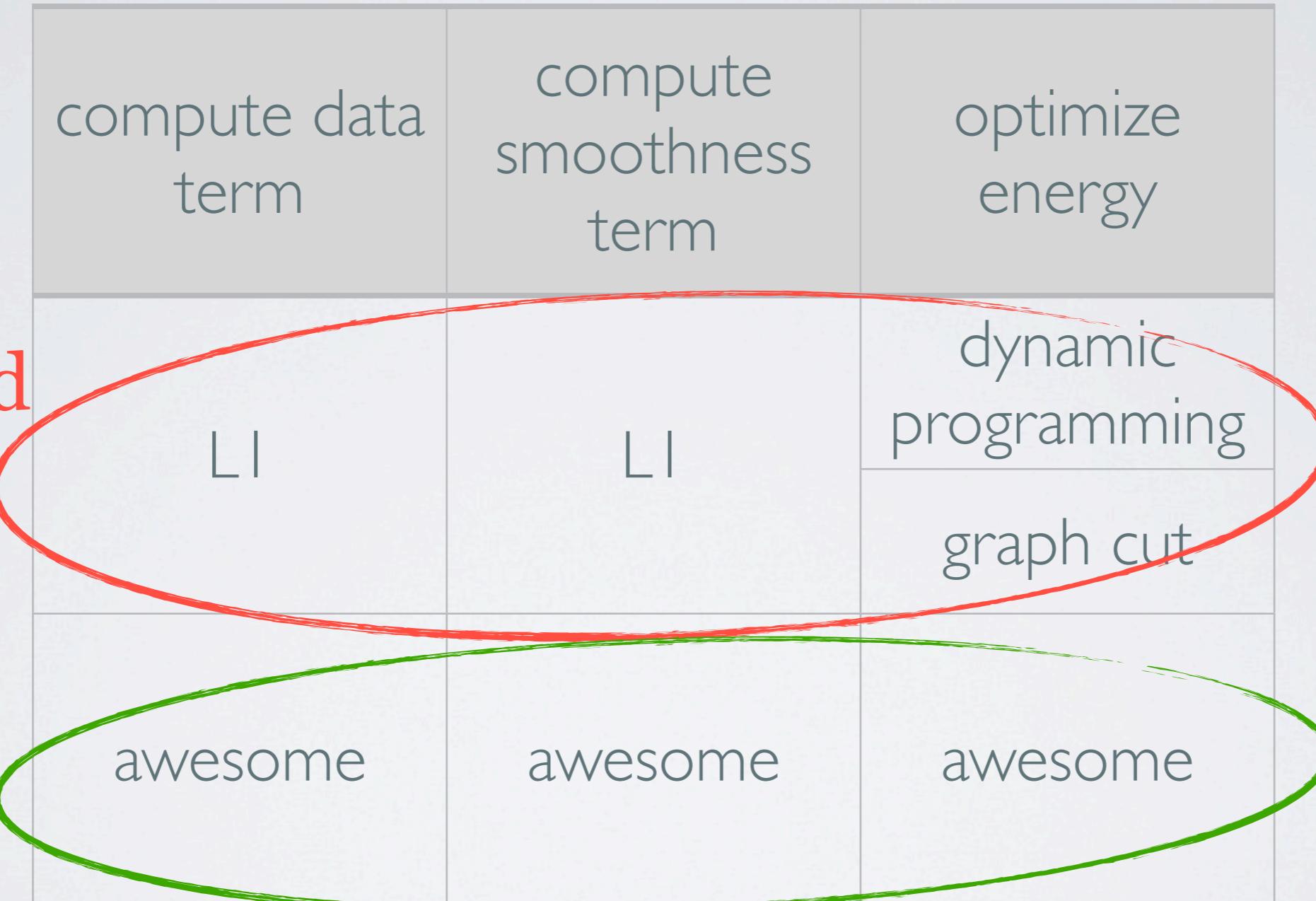
# ALGORITHMS TO IMPLEMENT

Required

compute data term	compute smoothness term	optimize energy
LI	LI	dynamic programming graph cut
awesome	awesome	awesome

# ALGORITHMS TO IMPLEMENT

Required



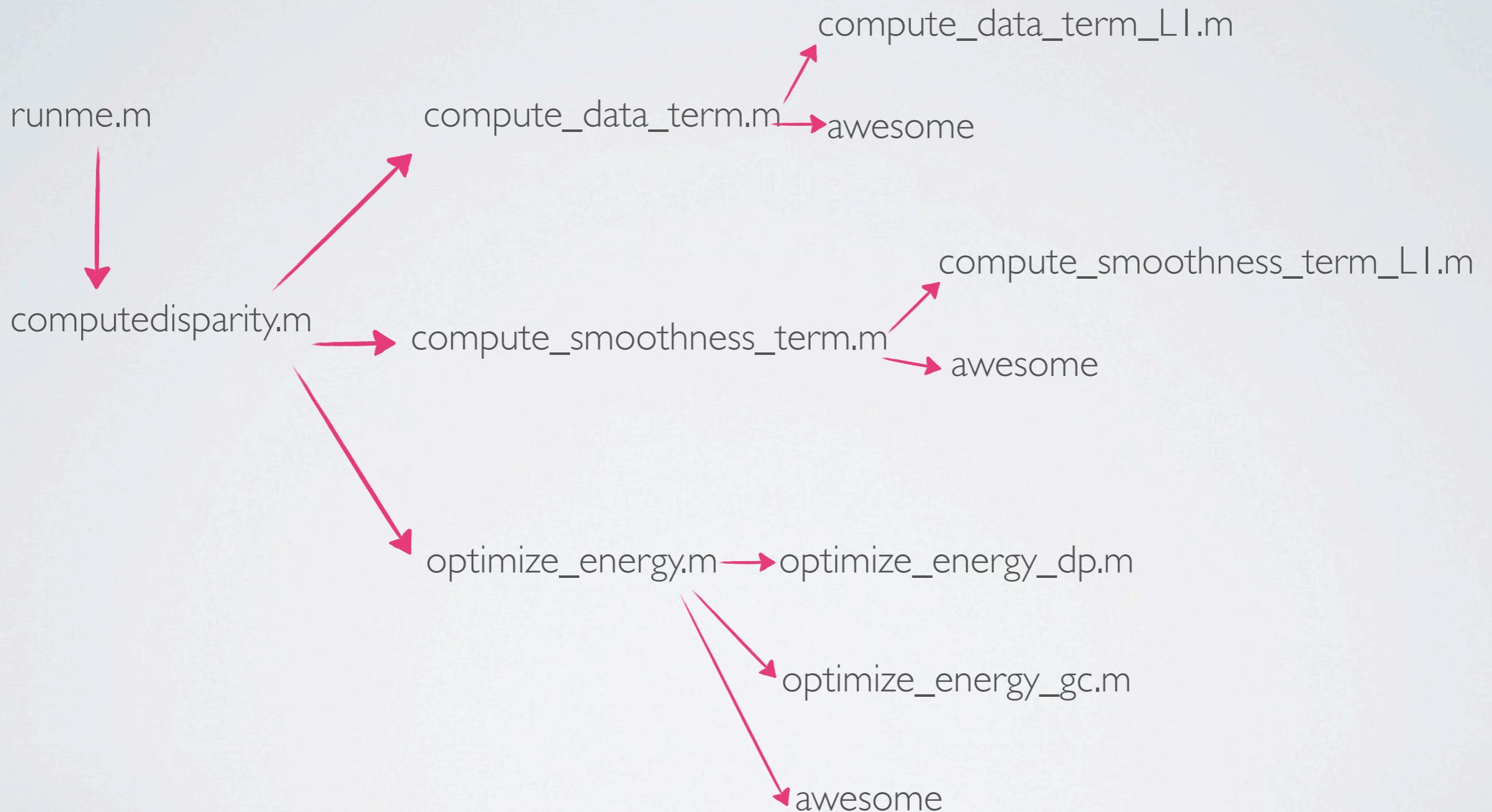
Implement at least  
one awesome

# HOW TO RUN

- runme.m

```
run_configurations = { ...
    % {'test01', 'L1', 'L1', 'graph_cut'}, ...
    % {'test01', 'awesome', 'L1', 'dynamic_programming'}, ...
    % {'test01', 'L1', 'awesome', 'dynamic_programming'}, ...
    % {'test01', 'L1', 'L1', 'awesome'}, ...
    {'all', 'L1', 'L1', 'dynamic_programming'}, ...
    {'all', 'L1', 'L1', 'graph_cut'}
};
```

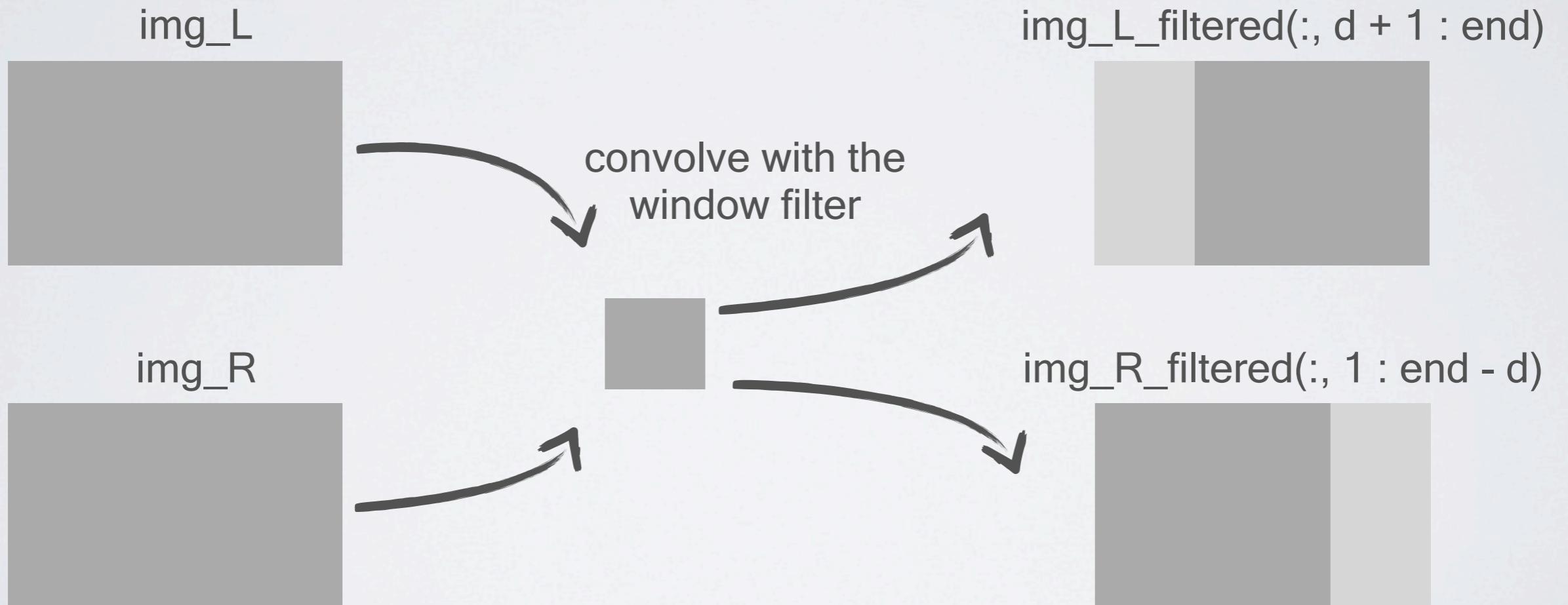
# HOW TO RUN



# TIPS AND SUGGESTIONS

# COMPUTE DATA TERM

$$data(y, x, d) = \lambda \cdot \min \left( \frac{\sum_{(y, x) \in \text{window}} |I_L(y, x) - I_R(y, x - d)|}{\text{window size}}, \tau \right)$$



# PARAMETER SETTING

- maximum\_data\_term\_value:  $\sim 10$
- maximum\_smoothness\_term\_value:  $\sim 1.7$
- max\_disparity =  $\sim 60$
- lambda/data\_term\_weight:  $\sim 0.04$

# GRAPH CUT

- GCMex
  - link: <http://vision.ucla.edu/~brian/gcmex.html>
- **[labels, energy, energyafter] = GCMex(class, unary, pairwise, labelcost, flag);**
- You might want to use diag() in MATLAB to create the adjacency matrix

START EARLY & GOOD LUCK!