



Master in Computer Vision *Barcelona*

Module: 3D Vision

Project: 3D recovery of urban scenes

Session 4

Gloria Haro

Session 3

Goal: Reconstruction from two images (known internal parameters)

Mandatory tasks:

- Triangulation with the homogeneous algebraic method (DLT)
- Compute the camera matrices from the Fundamental matrix and K
- Recover the 3D points by triangulation
- Compute the reprojection error
- Depth map computation by a local method (SSD cost)
- Depth map computation by a local method (NCC cost)
- Improve the matching cost by using bilateral weights

Optional tasks:

- Depth map computation by belief propagation
- Depth map computation by plane sweep
- Depth map fusion by averaging signed distance functions
- New view synthesis: view morphing

Session 3

Language: MATLAB

Provided functions: lab4.m, euclid.m, apply_H_v2.m, fundamental_matrix.m, ransac_fundamental_matrix.m, ncc_cost.m, normalise2dpts.m, optical_center.m

lab4.m is the guided file with the different steps of the lab session.

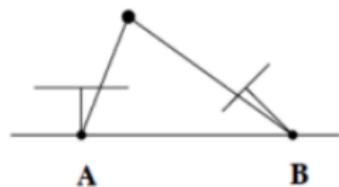
fundamental_matrix.m, ransac_fundamental_matrix.m are part of the solution of lab 3.

To Do:

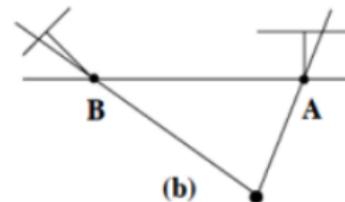
- Complete the code in lab4.m as indicated in the same file
- Write the functions triangulate.m and stereo_computation.m
- In the report, comment the results as asked in lab4.m

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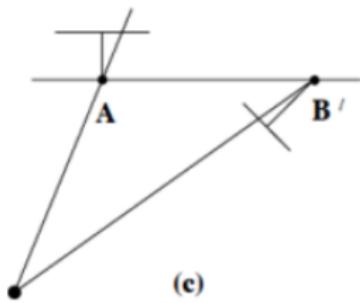
Compute the camera matrices from the Fundamental matrix and K



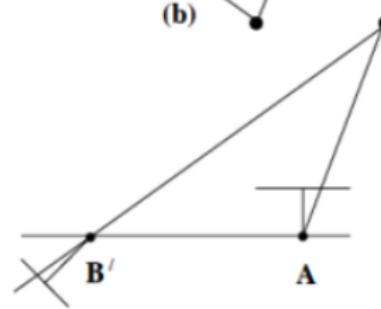
(a)



(b)



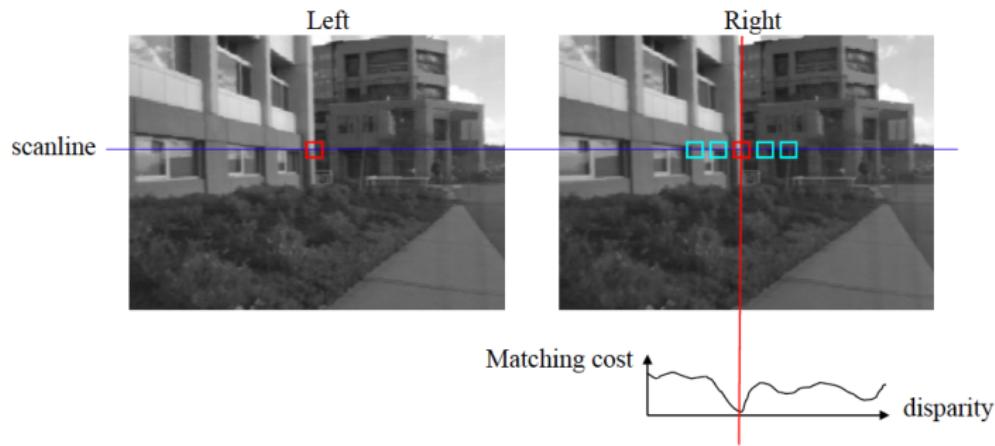
(c)



(d)

Session 3

Depth map computation by a local method with **stereo rectified images**



- Try different window sizes
- Try SSD and NCC costs

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Depth map computation by a local method with stereo rectified images

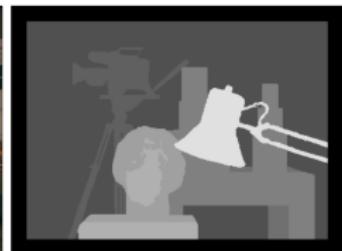
1st experiment: pair of Middlebury images with ground truth data



(a) Reference image.



(b) Target image.



(c) Ground-truth disparity map.

2nd experiment: Castle facade images



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Depth map computation by a local method with **stereo rectified images**

Improve the disparity computation by using weights based on color similarity and spatial distance (as those used in the bilateral filter) using the reference paper below.

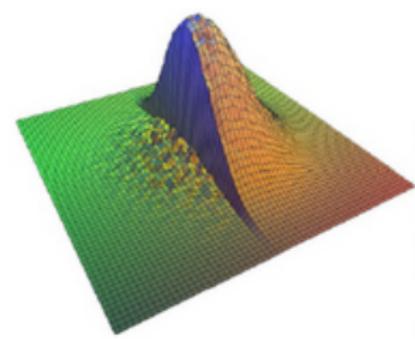
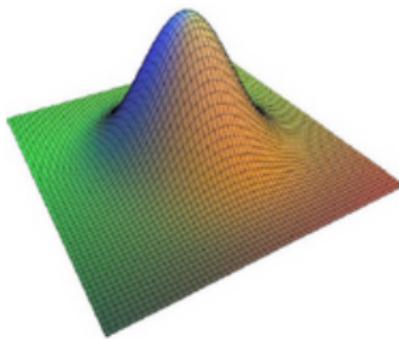
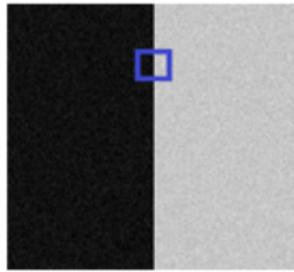
Adaptive Support-Weight Approach for Correspondence Search

Kuk-Jin Yoon, *Student Member, IEEE*, and
In So Kweon, *Member, IEEE*

IEEE TRANSACTIONS ON PATTERN ANALYSIS AND MACHINE INTELLIGENCE, VOL. 28, NO. 4, APRIL 2006

Session 3

Gaussian vs Bilateral weights



Session 3

Bilateral weights

Weights based on color similarity and spatial distance.

$$w(p, q) = \exp \left(-\frac{|I(p) - I(q)|}{\gamma_c} - \frac{\|p - q\|_2}{\gamma_p} \right)$$

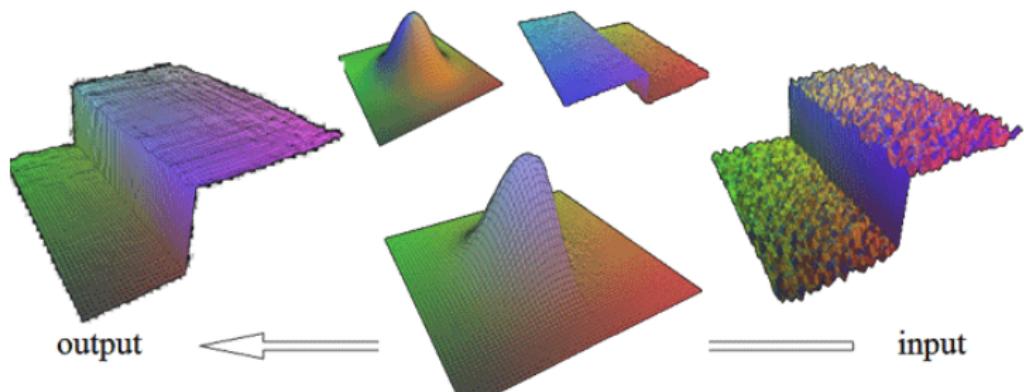
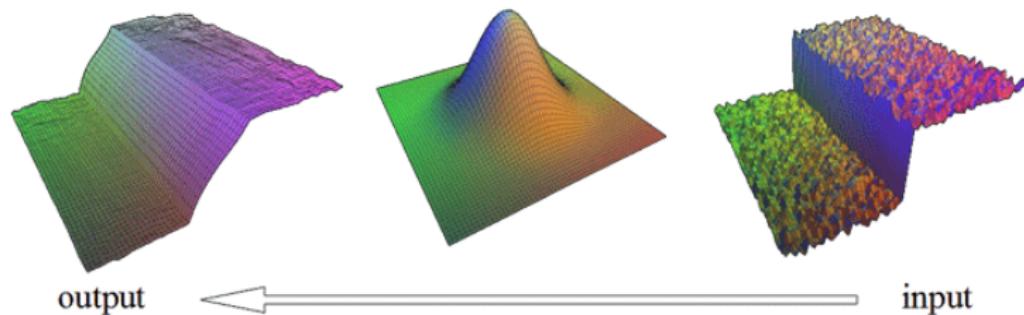
Modified **cost function**:

$$C(p, p_d) = \frac{\sum_{q \in N_p, q_d \in N_{p_d}} w(p, q) w(p_d, q_d) c(q, q_d)}{\sum_{q \in N_p, q_d \in N_{p_d}} w(p, q) w(p_d, q_d)}$$

where $p_d = p + (d, 0)^T$, $q_d = q + (d, 0)^T$, and d is the disparity.

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Gaussian vs Bilateral filtering



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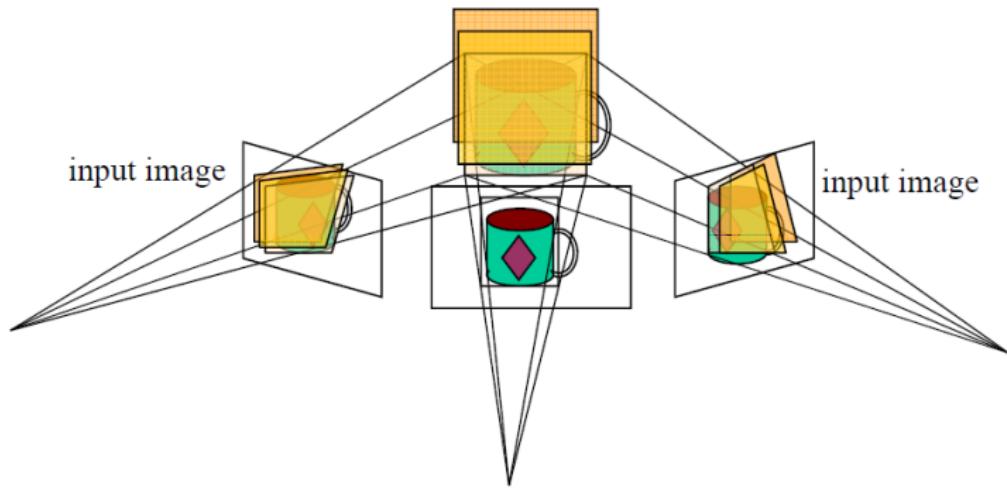
Depth map computation by belief propagation

Use a discrete energy with data and regularization terms and solve it by belief propagation as in module 2.

- Unary potentials: SSD or SAD
- Pairwise potentials: Potts model

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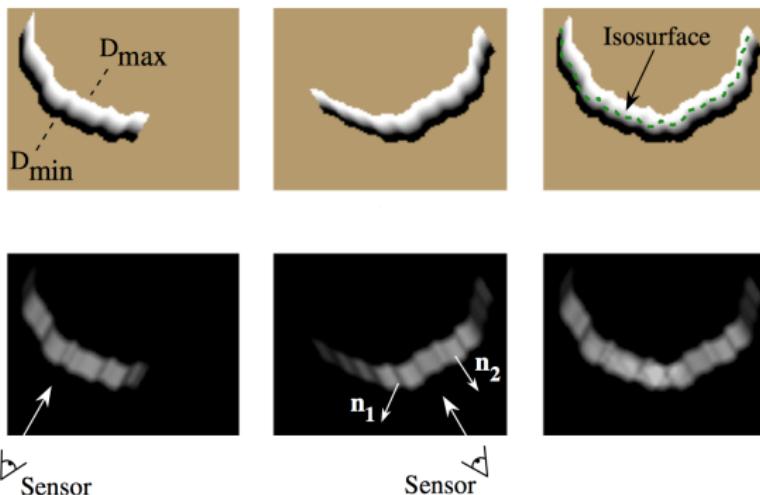
Depth map computation by plane sweep



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Depth map fusion by averaging signed distance functions

$$D(\mathbf{z}) = \frac{\sum_i w_i(\mathbf{z}) \text{sdf}_i(\mathbf{z})}{\sum_i w_i(\mathbf{z})}$$



No need to extract isosurface.

Just consider as surface voxels \mathbf{z} those satisfying $D(\mathbf{z}) \approx 0$

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Depth map fusion by averaging signed distance functions

In a first step you need to estimate depth maps from pairs of images in general position.

Possibilities for estimating the depth maps:

- Plane sweep method
- Local method + Stereo rectification method
http://demo.ipol.im/demo/m_quasi_euclidean_epipolar_rectification/
(Returns both rectified images and their corresponding rectifying homographies)

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New view synthesis: view morphing

Pair of stereo rectified (Middlebury) images with ground truth disparities

$$\mathbf{p} = (1 - s)(x, y) + s(x - d_\ell(x, y), y)$$

$$I(\mathbf{p}) = (1 - s) I_\ell(x, y) + s I_r(x - d_\ell(x, y), y)$$



Take into account occlusions!

Left-right consistency check: $\|d_\ell(x, y) - d_r(x - d_\ell(x, y), y)\| \leq \epsilon$

Evaluation

To deliver **before 9am of the day before** the next lab session:

- **Code deliverable:**
 - READY TO BE LAUNCHED on the provided images
- **Short document (around 10 pages):**
 - Results
 - Problems and comments

Evaluation

Grading:

- Report: **1.5 points**
- Triangulate function: **2 points**
- P, P' from F and K : **1 point**
- Reprojection error: **0.5 points**
- Depth map with SSD: **2 points**
- Depth map with NCC: **1 point**
- Bilateral weights: **1 point**
- Depth maps with facade images: **1 point**
- (Belief propagation): **+1.5 points**
- (Plane sweep): **+1.5 points**
- (Depth map fusion): **+1.5 points**
- (New view synthesis): **+1.5 points**