Multi-view Stereo

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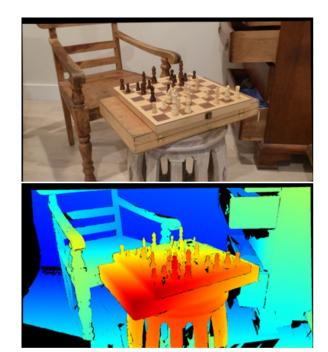
Multi-view Stereo

Goal: recover the disparity/depth map of the scene from two views

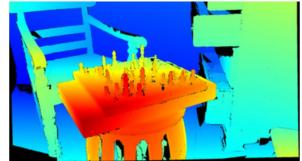
- Step 1: Camera calibration
 - Get the calibration matrix K
- Step 2: Estimation of the fundamental matrix F (sparse stereo matching)
 - 8-point algorithm with RANSAC
 - Extract the relative rotation and translation between two cameras
- Step 3: Dense stereo matching (window-based 1D search)
 - Given a point in one image, the corresponding point in the other image must lie on the **epipolar line** -> 1D search through the epipolar line
- Step 4: Depth map reconstruction
 - For each point in the reference image, compute the disparity between two images and the the depth is inversely proportional to the disparity

Middlebury Stereo Dataset

- A set of rectified two-view image pairs
- For each pair, we have
 - Left and right images
 - Ground truth disparity maps for both images
 - Calibration information including
 - **Intrinsic parameters**: the calibration matrix *K*
 - Extrinsic parameters: the baseline d
 - Image size: width, height
 - A conservative bound on the number of disparity levels







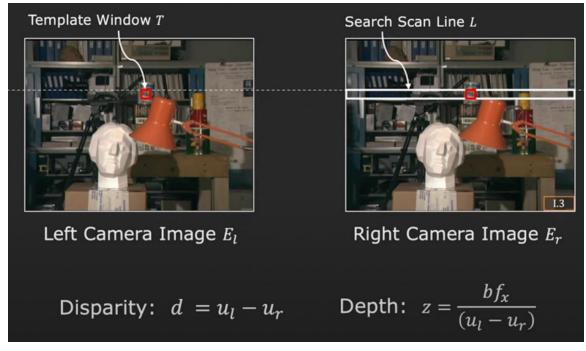
Multi-view Stereo

Goal: recover the disparity/depth map of the scene from two views

Step 1: Camera calibration



- Step 2: Estimation of the fundamental matrix F
- Step 3: Dense stereo matching
 - 1D search through the horizontal line
- Step 4: Depth map reconstruction



https://www.youtube.com/watch?v=hUVyDabn1Mg

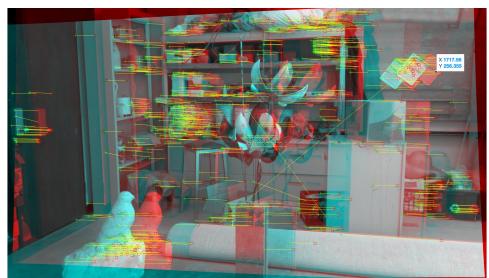
Estimation of the Fundamental Matrix

• Step 1: Find a set of corresponding features in left and right images using SIFT $\{(x_i, y_i)\}$





I_left



Correspondence Visualization

I_right

Estimation of the Fundamental Matrix

• Step 2:

Repeat n times {

- Randomly choose 8 matching pairs and solve for the following equation exactly by computing the null space of A
- Find the consensus set for the fundamental matrix in the set $\{(x_i, y_i)\}$.
- If the consensus set is sufficiently large, then we could refit the model using all points in the consensus set.
- Save the best model}

$$\begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Our estimation of F:

0.00000.00000.0009-0.00000.00000.3335-0.0010-0.33370.8887

Stereo Calibration

• Goal: extract the translation matrix T and rotation matrix R that represent the relative geometry between two cameras given the fundamental matrix F

$$\mathsf{E} = [T]_X R$$

• Theoretically, in our dataset, we only have a pure translation that is parallel to the

 The rotation matrix R and translation matrix T extracted from our estimated fundamental matrix using SVD:

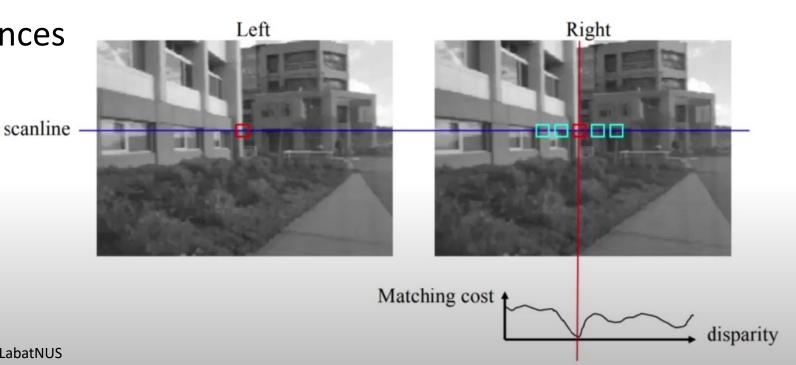
```
\begin{array}{ccccc} 1.000 & 0.0002 & 0.0000 \\ R = -0.0002 & 1.0000 & 0.0014 \\ -0.0000 & -0.0014 & 1.0000 \end{array}
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T = 0.9996 -0.0114 0.0274

Correspondence Search

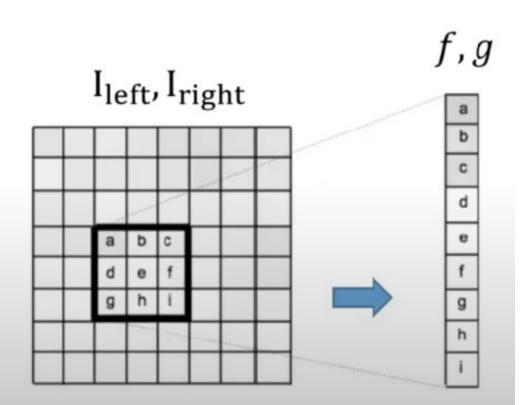
Methods to compute the matching cost:

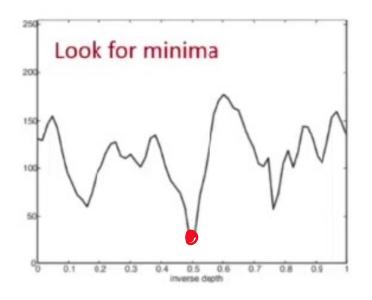
- Normalized Cross Correlation
- Sum of Squared Differences
- Sum of Absolute Differences
- Mutual Information



Sum of Squared Differences

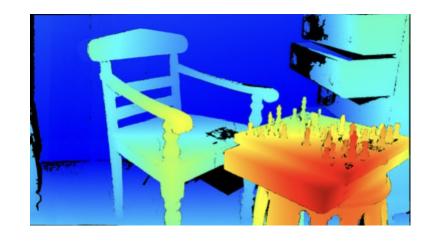
$$\rho_{SSD}(f,g) = ||f - g||^2$$

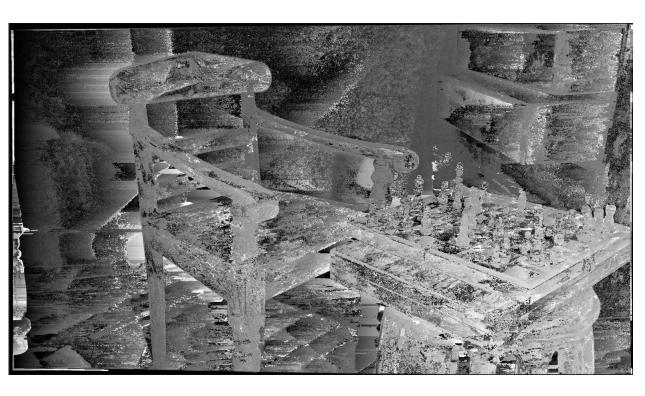


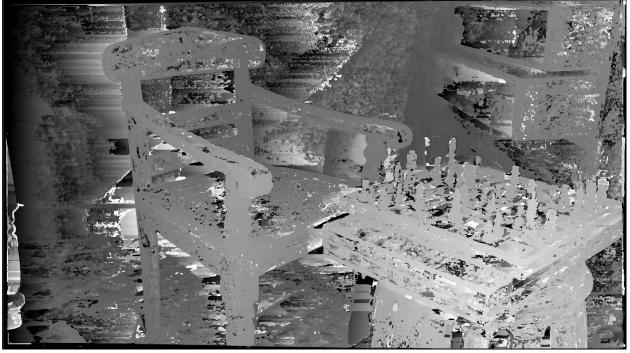


Chess room

Conservative estimation: 310







window size: 3 window size: 6

OpenCV depth reconstruction

Presented by Tad

StereoBM class and parameters

- Calculates a disparity map for a pair of rectified stereo images
- SAD as similarity metric
- Most important parameters
 - numDisparities
 - blockSize
- Demo