# Cameras and Stereo 

ECE P 596<br>Linda Shapiro

## Camera parameters

A camera is described by several parameters

- Translation T of the optical center from the origin of world coords
- Rotation R of the image plane
- focal length $f$, principal point ( $x^{\prime}{ }_{c}, y^{\prime}{ }_{c}$ ), pixel size ( $s_{x}, s_{y}$ )
- blue parameters are called "extrinsics," red are "intrinsics"

Projection equation

$$
\mathbf{x}=\left[\begin{array}{c}
w x \\
w y \\
w
\end{array}\right]=\left[\begin{array}{llll}
* & * & * & * \\
* & * & * & * \\
* & * & * & *
\end{array}\right]\left[\begin{array}{c}
X \\
Y \\
Z \\
1
\end{array}\right]=\boldsymbol{\Pi} \mathbf{X}
$$



- The projection matrix models the cumulative effect of all parameters
- Useful to decompose into a series of operations
- The definitions of these parameters are not completely standardized
- especially intrinsics-varies from one book to another


## Where does all this lead?

- We need it to understand stereo
- And 3D reconstruction
- It also leads into camera calibration, which is usually done in factory settings to solve for the camera parameters before performing an industrial task.
- The extrinsic parameters must be determined.
- Some of the intrinsic are given, some are solved for, some are improved.


## Camera Calibration



The idea is to snap images at different depths and get a lot of 2D-3D point correspondences.

x1, y1, z1, u1, v1<br>$\mathrm{x} 2, \mathrm{y} 2, \mathrm{z} 1, \mathrm{u} 2$, v2

xn, yn, zn, un, vn
Then solve a system of equations to get camera parameters.

## Stereo



## Amount of horizontal movement is

...inversely proportional to the distance from the camera


## Depth from Stereo

- Goal: recover depth by finding image coordinate $x^{\prime}$ that corresponds to $x$



## Depth from disparity



$$
\text { disparity }=x-x^{\prime}=\frac{B \cdot f}{z}
$$

Disparity is inversely proportional to depth.

## Depth from Stereo

- Goal: recover depth by finding image coordinate $x^{\prime}$ that corresponds to x
- Sub-Problems

1. Calibration: How do we recover the relation of the cameras (if not already known)?
2. Correspondence: How do we search for the matching point $x^{\prime}$ ?


## Correspondence Problem



- We have two images taken from cameras with different intrinsic and extrinsic parameters
- How do we match a point in the first image to a point in the second? How can we constrain our search?


## Key idea: Epipolar constraint



Potential matches for $x$ have to lie on the corresponding line $l$ '.

Potential matches for $x$ ' have to lie on the corresponding line $l$.

## Epipolar geometry: notation



- Baseline - line connecting the two camera centers
- Epipoles
= intersections of baseline with image planes
$=$ projections of the other camera center
- Epipolar Plane - plane containing baseline (1D family)


## Epipolar geometry: notation



- Baseline - line connecting the two camera centers
- Epipoles
= intersections of baseline with image planes
$=$ projections of the other camera center
- Epipolar Plane - plane containing baseline (1D family)
- Epipolar Lines - intersections of epipolar plane with image planes (always come in corresponding pairs)


## Example: Converging cameras



## Example: Motion parallel to image plane



## Epipolar constraint



- If we observe a point $\boldsymbol{x}$ in one image, where can the corresponding point $\boldsymbol{x}^{\prime}$ be in the other image?


## Epipolar constraint



- Potential matches for $\boldsymbol{x}$ have to lie on the corresponding epipolar line $l$ '.
- Potential matches for $\boldsymbol{x}$ ' have to lie on the corresponding epipolar line $\boldsymbol{l}$.


## Epipolar constraint example



## Epipolar constraint: Calibrated case



- Assume that the intrinsic and extrinsic parameters of the cameras are known
- We can multiply the projection matrix of each camera (and the image points) by the inverse of the calibration matrix to get normalized image coordinates
- We can also set the global coordinate system to the coordinate system of the first camera. Then the projection matrices of the two cameras can be written as $[\mathbf{I} \mid \mathbf{0}]$ and $[\mathbf{R} \mid \mathbf{t}]$


## Simplified Matrices for the 2 Cameras

$$
\begin{aligned}
& \left(\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right)=(\mathbf{I} \mid \mathbf{0}) \\
& \left(\begin{array}{l|l}
\mathbf{R} & \mathbf{t} \\
\hline \mathbf{0} & 1
\end{array}\right)=(\mathrm{R} \mid \mathrm{T})
\end{aligned}
$$

## Epipolar constraint: Calibrated case



The vectors $R x, t$, and $x$ ' are coplanar

## Epipolar constraint: Calibrated case



Essential Matrix E (Longuet-Higgins, 1981)

The vectors $\boldsymbol{R} \boldsymbol{x}, \mathrm{t}$, and $\boldsymbol{x}$ ' are coplanar

## Epipolar constraint: Calibrated case



- $\boldsymbol{E} \boldsymbol{x}$ is the epipolar line associated with $\boldsymbol{x}\left(I^{\prime}=\boldsymbol{E} \boldsymbol{x}\right)$
- $\boldsymbol{E}^{\top} \boldsymbol{x}^{\prime}$ is the epipolar line associated with $\boldsymbol{x}^{\prime}\left(\boldsymbol{I}=\boldsymbol{E}^{\top} \boldsymbol{x}^{\prime}\right)$
- $\boldsymbol{E} \boldsymbol{e}=0$ and $\boldsymbol{E}^{\top} \boldsymbol{e}^{\prime}=0$
- $\boldsymbol{E}$ is singular (rank two)
- $E$ has five degrees of freedom


## Epipolar constraint: Uncalibrated

## case



## Comparison of estimation



|  | 8-point | Normalized 8-point | Nonlinear least squares |
| :--- | :--- | :--- | :--- |
| Av. Dist. 1 | 2.33 pixels | 0.92 pixel | 0.86 pixel |
| Av. Dist. 2 | 2.18 pixels | 0.85 pixel | 0.80 pixel |

## Basic stereo matching algorithm



- If necessary, rectify the two stereo images to transform epipolar lines into scanlines
- For each pixel x in the first image
- Find corresponding epipolar scanline in the right image
- Search the scanline and pick the best match $x^{\prime}$
- Compute disparity $x-x^{\prime}$ and set depth $(x)=f B /\left(x-x^{\prime}\right)$


## Stereo image rectification



## Stereo image rectification

- Reproject image planes onto a common plane parallel to the line between camera centers
- Pixel motion is horizontal after this transformation
- Two homographies ( $3 \times 3$ transform), one for each input image reprojection
> C. Loop and Z. Zhang. Computing Rectifying Homographies for Stereo Vision. IEEE Conf. Computer Vision and Pattern Recognition, 1999.


## Example

## Unrectified



## Rectified




- Slide a window along the right scanline and compare contents of that window with the reference window in the left image
- Matching cost: SSD, SAD, or normalized correlation


## Correspondence search



## Correspondence search



Norm. corr

## Failures of correspondence search



Textureless surfaces


Occlusions, repetition


Non-Lambertian surfaces, specularities

## Results with window search <br> Data



Window-based matching
Ground truth


## Priors and constraints

- Uniqueness
- For any point in one image, there should be at most one matching point in the other image
- Ordering
- Corresponding points should be in the same order in both views
- Smoothness
- We expect disparity values to change slowly (for the most part)


## Real-time stereo



Nomad robot searches for meteorites in Antartica http://www.frc.ri.cmu.edu/projects/meteorobot/index.html

- Used for robot navigation (and other tasks)
- Several software-based real-time stereo


## Stereo reconstruction pipeline <br> - Steps

- Calibrate cameras
- Rectify images
- Compute disparity
- Estimate depth

What will cause errors?

- Camera calibration errors
- Poor image resolution
- Occlusions
- Violations of brightness constancy (specular reflections)
- Large motions
- Low-contrast image regions



## Using more than two images



Multi-View Stereo for Community Photo Collections M. Goesele, N. Snavely, B. Curless, H. Hoppe, S. Seitz
 Proceedings of ICCV 2007,

