# Introduction to Computer Vision 

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## What is Computer Vision?

## Computer Graphics



Images

## Computer Vision



The inverse problem: Generate 3D model from images.

## Multiview Reconstruction

Given Images


4 images out of a sequence with 435 images.


## Reconstruction Pipeline

Point Detection and Matching


## Reconstruction Pipeline

Point Detection and Matching


Detect interesting (descriptive) points in all images.

## Reconstruction Pipeline

Point Detection and Matching


Match points between images.

## Reconstruction Pipeline

## Geometric Computations (main part of the Computer Vison course!)



Compute 3D-positions of the matched points, position and orientation of the cameras.

## Video




## Camera Model

## The Pinhole Camera

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Sic nos exactè Anno.1544. Louanii cclipfim Solis obferuauimus, inuenimuś́; deficere paulò plus $\not$ ঞ̈ dex-

## Using a pinhole camera to create an image



Reinerus Gemma-Frisus
camera obscura from 1544.

## Camera Model

## The Pinhole Camera



## Pinhole Projection



$$
\begin{aligned}
C=\left(\begin{array}{l}
0 \\
0 \\
0
\end{array}\right) & \Rightarrow \lambda\left(\begin{array}{c}
x_{1} \\
x_{2} \\
1
\end{array}\right)=\left(\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right), \lambda \in \mathbb{R} \\
& \Rightarrow\left(x_{1}, x_{2}\right)=\left(\frac{x_{1}}{X_{3}}, \frac{x_{2}}{x_{3}}\right)
\end{aligned}
$$

## Ex 1

Compute the image of the cube with corners in $( \pm 1, \pm 1,2)$ and $( \pm 1, \pm 1,4)$.

## Homogeneous Coordinates \& Moving Cameras

Projections with homogeneous coord:

$$
\lambda \mathbf{x}=\left[\begin{array}{ll}
I & 0
\end{array}\right] \mathbf{X}, \quad \text { where } \mathbf{x}=\left(\begin{array}{c}
x_{1} \\
x_{2} \\
1
\end{array}\right), \mathbf{X}=\left(\begin{array}{c}
x_{1} \\
x_{2} \\
x_{3} \\
1
\end{array}\right),
$$

Camera motion can be modeled using

$$
\lambda \mathbf{x}=\left[\begin{array}{ll}
R & t
\end{array}\right] \mathbf{X}
$$

Extrinsic parameters: $R$-rotation matrix, $t$-translation vector.

## Ex 2

Compute the projection of $X=(0,0,1)$ in

$$
P_{1}=\left(\begin{array}{cccc}
\frac{1}{\sqrt{2}} & 0 & -\frac{1}{\sqrt{2}} & 0 \\
0 & 1 & 0 & 0 \\
\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 1
\end{array}\right) \text { and } \sqrt{2} P_{1} .
$$

## The Inner Parameters - K

$$
\left(\begin{array}{c}
f x+x_{0} \\
f y+y_{0} \\
1
\end{array}\right)=\left[\begin{array}{ccc}
f & 0 & x_{0} \\
0 & f & y_{0} \\
0 & 0 & 1
\end{array}\right]\left(\begin{array}{c}
x \\
y \\
1
\end{array}\right)
$$

$f$-focal length, $\left(x_{0}, y_{0}\right)$ principal point. Re-centers and scales (e.g. meters $\rightarrow$ pixels) the image. Typically transforms the point $(0,0,1)$ to the middle of the image.


## The Inner Parameters - $K$

The most general version of $K$ is the upper triangular matrix:

$$
K=\left[\begin{array}{ccc}
\gamma f & s f & x_{0} \\
0 & f & y_{0} \\
0 & 0 & 1
\end{array}\right]
$$

- $f$ - focal length
- $\gamma$ - aspect ratio
- s-skew
- $\left(x_{0}, y_{0}\right)$ - principal points

Camera equations:

$$
\lambda \mathbf{x}=K\left[\begin{array}{ll}
R & t
\end{array}\right] \mathbf{X}=P \mathbf{X}
$$

RQ factorization: Any $3 \times 4$ matrix $P$ can be written $K[R$ with $K$ triangular and $R$ orthogonal.

## Relative Orientation: Problem Formulation

## Given



Two images and corresponding points.

## Compute



The structure (3D-points) and the motion (camera matrices).

## Relative Orientation

## Problem Formulation

Given corresponding points $\mathbf{x}_{i}$ and $\overline{\mathbf{x}}_{i}, i=1, \ldots, n$ find cameras $P_{1}$ and $P_{2}$ and 3D points $\mathbf{X}_{i}$ such that

$$
\lambda_{i} \mathbf{x}_{i}=P_{1} \mathbf{X}_{i}
$$

and

$$
\bar{\lambda}_{i} \overline{\mathbf{x}}_{i}=P_{2} \mathbf{X}_{i} .
$$

## Ambiguities (uncalibrated case)

Can always apply a projective transformation $H$ to archive a different solution

$$
\lambda_{i} \mathbf{x}_{i}=P_{1} H H^{-1} \mathbf{X}_{i}=\tilde{P}_{1} \tilde{\mathbf{X}}_{i}
$$

and

$$
\bar{\lambda}_{i} \overline{\mathbf{x}}_{i}=P_{2} H H^{-1} \mathbf{X}_{i}=\tilde{P}_{2} \tilde{\mathbf{x}}_{i}
$$

## Relative Orientation: Problem Formulation

## Simplification

If $P_{1}=\left[\begin{array}{ll}A_{1} & t_{1}\end{array}\right]$ and $P_{2}=\left[\begin{array}{ll}A_{2} & t_{2}\end{array}\right]$, apply the transformation

$$
H=\left[\begin{array}{cc}
A_{1}^{-1} & -A_{1}^{-1} t_{1} \\
0 & 1
\end{array}\right] .
$$

Then

$$
P_{1} H=\left[\begin{array}{ll}
A_{1} & t_{1}
\end{array}\right]\left[\begin{array}{cc}
A_{1}^{-1} & -A_{1}^{-1} t_{1} \\
0 & 1
\end{array}\right]=\left[\begin{array}{ll}
I & 0
\end{array}\right]
$$

Hence, we may assume that the cameras are

$$
P_{1}=\left[\begin{array}{ll}
I & 0
\end{array}\right] \text { and } P_{2}=\left[\begin{array}{ll}
A & t
\end{array}\right]
$$

Still hard to solve since $P_{2}$ and $\mathbf{X}_{i}$ are unknown!

## Epipolar Geometry



Consider a single point $\mathbf{x}$ in the first image. Any point on the line projects to this point.

## Epipolar Geometry



Any point on the projection of the 3D line can correspond to $\mathbf{x}$.

## Epipolar Geometry



## Epipolar Geometry



The projected lines should all meet in a point. The so called epipole is the projection of the camera center of the other camera.

## Epipolar Geometry



The epipole $e_{1}$ is the projection of the $C_{2}$ in $P_{1}$. The epipole $e_{2}$ is the projection of the $C_{1}$ in $P_{2}$. $e_{1}, e_{2}$ usually outside field of view.

## The Fundamental Matrix

If $\mathbf{x}$ and $\overline{\mathbf{x}}$ are projections of $X$ in $P_{1}=\left[\begin{array}{ll}I & 0\end{array}\right]$ and $P_{2}=\left[\begin{array}{ll}A & t\end{array}\right]$ then

$$
\overline{\mathbf{x}}^{T} F \mathbf{x}=0
$$

where $F=[t]_{\times} A$.

- $F$ - Fundamental matrix.
- $\overline{\mathbf{x}}^{T} F \mathbf{x}=0$ - epipolar constraint,

$$
e_{2}^{T} F \mathbf{x}=0 \quad \forall \bar{x} \Rightarrow e_{2}^{T} F=0 \Rightarrow \operatorname{det}\left(F^{T}\right)=\operatorname{det}(F)=0
$$

Search for the fundamental matrix song on youtube!

## Ex 3

If $P_{1}=\left[\begin{array}{ll}I & 0\end{array}\right]$ and $P_{2}=\left[\begin{array}{ll}I & t\end{array}\right]$, where $t=\left(\begin{array}{l}1 \\ 1 \\ 1\end{array}\right)$, which of $\overline{\mathbf{x}}=\left(\begin{array}{l}1 \\ 2 \\ 1\end{array}\right)$ and
$\overline{\mathbf{y}}=\left(\begin{array}{l}3 \\ 2 \\ 1\end{array}\right)$ (in image 2) can correspond to $\mathbf{x}=\left(\begin{array}{c}1 \\ -1 \\ 1\end{array}\right)$ (in image 1 )?

## The Fundamental Matrix

The epipolar constraints only contain camera information. The 3D-points have been eliminated.

## Estimating $F$

If $\mathbf{x}_{i}$ and $\overline{\mathbf{x}}_{i}$ corresponding points

$$
\overline{\mathbf{x}}_{i}^{T} F \mathbf{x}_{i}=0
$$

If $\mathbf{x}_{i}=\left(x_{i}, y_{i}, z_{i}\right)$ and $\overline{\mathbf{x}}_{i}=\left(\bar{x}_{i}, \bar{y}_{i}, \bar{z}_{i}\right)$ then

$$
\begin{aligned}
\overline{\mathbf{x}}_{i}^{T} F \mathbf{x}_{i}= & F_{11} \bar{x}_{i} x_{i}+F_{12} \bar{x}_{i} y_{i}+F_{13} \bar{x}_{i} z_{i} \\
& +F_{21} \bar{y}_{i} x_{i}+F_{22} \bar{y}_{i} y_{i}+F_{233} \bar{y}_{i} z_{i} \\
& +F_{31} \bar{z}_{i} x_{i}+F_{32} \bar{z}_{i} y_{i}+F_{33} \bar{z}_{i} z_{i}
\end{aligned}
$$

## The Fundamental Matrix

## Estimating $F$

In matrix form (one row for each correspondence):

$$
\underbrace{\left[\begin{array}{ccccc}
\bar{x}_{1} x_{1} & \bar{x}_{1} y_{1} & \bar{x}_{1} z_{1} & \ldots & \bar{z}_{1} z_{1} \\
\bar{x}_{2} x_{2} & \bar{x}_{2} y_{2} & \bar{x}_{2} z_{2} & \ldots & \bar{z}_{2} z_{2} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
\bar{x}_{n} x_{n} & \bar{x}_{n} y_{n} & \bar{x}_{n} z_{n} & \cdots & \bar{z}_{n} z_{n}
\end{array}\right]}_{M}\left[\begin{array}{c}
F_{11} \\
F_{12} \\
F_{13} \\
\vdots \\
F_{33}
\end{array}\right]=\left[\begin{array}{c}
0 \\
0 \\
0 \\
\vdots \\
0
\end{array}\right]
$$

Solve using homogeneous least squares (svd).
$F$ has 9 entries (but the scale is arbitrary). Need at least 8 equations (point correspondences).

## The Fundamental Matrix

## Issues

Resulting $F$ may not have $\operatorname{det}(F)=0$.
Pick the closest matrix $A$ with $\operatorname{det}(A)=0$.
Can be solved using svd, in matlab:

$$
\begin{aligned}
& {[U, S, V]=\operatorname{svd}(F)} \\
& S(3,3)=0 \\
& A=U * S * V^{\prime}
\end{aligned}
$$



## The Fundamental Matrix

## Issues

Normalization needed (see DLT).
If $x_{1}$ and $\bar{x}_{1} \approx 1000$ pixels, the coefficients $z_{1} \bar{z}_{1}=1, x_{1} \bar{z}_{1}=1000$ and $x_{1} \bar{x}_{1}=1000000$. May give poor numerics.

Not normalized:


Normalized:


## Extracting cameras from F

A family of solutions:

$$
\begin{gathered}
P_{1}=\left[\begin{array}{ll}
I & 0
\end{array}\right] \\
P_{2}=\left[\begin{array}{ll}
{\left[e_{2}\right]_{\times} F+e_{2} v^{T}} & \lambda e_{2}
\end{array}\right],
\end{gathered}
$$

$e_{2} \in \operatorname{Null}(F), v \in \mathbb{R}^{3}, \lambda \in \mathbb{R}$.

## Calibrated Relative Orientation: Problem Formulation

## Simplification

If $P_{1}=\left[\begin{array}{ll}R_{1} & t_{1}\end{array}\right]$ and $P_{2}=\left[\begin{array}{ll}R_{2} & t_{2}\end{array}\right]$, apply the transformation

$$
H=\left[\begin{array}{cc}
R_{1}^{T} & -R_{1}^{T} t_{1} \\
0 & 1
\end{array}\right] .
$$

Then

$$
P_{1} H=\left[\begin{array}{ll}
R_{1} & t_{1}
\end{array}\right]\left[\begin{array}{cc}
R_{1}^{T} & -R_{1}^{T} t_{1} \\
0 & 1
\end{array}\right]=\left[\begin{array}{ll}
I & 0
\end{array}\right] .
$$

Hence, we may assume that the cameras are

$$
P_{1}=\left[\begin{array}{ll}
I & 0
\end{array}\right] \text { and } P_{2}=\left[\begin{array}{ll}
R & t
\end{array}\right]
$$

## The Essential Matrix

## The Essential Matrix

The camera pair $P_{1}=\left[\begin{array}{ll}I & 0\end{array}\right]$ and $P_{2}=\left[\begin{array}{ll}R & t\end{array}\right]$ has the fundamental matrix

$$
E=[t]_{\times} R
$$

$E$ is called the essential matrix.

- $R$ has 3 dof, $t 3$ dof, but the scale is arbitrary, therefore $E$ has 5 dof.
- $E$ has $\operatorname{det}(E)=0$
- $E$ has two nonzero equal singular values.

Solve using the (non-linear) 5-point solver. (Gives 10 solutions.)

## Computing the cameras

Want to find $P_{2}=[R t]$ such that $E=[t]_{\times} R$.
Outline:

- Ensure that E has the SVD

$$
E=U S V^{T}
$$

where $\operatorname{det}\left(U V^{T}\right)=1$.

- Compute a factorization $E=S R$ where $S$ is skew symmetric and $R$ a rotation.
- Compute a $t$ such that $[t]_{\times}=S$.
- Form the camera $P_{2}=[R t]$.



## 4 Solutions



