## "Bag of lines" models <br> for non-central linear cameras



## "Bag of lines" models

## for non-central linear cameras

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## Linear cameras


pinhole


## Goal

Stereo-vision between
ANY pair of
linear cameras


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PROBLEMS $\left\{\begin{array}{l}\text { unified model } \\ \text { computations }\end{array}\right.$

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Stereo-vision between

ANY pair of
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## PROBLEMS

## unified model computations



Admissible Map
[Pajdla,02]

## OUTLINE

## 1. Stereo-vision with pinhole cameras

2. The admissible map model
3. Stereo-vision with linear cameras
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## Inverse projection



Inverse projection


# Inverse projection 



Inverse projection

stereo correspondance
intersection of inverse projections


## Normalized coordinates



## Fundamental \& Essential matrix

Fundamental matrix

$$
\left(\begin{array}{l}
1 \\
u_{1} \\
u_{2}
\end{array}\right)
$$

$u^{T} \mathcal{F} u^{\prime}=0$
$3 \times 3$ matrix
8 correspondances needed to determine $\mathcal{F}$

## Essential matrix

$$
u^{T} \mathcal{E} u^{\prime}=0
$$

$\mathcal{E}$ depends only on $W$
6 correspondances needed to determine $\mathcal{E}$

## OUR APPROACH


OUR APPROACH
Camera
bag of lines $+$
retina

## OUTLINE

## 1. Stereo-vision with pinhole cameras

2. The admissible map model
3. Stereo-vision with linear cameras

## OUTLINE

## "bag of lines" level

1. Stereo-vision with pinhole cameras
2. The admissible map model
3. Stereo-vision with linear cameras
adding the retina

## Projective space

$\mathbb{P}^{3}(\mathbb{R}) \sim$ set of lines of $\mathbb{R}^{4}$ passing through the origin

Homogeneous coordinates


$$
\left[x_{0}: x_{1}: x_{2}: x_{3}\right] \sim\left[\lambda x_{0}: \lambda x_{1}: \lambda x_{2}: \lambda x_{3}\right]
$$

Intuition in the affine space $\mathbb{R}^{3}$

$$
\mathbb{P}^{3}(\mathbb{R})=\underset{[1: x: y: z]}{\text { points }}+\underset{[0: u: v: w]}{ }+\begin{gathered}
\text { directions } \\
\end{gathered}
$$



$$
\ell \cap P=\{[\vec{u}]\}
$$

## Linear congruences

## 3D field of view + order 1: $x \stackrel{!}{\mapsto} \ell \underset{\substack{\text { for almost } \\ \text { al } \\ \text { (fit }}}{\substack{\text { pen }}}$ $\Rightarrow$ 2-parameter set of lines

hyperbolic congruence

X-slit

camera
 pencil camera

## Projective

 classification
bundle pinhole camera

elliptic congruence linear oblique camera

degenerate congruence (no name)

## Linear congruences

## 3D field of view + order $1 \cdot x \stackrel{!}{\stackrel{ }{n}}$ (for almost all $x$ ) $\Rightarrow$ 2-parameter set of lines



## Admissible maps

Idea: a linear map $A$ that globally preserves each line of the bag


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$\Rightarrow \quad A^{2} x=\lambda_{x} A x+\mu_{x} x$
(for almost all $x$ )
$\Rightarrow \quad A^{2}=\lambda A+\mu \mathrm{Id} \quad$ (linear algebra) $A=\left(\begin{array}{l}* * * * \\ * * * * \\ * * * * \\ * * * *\end{array}\right)=\mathbb{R}^{4 \times 4}$
Definition: $A$ has a minimal polynomial $\pi_{A}$ of degree 2 .

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Definition: $A$ has a minimal polynomial $\pi_{A}$ of degree 2 .
$\Rightarrow \quad \mathcal{L}=\{x \vee A x$ for $x$ not an eigenvector $\}$ has order 1 .

Ambiguity locus

$$
\text { of } \mathcal{L}
$$

Union of eigenspaces of $A$

## Admissible maps

| $\pi_{A}$ | Eigenspaces | Reduced form of $A$ | $\mathcal{L}$ |
| :---: | :---: | :---: | :---: |
| $(X-\alpha)(X-\beta)$ | plane + point | $\left[\begin{array}{c\|c}\alpha I_{3} & 0 \\ \hline 0 & \beta\end{array}\right]$ |  |
| $(X-\alpha)(X-\beta)$ | 2 lines | $\left[\begin{array}{c\|c}\alpha I_{2} & 0 \\ \hline 0 & \beta I_{2}\end{array}\right]$ |  |
| $(X-\alpha)^{2}$ | plane | $\left[\right.$$\alpha I_{2}$ 0  <br> 0 $\alpha$ 0 <br>  $\lambda$ $\alpha$$]$ |  |
| $(X-\alpha)^{2}$ | 1 line | $\left[\right.$$\alpha$ 0 0  <br> $\lambda$ $\alpha$   <br> 0 $\alpha$ 0  <br>   $\mu$ $\alpha$$]$ |  |
| $\Delta<0$ | $\emptyset$ | $\left[\begin{array}{cc\|c}\alpha & -\beta & 0 \\ \beta & \alpha & 0 \\ \hline 0 & \left.\begin{array}{cc}\alpha & -\beta \\ \beta & \alpha\end{array}\right]\end{array}\right.$ |  |

## Geometric model

(linear congruences)
$\sqrt{V}$

## Analytical model

(admissible maps)

## Application：Ray－tracing（using pbrt）


class LinearCamera ：public Camera \｛
〈 LinearCamera Constructor〉
float GenerateRay（Sample \＆s，Ray＊r）；
Transform AdMap；
Transform RasterToWorld；
Vector vview；
〈Other Attributes 〉\};

## Application: Ray-tracing (using pbrt)


float GenerateRay (Sample \&s,Ray *r) \{
Point pras = Point (sample.imageX, sample.imageY,0);
Point pret = RasterToWorld(pras);
ray->o = pret;
ray->d = Normalize(AdMap(pret)-pret);
if (Dot (ray->d, vview) < 0) ray->d = -(ray->d);
〈Setting Ray Time and Endpoints 〉\};

## OUTLINE

1. Stereo-vision with pinhole cameras
2. The admissible map model
3. Stereo-vision with linear cameras 2 key ingredients: * inverse projection * normalized coordinates

## Inverse projection

$\forall x$ non-ambiguous $\quad x \longmapsto x \vee A x \in \mathcal{L}$ where $A$ is an admissible map for $\mathcal{L}$

$$
\begin{gathered}
x \vee A x=\left[\xi_{0}: \xi_{1}: \xi_{2}: \xi_{3}: \xi_{4}: \xi_{5}\right] \\
x=\sum_{i} u_{i} y^{i} \text { Inverse projection } \\
x_{y^{0} \times{ }^{\circ} y^{1} y^{2}} \quad \pi_{i}:\left(u_{i}\right) \longmapsto\left(\xi_{i}\right)
\end{gathered}
$$



## Inverse projection

$$
\begin{gathered}
\pi_{i}(u)=\left(\sum_{i=0}^{2} u_{i} y^{i}\right) \vee\left(\sum_{i=0}^{2} u_{i} A y^{i}\right) \\
\pi_{i}(u)=\sum_{i=0}^{2} \frac{u_{i}^{2}}{2} \zeta^{i i}+\sum_{0 \leq i<j \leq 2} u_{i} u_{j} \zeta^{i j} \\
\text { where } \zeta^{i j}=y^{i} \vee A y^{j}+y^{j} \vee A y^{i} \\
\widetilde{\mathcal{P}}=\left[\zeta^{00}, \zeta^{01}, \zeta^{02}, \zeta^{11}, \zeta^{12}, \zeta^{22}\right] \\
(6 \times 6 \text { matrix }) \\
\mu(u)=\left(u_{0}^{2}, u_{0} u_{1}, u_{0} u_{2}, u_{1}^{2}, u_{1} u_{2}, u_{2}^{2}\right)^{T} \\
\quad(6 \text {-vector })
\end{gathered}
$$

## Fundamental matrix



Image coordinates $u$ and $u^{\prime}$ are in stereo correspondance

$$
\pi_{i}(u) \odot \pi_{i}\left(u^{\prime}\right)=0
$$

$\xi=\left[\xi_{0}: \xi_{1}: \xi_{2}: \xi_{3}: \xi_{4}: \xi_{5}\right]^{T}$
$\xi \odot \xi^{\prime}=\xi^{T} \cdot\left(\xi^{\prime}\right)^{*}$
$\xi^{*}=\left[\xi_{3}: \xi_{4}: \xi_{5}: \xi_{0}: \xi_{1}: \xi_{2}\right]^{T}$
side-operator

$$
\ell \cap \ell^{\prime} \neq \varnothing \quad \Leftrightarrow \quad \xi \odot \xi^{\prime}=0
$$

## Fundamental matrix



Image coordinates $u$ and $u^{\prime}$ are in stereo correspondance


$$
\pi_{i}(u) \odot \pi_{i}\left(u^{\prime}\right)=0
$$

$\mu(u)^{T} \widetilde{\mathcal{P}}^{T}\left(\widetilde{\mathcal{P}}^{\prime}\right)^{*} \mu\left(u^{\prime}\right)=0$
$6 \times 6$ fundamental matrix 35 correspondances needed

## Normalized coordinates

Idea: the bag of lines is spanned by at most 4 lines

- express $\pi_{i}$ in that "base" of lines


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Idea: the bag of lines is spanned by at most 4 lines

- express $\pi_{i}$ in that "base" of lines


X-slit camera

$$
\begin{aligned}
& \left(p^{0}, p^{1}, p^{2}, p^{3}\right)=\text { basis of the camera } \\
& \left(p^{0}, p^{1}, p^{2}\right)=\text { basis of the retina } \\
& \begin{cases}\zeta^{0}=p^{0} \vee p^{3} & \text { basis of } \\
\zeta^{1}=p^{1} \vee\left(p^{0}+p^{3}\right) & \text { the bag of } \\
\zeta^{2}=p^{2} \vee p^{3} & \text { lines } \\
\zeta^{3}=p^{1} \vee p^{2} & \text { and }\end{cases}
\end{aligned}
$$

## Normalized coordinates

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X-slit camera

$$
\begin{aligned}
& \left(p^{0}, p^{1}, p^{2}, p^{3}\right)=\text { basis of the camera } \\
& \left(p^{0}, p^{1}, p^{2}\right)=\text { basis of the retina } \\
& \begin{cases}\zeta^{0}=p^{0} \vee p^{3} & \text { basis of } \\
\zeta^{1}=p^{1} \vee\left(p^{0}+p^{3}\right) & \text { the bag of } \\
\zeta^{2}=p^{2} \vee p^{3} & \text { lines }\end{cases}
\end{aligned}
$$

$$
A_{\mathrm{s}}=\frac{1}{2}\left(\begin{array}{cccc}
1 & 0 & 0 & 0 \\
0 & -1 & 0 & 0 \\
0 & 0 & 1 & 0 \\
2 & 0 & 0 & -1
\end{array}\right) \quad \widetilde{P}_{\mathrm{s}} \times \mu_{\mathrm{s}}(u)=\left(\begin{array}{cccc}
0 & -1 & 0 & 0 \\
0 & 0 & 0 & 0 \\
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & -1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{array}\right) \times\left(\begin{array}{c}
u_{0}{ }^{2} \\
u_{0} u_{1} \\
u_{0} u_{2} \\
u_{1} u_{2}
\end{array}\right)
$$

## Normalized coordinates



Pencil camera


$$
\widetilde{P}_{\mathrm{s}} \times \mu_{\mathrm{s}}(u)=\left(\begin{array}{cccc}
0 & 0 & 1 & 0 \\
0 & 0 & 0 & 0 \\
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & -1 & 0 & 0 \\
0 & 0 & 0 & -1
\end{array}\right) \times\left(\begin{array}{c}
u_{0}{ }^{2} \\
u_{0} u_{1} \\
u_{0} u_{2} \\
u_{2}{ }^{2}
\end{array}\right)
$$

$\widetilde{\mathcal{P}}_{\mathrm{s}}$ and $\mu_{\mathrm{s}}$ depend only on the type of the camera.

$$
\widetilde{P}_{\mathrm{s}} \times \mu_{\mathrm{s}}(u)=\left(\begin{array}{cccc}
0 & 0 & 1 & 0 \\
0 & -1 & 0 & 0 \\
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & -1 & 0 & 0 \\
0 & 0 & 0 & -1
\end{array}\right) \times\left(\begin{array}{c}
u_{0}{ }^{2} \\
u_{0} u_{1} \\
u_{0} u_{2} \\
u_{1}{ }^{2}+u_{2}{ }^{2}
\end{array}\right)
$$

Linear oblique camera

## Normalized coordinates

Also for pinhole cameras!


$$
\widetilde{P}_{\mathrm{s}} \times \mu_{\mathrm{s}}(u)=\left(\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
1 & 0 & 0 \\
0 & 0 & 1 \\
0 & 1 & 0 \\
0 & 0 & 0
\end{array}\right) \times\left(\begin{array}{l}
u_{0} \\
u_{1} \\
u_{2}
\end{array}\right)
$$

## Normalized coordinates

## Calibration

$\star$ position of the camera in the world $\rightarrow W$

$\star$ position of the captor lattice in the retina $\rightarrow K$
retina coordinates

$$
\text { in }\left(p^{0}, p^{1}, p^{2}\right)
$$

invertible
$3 \times 3$ matrix
physical / pixel coordinates

## Normalized coordinates

## Calibration

$\star$ position of the camera in the world $\rightarrow W$


* position of the captor lattice in the retina $\rightarrow K$

$$
\begin{aligned}
& \text { retina coordinates } \\
& \text { in }\left(p^{0}, p^{1}, p^{2}\right)
\end{aligned}
$$

$\xrightarrow{3 \times 3 \text { matrix } \rightarrow}$| physical / pixel |
| :---: |
| coordinates |

$$
\pi_{i}(u)=\left(\Lambda^{2} W\right) \widetilde{\mathcal{P}}_{\mathrm{s}} \times \mu_{\mathrm{s}}\left(K^{-1} u\right)
$$

$\Lambda^{2} W$ is a $6 \times 6$ matrix encoding the action of $W$ on lines.

## Essential matrix



Image coordinates $u$ and $u^{\prime}$ are in stereo correspondance


$$
\pi_{i}(u) \odot \pi_{i}\left(u^{\prime}\right)=0
$$

$$
\mu_{\mathrm{s}}\left(K^{-1} u\right)^{T} \quad \widetilde{\mathcal{P}}_{\mathrm{s}}^{T} \Lambda^{2} W^{T}\left(\Lambda^{2} W^{\prime}\right)^{*}\left(\widetilde{\mathcal{P}}_{\mathrm{s}}^{\prime}\right)^{*} \quad \mu_{\mathrm{s}}\left(K^{\prime-1} u^{\prime}\right)=0
$$

$4 \times 4$ essential matrix
15 correspondances needed

## Euclidean framework

Exactly the same machinery

Only 6 correspondances needed to build $\mathcal{E}$

Need orthonormal normalized basis $\left(p^{0}, p^{1}, p^{2}, p^{3}\right)$

Induce more involved intrinsic parameters

Linear
Congruences
Grassmannian Sections

Admissible Maps

$6 \times 6$ fundamental matrix (uncalibrated case) $4 \times 4$ essential matrix (calibrated case) between $A N Y$ pair of linear cameras

## FURTHER QUESTIONS



What are the positions of the retina that minimize the distorsions of the image?

Do algebraic line congruences suggest interesting imaging devices? Can they admit analogues of admissible maps?


[Pajdla et al.]


Can we get clear pictures from our pencil camera?

## Thank you!

[1] J. Ponce, What is a Camera?, CVPR'09
[2] G. B., X. Goaoc, J. Ponce, Admissible Linear Map Model for Linear Cameras, CVPR'10
[3] The "linear camera" model (in preparation)

## Pencil camera



## Synthesis

|  | projective <br> linear <br> camera | euclidean <br> linear <br> camera | euclidean <br> pinhole <br> camera |
| :---: | :---: | :---: | :---: |
| extrinsic parameters | 15 | 6 | 6 |
| intrinsic parameters | 6 | 12 | 5 |
| fundamental matrix | 27 |  | 7 |

## Linearity on lines

## Geometric axiomatisation <br> [Veblen\&Young,1910] for linear dependance

Initialisation

two lines

line pencil

regulus
$\ell$ depends linearly on $L=\left\{\ell_{1}, \ldots, \ell_{k}\right\}$ iff
Heredity $\exists L \rightarrow \ell_{k+1} \rightarrow \cdots \rightarrow \ell_{k+n}=\ell$
where $\rightarrow$ is a line pencil/regulus construction.

Exemples

degenerate regulus

bundle
(Need 2 line pencil steps)

