
5. Segmentation

9. Stitching
12. 3D Shape


3. Image Processing


6-7. Structure from Motion

10. Computational Photography

4. Features

8. Motion

11. Stereo

13. Image-based Rendering

14. Recognition

## Image Formation

2.1 Geometric primitives and transformations ..... 31
2.1.1 Geometric primitives ..... 32
2.1.2 2D transformations ..... 35
2.1.3 3D transformations ..... 39
2.1.4 3D rotations ..... 41
2.1.5 3D to 2D projections ..... 46
2.1.6 Lens distortions ..... 58
2.2 Photometric image formation ..... 60
2.2.1 Lighting ..... 60
2.2.2 Reflectance and shading ..... 62
2.2.3 Optics ..... 68
2.3 The digital camera ..... 73
2.3.1 Sampling and aliasing ..... 77
2.3.2 Color ..... 80
2.3.3 Compression ..... 90
2.4 Additional reading ..... 93
2.5 Exercises ..... 93

## Image Formation

2.1 Geometric primitives and transformations ..... 31
2.1.1 Geometric primitives ..... 32
2.1.2 2D transformations ..... 35
2.1.3 3D transformations ..... 39
2.1.4 3D rotations ..... 41
2.1.5 3D to 2 D projections ..... 46
2.1.6 Lens distortions ..... 58
2.2 Photometric image formation ..... 60
2.2.1 Lighting ..... 60
2.2.2 Reflectance and shading ..... 62
2.2.3 Optics ..... 68
2.3 The digital camera ..... 73
2.3.1 Sampling and aliasing ..... 77
2.3.2 Color ..... 80
2.3.3 Compression ..... 90
2.4 Additional reading ..... 93
2.5 Exercises ..... 93

### 2.1.1 Geometric Primitives

- 2D points:
- 2D lines:
- 2D conics:
- 3D points:
- 3D planes:
- 3D lines:


## 2D Coordinate Frames \& Points

- coordinates $x$ and $y$


2D Lines

- Line I = ax+by=c



## Homogeneous Coordinates

- Uniform treatment of points and lines
- Line-point incidence: $\left.\right|^{\top} p=0$



## Join = cross product !

- Join of two lines is a point: $\mathrm{p}=\mathrm{l}_{1} \times \mathrm{l}_{2}$

- Join of two points is a line:
$\mathrm{l}=\mathrm{p}_{1} \times \mathrm{p}_{2}$


Automatic estimation of vanishing points and lines


## Joining two parallel lines?

(a,b,c)
$p=\left|\begin{array}{lll}i & j & k \\ a & b & c \\ a & b & d\end{array}\right|=\left[\begin{array}{c}b d-c b \\ c a-a d \\ 0\end{array}\right]$
(a,b,d)

## Points at Infinity !



## Homogeneous coordinates

## Conversion

Converting to homogeneous coordinates

$$
\begin{array}{cc}
(x, y) \Rightarrow\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] & (x, y, z) \Rightarrow\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right] \\
\begin{array}{cc}
\text { homogeneous image } & \text { homogeneous scene } \\
\text { coordinates } & \text { coordinates }
\end{array}
\end{array}
$$

Converting from homogeneous coordinates

$$
\left[\begin{array}{c}
x \\
y \\
w
\end{array}\right] \Rightarrow(x / w, y / w) \quad\left[\begin{array}{l}
x \\
y \\
z \\
w
\end{array}\right] \Rightarrow(x / w, y / w, z / w)
$$

### 2.1.1 Geometric Primitives

homogeneous augmented

- 2D points: $(x, y), \tilde{\boldsymbol{x}}=(\tilde{x}, \tilde{y}, \tilde{w})=\tilde{w}(x, y, 1)=\tilde{w} \overline{\boldsymbol{x}}$
- 2D lines: $\overline{\boldsymbol{x}} \cdot \tilde{\boldsymbol{l}}=a x+b y+c=0$
- 2D conics:
- 3D points:
- 3D planes:
- 3D lines:


### 2.1.1 Geometric Primitives

- 2D points: $(x, y), \tilde{\boldsymbol{x}}=(\tilde{x}, \tilde{y}, \tilde{w})=\tilde{w}(x, y, 1)=\tilde{w} \overline{\boldsymbol{x}}$
- 2D lines: $\overline{\boldsymbol{x}} \cdot \tilde{\boldsymbol{l}}=a x+b y+c=0$
- 2D conics:
- 3D points: $\boldsymbol{x}=(x, y, z) \tilde{\boldsymbol{x}}=(\tilde{x}, \tilde{y}, \tilde{z}, \tilde{w})$
- 3D planes: $\overline{\boldsymbol{x}} \cdot \tilde{\boldsymbol{m}}=a x+b y+c z+d=0$
- 3D lines:


### 2.1.1 Geometric Primitives

- 2D points: $(x, y), \tilde{\boldsymbol{x}}=(\tilde{x}, \tilde{y}, \tilde{w})=\tilde{w}(x, y, 1)=\tilde{w} \overline{\boldsymbol{x}}$
- 2D lines: $\overline{\boldsymbol{x}} \cdot \tilde{\boldsymbol{l}}=a x+b y+c=0$
- 2D conics: $\tilde{x}^{T} Q \tilde{x}=0$
- 3D points: $\boldsymbol{x}=(x, y, z) \tilde{\boldsymbol{x}}=(\tilde{x}, \tilde{y}, \tilde{z}, \tilde{w})$
- 3D planes: $\overline{\boldsymbol{x}} \cdot \tilde{\boldsymbol{m}}=a x+b y+c z+d=0$
- 3D lines: $r=(1-\lambda) p+\lambda q$

$$
\begin{aligned}
\tilde{\boldsymbol{r}} & =\mu \tilde{\boldsymbol{p}}+\lambda \tilde{\boldsymbol{q}} \\
\boldsymbol{r} & =\boldsymbol{p}+\lambda \hat{\boldsymbol{d}}
\end{aligned}
$$

### 2.1.2: 2D Transformations



### 2.1.2: 2D Transformations


translation

affine

rotation

perspective

aspect

cylindrical

## 2D planar transformations



## 2D planar transformations



How would you implement scaling?

- Each component multiplied by a scalar
- Uniform scaling - same scalar for each component


## 2D planar transformations

$$
\begin{gathered}
x^{\prime}=a x \\
y^{\prime}=b y
\end{gathered}
$$



What's the effect of using different scale factors?

- Each component multiplied by a scalar
- Uniform scaling - same scalar for each component


## 2D planar transformations



$$
\begin{aligned}
x^{\prime} & =a x \\
y^{\prime} & =b y
\end{aligned}
$$

matrix representation of scaling:

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\underbrace{\left[\begin{array}{ll}
a & 0 \\
0 & b
\end{array}\right]}_{\text {scaling matrix S }}\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

- Each component multiplied by a scalar
- Uniform scaling - same scalar for each component


## 2D planar transformations

$y$


## 2D planar transformations



## 2D planar transformations

$y$

## 2D planar and linear transformations

| Scale | Flip across y |
| :---: | :---: |
| $\mathbf{M}=\left[\begin{array}{cc}s_{x} & 0 \\ 0 & s_{y}\end{array}\right]$ | $\mathbf{M}=\left[\begin{array}{cc} -1 & 0 \\ 0 & 1 \end{array}\right]$ |
| Rotate | Flip across origin |
| $\mathbf{M}=\left[\begin{array}{cc}\cos \theta & -\sin \theta \\ \sin \theta & \cos \theta\end{array}\right]$ | $\mathbf{M}=\left[\begin{array}{cc} -1 & 0 \\ 0 & -1 \end{array}\right]$ |
| Shear | Identity |
| $\mathbf{M}=\left[\begin{array}{cc}1 & s_{x} \\ s_{y} & 1\end{array}\right]$ | $\mathbf{M}=\left[\begin{array}{ll} 1 & 0 \\ 0 & 1 \end{array}\right]$ |

## 2D translation



## 2D translation



## 2D translation



## 2D translation



## 2D translation



## 2D translation using homogeneous coordinates

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{llc}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]=\left[\begin{array}{c}
x+t_{x} \\
y+t_{y} \\
1
\end{array}\right]
$$



## 2D Transformations in homogeneous coordinates

## Reminder: Homogeneous coordinates

Conversion:
Special points:

- inhomogeneous $\rightarrow$
augmented/homogeneous

$$
\left[\begin{array}{l}
x \\
y
\end{array}\right] \Rightarrow\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

- homogeneous $\rightarrow$ inhomogeneous

$$
\left[\begin{array}{c}
x \\
y \\
w
\end{array}\right] \Rightarrow\left[\begin{array}{l}
x / w \\
y / w
\end{array}\right]
$$

- point at infinity
- undefined

$$
\left[\begin{array}{lll}
x & y & 0
\end{array}\right]
$$

- scale invariance

$$
\left[\begin{array}{lll}
x & y & w
\end{array}\right]^{\top}=\lambda\left[\begin{array}{lll}
x & y & w
\end{array}\right]^{\top}
$$

## 2D transformations

Re-write these transformations as $3 \times 3$ matrices:

$$
\begin{aligned}
& \begin{array}{c}
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=} \\
\text { translation }
\end{array} \\
& {\left[\begin{array}{c}
x^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]=\underset{\text { scaling }}{\left[\begin{array}{l}
?
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]}} \\
& {\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=[ } \\
& ]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] \\
& \text { rotation } \\
& {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{l}
?
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]} \\
& \text { shearing }
\end{aligned}
$$

## 2D transformations

Re-write these transformations as $3 \times 3$ matrices:

$$
\begin{aligned}
& \begin{array}{c}
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=} \\
\text { translation }
\end{array}
\end{aligned}
$$

$$
\begin{aligned}
& {\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=[ } \\
& ]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] \\
& \text { rotation } \\
& {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{l}
?
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]} \\
& \text { shearing }
\end{aligned}
$$

## 2D transformations

Re-write these transformations as $3 \times 3$ matrices:

$$
\begin{aligned}
& \begin{aligned}
& {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right] }=\left[\begin{array}{lll}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] \\
& \text { translation }
\end{aligned}
\end{aligned}
$$

$$
\begin{aligned}
& {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=[ } \\
& \text { ? } \quad]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] \\
& \text { rotation } \\
& {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
1 & \beta_{x} & 0 \\
\beta_{y} & 1 & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]} \\
& \text { shearing }
\end{aligned}
$$

## 2D transformations

Re-write these transformations as $3 \times 3$ matrices:

$$
\begin{array}{cc}
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=} & {\left[\begin{array}{lll}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]} \\
\text { translation } & {\left[\begin{array}{c}
\boldsymbol{x}^{\prime} \\
\boldsymbol{y}^{\prime} \\
1
\end{array}\right]=} \\
{\left[\begin{array}{c}
{\left[\begin{array}{ccc}
\boldsymbol{s}_{\boldsymbol{x}} & 0 & 0 \\
0 & \boldsymbol{s}_{\boldsymbol{y}} & 0 \\
0 & 0 & 1
\end{array}\right]}
\end{array}\left[\begin{array}{l}
\boldsymbol{x} \\
\boldsymbol{y} \\
1
\end{array}\right]\right.} \\
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\underset{\text { scaling }}{\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]}\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\underset{\text { shearing }}{\left[\begin{array}{ccc}
1 & \beta_{x} & 0 \\
\boldsymbol{\beta}_{y} & 1 & 0 \\
0 & 0 & 1
\end{array}\right]}\left[\begin{array}{c}
x \\
y \\
1
\end{array}\right]}
\end{array}
$$

## Matrix composition

Transformations can be combined by matrix multiplication:

$$
\begin{aligned}
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right] } & =\left(\left[\begin{array}{lll}
1 & 0 & t x \\
0 & 1 & t y \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
s x & 0 & 0 \\
0 & s y & 0 \\
0 & 0 & 1
\end{array}\right]\right)\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right] \\
\mathbf{p}^{\prime} & =? ?
\end{aligned}
$$

## Matrix composition

Transformations can be combined by matrix multiplication:

$$
\begin{aligned}
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right] } & =\left(\left[\begin{array}{lll}
1 & 0 & t x \\
0 & 1 & t y \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
\cos \Theta & -\sin \Theta & 0 \\
\sin \Theta & \cos \Theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
s x & 0 & 0 \\
0 & s y & 0 \\
0 & 0 & 1
\end{array}\right]\right)\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right] \\
\mathbf{p}^{\prime} & =\operatorname{translation}\left(\mathrm{t}_{\left.\mathrm{x}, \mathrm{t}_{\mathrm{y}}\right)} \quad \operatorname{rotation}(\theta)\right.
\end{aligned}
$$

## Classification of 2D transformations

## Classification of 2D transformations



## Classification of 2D transformations

| Name | Matrix | \# D.O.F. |
| :--- | :---: | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]$ | $?$ |
| rigid (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]$ | $?$ |
| similarity | $[s \boldsymbol{R} \mid \boldsymbol{t}]$ | $?$ |
| affine | $[\boldsymbol{A}]$ | $?$ |
| projective | $[\tilde{\boldsymbol{H}}]$ | $?$ |

## Translation

Translation: $\left[\begin{array}{ccc}1 & 0 & t_{1} \\ 0 & 1 & t_{2} \\ 0 & 0 & 1\end{array}\right]$

How many degrees of freedom?


## Euclidean/Rigid

$\underset{\text { Euclidean (rigid): }}{\text { rotation + translation }} \quad\left[\begin{array}{ccc}\cos \theta & -\sin \theta & r_{3} \\ \sin \theta & \cos \theta & r_{6} \\ 0 & 0 & 1\end{array}\right]$

How many degrees of freedom?


## Affine



Are there any values that are related?


## Affine transformations

Affine transformations are combinations of

- arbitrary (4-DOF) linear transformations
-     + translations

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right]=\left[\begin{array}{lll}
a & b & c \\
d & e & f \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right]
$$

Properties of affine transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines map to parallel lines
- ratios are preserved
- compositions of affine transforms are also

Does the last coordinate w ever change?

## Projective transformations

Projective transformations are combinations of

- affine transformations;
-     + projective wraps

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right]=\left[\begin{array}{lll}
a & b & c \\
d & e & f \\
g & h & i
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right]
$$

8 DOF: vectors (and therefore matrices) are defined up to scale)
Properties of projective transformations:

- origin does not necessarily map to origin
- lines map to lines
- parallel lines do not necessarily map to parallel lines
- ratios are not necessarily preserved
- compositions of projective transforms are also projective transforms


## Classification of 2D transformations

| Name | Matrix | \# D.O.F. |
| :--- | :---: | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]$ | $?$ |
| rigid (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]$ | $?$ |
| similarity | $[s \boldsymbol{R} \mid \boldsymbol{t}]$ | $?$ |
| affine | $[\boldsymbol{A}]$ | $?$ |
| projective | $[\tilde{\boldsymbol{H}}]$ | $?$ |

## Classification of 2D transformations

| Name | Matrix | \# D.O.F. |
| :--- | :---: | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]$ | 2 |
| rigid (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]$ | 3 |
| similarity | $[s \boldsymbol{R} \mid \boldsymbol{t}]$ | 4 |
| affine | $[\boldsymbol{A}]$ | 6 |
| projective | $[\tilde{\boldsymbol{H}}]$ | 8 |

### 2.1.3: 3D Transformations

- Need a way to specify the six degrees-of-freedom of a rigid body.
- Why are their 6 DOF?


A rigid body is a collection of points whose positions relative to each other can't change


Fix one point, three DOF


Fix second point, two more DOF (must maintain distance constraint)


Third point adds one more DOF, for rotation around line

## Notations

- Superscript references coordinate frame
- $\quad A P$ is coordinates of $P$ in frame $A$
- $\quad{ }^{B} P$ is coordinates of $P$ in frame $B$
- Example :



## Translation

- Using augmented/homogeneous coordinates, translation isexpressed as a matrix multiplication.
${ }^{B} P={ }^{A} P+{ }^{B} O_{A}$
$\left[\begin{array}{l}{ }^{B} P \\ 1\end{array}\right]=\left[\begin{array}{cc}I & { }^{B} O_{A} \\ 0 & 1\end{array}\right]\left[\begin{array}{l}{ }^{A} P \\ 1\end{array}\right]$
- Translation is communicative


## Rotation in homogeneous coordinates

- Using homogeneous coordinates, rotation can be expressed as a matrix multiplication.

$$
\begin{aligned}
{ }^{B} P & ={ }_{A}^{B} R^{A} P \\
{\left[\begin{array}{l}
B \\
1
\end{array}\right] } & =\left[\begin{array}{cc}
{ }_{A}^{B} R & 0 \\
0 & 1
\end{array}\right]\left[\begin{array}{l}
{ }^{A} P \\
1
\end{array}\right]
\end{aligned}
$$

- R is a rotation matrix:
- Columns are unit vectors
- Columns are mutually orthogonal
- Inverse is transpose
- Rotation is not communicative


## 3D Rigid transformations



$$
{ }^{B} P={ }_{A}^{B} R^{A} P+{ }^{B} O_{A}
$$

## 3D Rigid transformations

- Unified treatment using homogeneous coordinates.

$$
\begin{aligned}
{\left[\begin{array}{l}
{ }^{B} P \\
1
\end{array}\right] } & =\left[\begin{array}{cc}
1 & { }^{B} O_{A} \\
0 & 1
\end{array}\right]\left[\begin{array}{cc}
{ }_{A}^{B} R & 0 \\
0 & 1
\end{array}\right]\left[\begin{array}{l}
{ }^{A} P \\
1
\end{array}\right] \\
& =\left[\begin{array}{cc}
{ }_{A}^{B} R & { }^{B} O_{A} \\
0^{T} & 1
\end{array}\right]\left[\begin{array}{l}
{ }^{A} P \\
1
\end{array}\right] \\
& {\left[\begin{array}{c}
{ }^{B} P \\
1
\end{array}\right]={ }_{A}^{B} T\left[\begin{array}{c}
{ }^{A} P \\
1
\end{array}\right] }
\end{aligned}
$$

## Hierarchy of 3D <br> Transforms

- Subgroup Structure:
- Translation (? DOF)
- Rigid 3D (? DOF)
- Affine (? DOF)
- Projective (? DOF)

Hierarchy of 3D Transforms

- Subgroup Structure:
- Translation (3 DOF)
- Rigid 3D (6 DOF)
- Affine (12 DOF)
- Projective (15 DOF)


### 2.1.5: 3D to 2D: Projection

3D world
2D image


Point of observation

## Orthographic Projection



## Pinhole camera



## Camera obscura: the pre-camera

- Known during classical period in China and Greece (e.g. Mo-Ti, China, 470BC to 390BC)


Illustration of Camera Obscura


Freestanding camera obscura at UNC Chapel Hill

## Camera Obscura used for Tracing



Lens Based Camera Obscura, 1568

## First Photograph

Oldest surviving photograph

- Took 8 hours on pewter plate


Joseph Niepce, 1826

Photograph of the first photograph


Stored at UT Austin

Niepce later teamed up with Daguerre, who eventually created Daguerrotypes

## Projection can be tricky...



## Projection can be tricky...







## Camera and World Geometry



## Pinhole Camera

- Fundamental equation:



## Homogeneous Coordinates

Linear transformation of homogeneous (projective) coordinates
$m=\left[\begin{array}{l}u \\ v \\ w\end{array}\right]=\left[\begin{array}{lll}I & 0\end{array}\right] M=\left[\begin{array}{llll}1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0\end{array}\right]\left[\begin{array}{c}X \\ Y \\ Z \\ T\end{array}\right]$
Recover image (Euclidean) coordinates by normalizing:
$\hat{u}=\frac{u}{w}=\frac{X}{Z}$
$\hat{v}=\frac{v}{w}=\frac{Y}{Z}$

## We can see infinity!

Railroad: parallel lines


## Vanishing points and lines



## Vanishing points and lines



## Pixel coordinates in 2D

$(0.5,0.5) \quad 640$


## Intrinsic Calibration

$3 \times 3$ Calibration Matrix K

Recover image (Euclidean) coordinates by normalizing:
$\hat{u}=\frac{u}{w}=\frac{\alpha X+s Y+u_{0}}{Z}$
$\hat{v}=\frac{v}{w}=\frac{\beta Y+v_{0}}{Z}$

## Camera Pose

In order to apply the camera model, objects in the scene must be expressed in camera coordinates.


## Projective Camera Matrix

Camera $=$ Calibration $\times$ Projection $\times$ Extrinsics

$$
\begin{aligned}
& m=\left[\begin{array}{l}
u \\
v \\
w
\end{array}\right]=\left[\begin{array}{lll}
\alpha & s & u_{0} \\
& \beta & v_{0} \\
& & 1
\end{array}\right]\left[\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right]\left[\begin{array}{ll}
R & t \\
0 & 1
\end{array}\right]\left[\begin{array}{l}
X \\
Y \\
Z \\
T
\end{array}\right] \\
& =K\left[\begin{array}{ll}
R & t
\end{array}\right] M=P M
\end{aligned}
$$

## Projective Geometry

## What is lost?

- Length



## Length and area are not preserved



Figure by David Forsyth

## Projective Geometry

## What is lost?

- Length
- Angles



## Projective Geometry

## What is preserved?

- Straight lines are still straight


Field of View (Zoom, focal length)



From London and Upton

### 2.1.6 Radial Distortion




Corrected Barrel Distortion

