# Geometric Model of Camera 

Dr. Gerhard Roth

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Version 1

## Similar Triangles


property (i): corresponding angles are equal

$$
\left(A=A^{\prime} \text { and } B=B^{\prime} \text { and } C=C^{\prime}\right)
$$

property (ii): corresponding sides have proportional lengths

$$
\left(\frac{a}{a^{\prime}}=\frac{b}{b^{\prime}}=\frac{c}{c^{\prime}}\right)
$$

## Geometric Model of Camera

Perspective projection


$$
\begin{aligned}
& \mathrm{P}(\mathrm{X}, \mathrm{Y}, \mathrm{Z}) \rightarrow p(\mathrm{x}, \mathrm{y}) \\
& x=f \frac{X}{Z} \quad y=f \frac{Y}{Z}
\end{aligned}
$$

## Parallel lines aren't...



Figure by David Forsyth

## Lengths can't be trusted...



## Coordinate Transformation - 2D

$$
\begin{aligned}
& p^{\prime}=\left[\begin{array}{l}
p_{x}{ }^{\prime} \\
p_{y}^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
\cos \phi & \sin \phi \\
-\sin \phi & \cos \phi
\end{array}\right]\left[\begin{array}{c}
p_{x} \\
p_{y}
\end{array}\right] \\
& p^{\prime \prime}=\left[\begin{array}{c}
p_{x}{ }^{\prime} \\
p_{y}{ }^{\prime}
\end{array}\right]+\left[\begin{array}{c}
T_{x} \\
T_{y}
\end{array}\right]=\left[\begin{array}{cc}
\cos \phi & \sin \phi \\
-\sin \phi & \cos \phi
\end{array}\right]\left[\begin{array}{c}
p_{x} \\
p_{y}
\end{array}\right]+\left[\begin{array}{c}
T_{x} \\
T_{y}
\end{array}\right]
\end{aligned}
$$

## Homogeneous Coordinates

Go one dimensional higher:

$$
\left[\begin{array}{l}
x \\
y
\end{array}\right] \rightarrow\left[\begin{array}{c}
w x \\
w y \\
w
\end{array}\right] \quad\left[\begin{array}{l}
x \\
y \\
z
\end{array}\right] \rightarrow\left[\begin{array}{c}
w x \\
w y \\
w z \\
w
\end{array}\right]
$$

$w$ is an arbitrary non-zero scalar, usually we choose 1.

From homogeneous coordinates to Cartesian coordinates:

$$
\left[\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right] \rightarrow\left[\begin{array}{l}
x_{1} / x_{3} \\
x_{2} / x_{3}
\end{array}\right] \quad\left[\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3} \\
x_{4}
\end{array}\right] \rightarrow\left[\begin{array}{l}
x_{1} / x_{4} \\
x_{2} / x_{4} \\
x_{3} / x_{4}
\end{array}\right]
$$

## 2D Transformation with Homogeneous Coordinates

2D coordinate transformation:

$$
p^{\prime \prime}=\left[\begin{array}{cc}
\cos \phi & \sin \phi \\
-\sin \phi & \cos \phi
\end{array}\right]\left[\begin{array}{l}
p_{x} \\
p_{y}
\end{array}\right]+\left[\begin{array}{l}
T_{x} \\
T_{y}
\end{array}\right]
$$

2D coordinate transformation using homogeneous coordinates:

$$
\left[\begin{array}{c}
p_{x}^{\prime \prime} \\
p_{y}^{\prime \prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
\cos \phi & \sin \phi & T_{x} \\
-\sin \phi & \cos \phi & T_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
p_{x} \\
p_{y} \\
1
\end{array}\right]
$$

## Homogeneous coordinates (In 2d)

Two points are equal if and only if:

$$
x^{\prime} / w^{\prime}=x / w \quad \text { and } \quad y^{\prime} / w^{\prime}=y / w
$$

$w=0$ : points at infinity

- useful for projections and curve drawing

Homogenize = divide by $w$. Homogenized points:

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

## Translations with homogeneous

$$
\begin{aligned}
& {\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right]=\left[\begin{array}{lll}
1 & 0 & t_{x} \\
0 & 1 & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
x \\
y \\
w
\end{array}\right]} \\
& \left\{\begin{array}{l}
\frac{x^{\prime}}{w^{\prime}}=\frac{x}{w}+t_{x} \\
\frac{y^{\prime}}{w^{\prime}}=\frac{y}{w}+t_{y}
\end{array}\right. \\
& \longrightarrow\left\{\begin{array}{l}
x^{\prime}=x+w t_{x} \\
y^{\prime}=y+w t_{y} \\
w^{\prime}=w
\end{array}\right.
\end{aligned}
$$

## Scaling with homogeneous

$$
\begin{aligned}
{\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right]=} & {\left[\begin{array}{lll}
s_{x} & 0 & 0 \\
0 & s_{y} & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right] }
\end{aligned} \underbrace{}_{\left\{\begin{array}{l}
\frac{x^{\prime}}{w^{\prime}}=S_{x} \frac{x}{w} \\
\frac{y^{\prime}}{w^{\prime}}=s_{x} x \\
y^{\prime}=s_{y} y \\
w^{\prime}=w
\end{array}\right.} \underbrace{\frac{y}{w}}
$$

## Rotation with homogeneous

$$
\begin{gathered}
{\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
w^{\prime}
\end{array}\right]=}
\end{gathered} \underbrace{\qquad\left\{\begin{array}{l}
x^{\prime}=\cos \theta x-\sin \theta y \\
y^{\prime}=\sin \theta x+\cos \theta y \\
w^{\prime}=
\end{array} w\right.}_{\left[\begin{array}{ccc}
\cos \theta & -\sin \theta & 0 \\
\sin \theta & \cos \theta & 0 \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
w
\end{array}\right] \quad\left\{\begin{array}{l}
\frac{x^{\prime}}{w^{\prime}}=\cos \theta \frac{x}{w}-\sin \theta \frac{y}{w} \\
\frac{y^{\prime}}{w^{\prime}}=\sin \theta \frac{x}{w}+\cos \theta \frac{y}{w}
\end{array}\right.} \begin{aligned}
& w
\end{aligned}
$$

## 3D Rotation Matrix

Rotate around each coordinate axis:
$R_{1}(\alpha)=\left[\begin{array}{ccc}1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha\end{array}\right] \quad R_{2}(\beta)=\left[\begin{array}{ccc}\cos \beta & 0 & \sin \beta \\ 0 & 1 & 0 \\ -\sin \beta & 0 & \cos \beta\end{array}\right] \quad R_{3}(\gamma)=\left[\begin{array}{ccc}\cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1\end{array}\right]$

Combine the three rotations:

$$
R=R_{1} R_{2} R_{3}
$$

3 D rotation matrix has three parameters, no matter how it is specified.

## Rotation Matrices

- Both 2d and 3d rotation matrices have two characteristics
- They are orthogonal (also called orthonormal)

$$
R^{T} R=I \quad R^{T}=R^{-1}
$$

- Their determinant is 1
- Matrix below is orthogonal but not a rotation matrix because the determinate is not 1

$$
\left(\begin{array}{ccc}
1 & 0 & 0 \\
0 & -1 & 0 \\
0 & 0 & 1
\end{array}\right)
$$

$\longleftarrow$ this is a reflection matrix

## Homogeneous co-ordinates

- Transformation - transform a point in an $n$ dimensional space to another $n$ dim point
- Transformations are scale, rotations, translations, etc.
- You can represent all these by multiplication by one appropriate matrix using homogeneous co-ordinates
- Projection - transform a point in an n dimensional space to an m dim point
- For projection $m$ is normally less than $n$
- Perspective projection is a projection (3d to 2d)
- From the 3d world to a 2d point in the image
- You can also represent a projection as matrix multiplication with one appropriate matrix and homogeneouse co-ordinates


## Four Coordinate Frames



Camera model: $\quad p_{i m}=\left[\begin{array}{c}\text { transformation } \\ \text { matrix }\end{array}\right] P_{w}$

## Perspective Projection



$$
x=f \frac{X}{Z} \quad y=f \frac{Y}{Z}
$$

These are nonlinear.

Using homogenous coordinate, we have a linear relation:

$$
\begin{gathered}
{\left[\begin{array}{c}
u \\
v \\
w
\end{array}\right]=\left[\begin{array}{llll}
f & 0 & 0 & 0 \\
0 & f & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right]\left[\begin{array}{l}
X \\
Y \\
Z \\
1
\end{array}\right]} \\
x=u / w \quad y=v / w
\end{gathered}
$$

## World to Camera Coordinate

Transformation between the camera and world coordinates. Here we rotate, then translate, to go from world to camera co-ordinates which is opposite of book, but is simpler and is the way in which OpenCV routines do it:


$$
\begin{gathered}
\mathbf{X}_{c}=\mathbf{R} \mathbf{X}_{w}+\mathbf{T} \\
{\left[\begin{array}{c}
X_{c} \\
Y_{c} \\
Z_{c} \\
1
\end{array}\right]=\left[\begin{array}{cc}
\mathbf{R} & \mathbf{T} \\
\mathbf{0}^{\mathrm{T}} & 1
\end{array}\right]\left[\begin{array}{c}
X_{w} \\
Y_{w} \\
Z_{w} \\
1
\end{array}\right]}
\end{gathered}
$$

After R, and T we have converted from world to camera frame. In the camera frame the z axis is along the optical center.

## Camera Coordinates to Pixel Coordinates

$$
x=\left(o_{x}-x_{i m}\right) s_{x} \quad y=\left(o_{y}-y_{i m}\right) s_{y}
$$

$s_{x}, s_{y}:$ pixel sizes in millimeters per pixel


Camera co-ordinates $x$, and $y$ are in millimetres
Image co-ordinates $x_{i m}, y_{i m}$, are in pixels
Center of projection $\mathrm{o}_{\mathrm{x}}, \mathrm{o}_{\mathrm{y}}$ is in pixels
Sign change because horizontal and vertical axis of the image and camera frame have opposite directions.

## Image and Camera frames

Now we look from the camera outward and image origin is the top left pixel $(0,0)$


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## Put All Together - World to Pixel

$$
\begin{aligned}
& {\left[\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right]=\left[\begin{array}{ccc}
-1 / s_{x} & 0 & o_{x} \\
0 & -1 / s_{y} & o_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{c}
u \\
v \\
w
\end{array}\right]} \\
& =\left[\begin{array}{ccc}
-1 / s_{x} & 0 & o_{x} \\
0 & -1 / s_{y} & o_{y}
\end{array}\right]\left[\begin{array}{cccc}
f & 0 & 0 & 0 \\
0 & f & 0 & 0
\end{array}\right]\left[\begin{array}{c}
X_{c} \\
Y_{c} \\
Z_{c}
\end{array}\right] \quad \text { Add projection } \\
& =\left[\begin{array}{ccc}
-1 / s_{x} & 0 & o_{x} \\
0 & -1 / s_{y} & o_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{cccc}
f & 0 & 0 & 0 \\
0 & f & 0 & 0 \\
0 & 0 & 1 & 0
\end{array}\right]\left[\begin{array}{cc}
R & T \\
0 & 1
\end{array}\right]\left[\begin{array}{c}
X_{w} \\
Y_{w} \\
Z_{w} \\
1
\end{array}\right] \\
& \text { Add Rotation } \\
& \text { And Translation } \\
& x_{i m}=x_{1} / x_{3} \quad y_{i m}=x_{2} / x_{3}
\end{aligned}
$$

## Camera Parameters

- Extrinsic parameters define the location and orientation of the camera reference frame with respect to a world reference frame
- Depend on the external world, so they are extrinsic
- Intrinsic parameters link pixel co-ordinates in the image with the corresponding coordinates in the camera reference frame
- An intrinsic characteristic of the camera
- Image co-ordinates are in pixels
- Camera co-ordinates are in millimetres
- In formulas that do conversions the units must match!


## Intrinsic Camera Parameters

$$
K=\left[\begin{array}{ccc}
-f / s_{x} & 0 & o_{x} \\
0 & -f / s_{y} & o_{y} \\
0 & 0 & 1
\end{array}\right]
$$

K is a $3 \times 3$ upper triangular matrix, called the Camera Calibration Matrix.

There are five intrinsic parameters:
(a) The pixel sizes in x and y directions $s_{x}, s_{y}$ in millimeters/pixel
(b) The focal length $f$ in millimeters
(c) The principal point ( $\mathrm{ox}, \mathrm{oy}$ ) in pixels, which is the point where the optic axis intersects the image plane.
(d) The units of $f / S x$ and $f / S y$ are in pixels, why is this so?

## Camera intrinsic parameters

- Can write three of these parameters differently by letting $\mathrm{f} / \mathrm{sx}=\mathrm{fx}$ and $\mathrm{f} / \mathrm{sy}=\mathrm{fy}$
- Then intrinsic parameters are ox,oy,fx,fy
- The units of these parameters are pixels!
- In practice pixels are square ( $s x=s y$ ) so that means fx should equal fy for most cameras
- However, every explicit camera calibration process (using calibration objects) introduces some small errors
- These calibration errors make fx not exactly equal to fy
- So in OpenCV the intrinsic camera parameters are the four following ox,oy,fx,fy
- However fx is usually very close to fy and if this is not the case then there is a problem


## Extrinsic Parameters and Proj. Matrix

$$
p_{i m}=\left[\begin{array}{c}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right]=K\left[\begin{array}{ll}
R & T
\end{array}\right]\left[\begin{array}{c}
X_{w} \\
Y_{w} \\
Z_{w} \\
1
\end{array}\right]=M\left[\begin{array}{c}
X_{w} \\
Y_{w} \\
Z_{w} \\
1
\end{array}\right]
$$

$[R \mid T]$ defines the extrinsic parameters.
The $3 \times 4$ matrix $M=K[R \mid T]$ is called the projection matrix.
It takes 3d points in the world co-ordinate system and maps them to the appropriate image co-ordinates in pixels

## Using the projection matrix - example

$$
\begin{aligned}
& {\left[\begin{array}{llll}
\mathbf{P}_{11} & \mathbf{P}_{12} & \mathbf{P}_{13} & \mathbf{P}_{14} \\
\mathbf{P}_{21} & \mathbf{P}_{22} & \mathbf{P}_{23} & \mathbf{P}_{24} \\
\mathbf{P}_{31} & \mathbf{P}_{32} & \mathbf{P}_{33} & \mathbf{P}_{34}
\end{array}\right]=\left[\begin{array}{cccc}
1000 & 0 & 320 & -19600 \\
0 & 1000 & 240 & -27200 \\
0 & 0 & 1 & -30
\end{array}\right]} \\
& \quad u=\frac{u^{\prime}}{w^{\prime}} \\
& v=\frac{v^{\prime}}{w^{\prime}}
\end{aligned}
$$

- Where would the point $(20,50,200)$ project to in the image?

$$
\begin{aligned}
& u=u^{\prime} / w^{\prime}=64400 / 170=378.8 \\
& {\left[\begin{array}{l}
u^{\prime} \\
v^{\prime} \\
w^{\prime}
\end{array}\right]=\left[\begin{array}{r}
64400 \\
70800 \\
170
\end{array}\right]} \\
& v=v^{\prime} / w^{\prime}=70800 / 170=416.5 \\
& \text { - World point }(20,50,200) \text { project to pixel } \\
& \text { With co-ordinates of }(379,414)
\end{aligned}
$$

## Effect of change in focal length

Small f is wide angle, large f is telescopic

perspective
perspective


weak perspective


## Orthographic Projection

Orthographic Projection


## Orthographic Projection

Orthographic Projection

## Weak Perspective Model

Assume the relative distance between any two points in an object along the principal axis is much smaller ( $1 / 20^{\text {th }}$ at most) than the $\bar{Z}$ average distance of the object. Then the camera projection can be approximated as:

$$
\begin{aligned}
& x=f \frac{X}{Z} \approx \frac{f}{Z} X \\
& y=f \frac{Y}{Z} \approx \frac{f}{Z} Y
\end{aligned}
$$

This is the weak-perspective camera model. Sometimes called scaled orthography.

## Weak Perspective Projection


plane

## Weak Perspective

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5 Camera Models


## Impact of different projections

- Perspective projections
- Parallel lines in world are not parallel in the image
- Object projection gets smaller with distance from camera
- Weak perspective projection
- Parallel lines in the world are parallel in the image
- Object projection gets smaller with distance from camera
- Orthographic projection
- Parallel lines in the world are parallel in the image
- Object projection is unchanged with distance from camera


## Image distortion due to optics

- Radial distortion which depends on radius $r$, distance of each point from center of image
- $r^{2}=(x-0 x)^{2}+(y-O y)^{2}$


Radial distortion


$$
\begin{aligned}
& x_{\text {conserted }}=x\left(1+k_{1} r^{2}+k_{2} r^{4}+k_{3} r^{6}\right) \\
& y_{\text {cartected }}=y\left(1+k_{1} r^{2}+k_{2} r^{4}+k_{3} r^{6}\right)
\end{aligned}
$$

Correction uses three parameters, $\mathrm{k}_{1}, \mathrm{k}_{2}, \mathrm{k}_{3}$

## Correcting Radial Distortions



## Tangential Distortion

- Lens not exactly parallel to the image plane

$$
\begin{aligned}
& x_{\text {cormesead }}=x+\left[2 p_{1} y+p_{2}\left(r^{2}+2 x^{2}\right)\right] \\
& y_{\text {cormestad }}=y+\left[p_{1}\left(r^{2}+2 y^{2}\right)+2 p_{2} x\right]
\end{aligned}
$$



Tangential distortion

- Correction uses two parameter $\mathrm{p}_{1}, \mathrm{p}_{2}$
- Both types of distortion are removed (image is un-distorted) and only then does standard calibration matrix K apply to the image
- Camera calibration computes both K and these five distortion parameters


## How to find the camera parameters

## - Can use the EXIF tag for any digital image

- Has focal length fin millimeters but not the pixel size
- But you can get the pixel size from the camera manual
- There are only a finite number of different pixels sizes because the number of sensing element sizes is limited
- If there is not a lot of image distortion due to optics then this approach is sufficient (this is only a linear calibration)
- Can perform explicit camera calibration
- Put a calibration pattern in front of the camera
- Take a number of different pictures of this pattern
- Now run the calibration algorithm (different types)
- Result is intrinsic camera parameters (linear and non-linear) and the extrinsic camera parameters of all the images

