# Lab Course Kinect Programming for Computer Vision 

# Transformations and Camera Calibration 

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## Lecture Outline

## Part I: Basics of Transformations

## Part II: Basics of Cameras

## Outline Part I

- 2D Transformations
- Homogeneous Coordinates in 2D
- Hierarchy of Transformations in 2D
- Hierarchy of Transformations in 3D


## 2D Transformations

The basic building blocks of transformations are:

- Scaling (isotropic, non-isotropic)
- Rotation (around one rotation axis)
- Translation (motion along the two axes)

Transformations are defined by their action on point coordinates.

Notation:

$$
\underset{\text { original point }}{\mathbf{p}=} \underset{\substack{(x, y)^{\top}}}{\left[\begin{array}{l}
x \\
y
\end{array}\right]} \quad \mathbf{p}^{\prime}=\left(x^{\prime}, y^{\prime}\right)^{\top}=\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]
$$

## 2D Transformations - Scaling

Isotropic scaling: same scaling factor in both spatial dimensions


$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{ll}
s & 0 \\
0 & s
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## 2D Transformations - Scaling

Non-isotropic scaling: different scaling factors in the two spatial dimensions


$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
s_{x} & 0 \\
0 & s_{y}
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## 2D Transformations - Rotation



$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{rr}
\cos \theta & -\sin \theta \\
\sin \theta & \cos \theta
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

## 2D Transformations - Rotation

$$
\mathbf{p}=\binom{x}{y}=\binom{r \cos (\alpha)}{r \sin (\alpha)}
$$

$$
\begin{aligned}
\mathbf{p}^{\prime} & =\binom{r \cos (\alpha+\beta)}{r \sin (\alpha+\beta)} \\
& =\binom{r \cos (\alpha) \cos (\beta)-r \sin (\alpha) \sin (\beta)}{r \sin (\alpha) \cos (\beta)+r \cos (\alpha) \sin (\beta)} \\
& =\binom{x \cos (\beta)-y \sin (\beta)}{y \cos (\beta)+x \sin (\beta)} \\
& =\left(\begin{array}{cc}
\cos (\beta) & -\sin (\beta) \\
\sin (\beta) & \cos (\beta)
\end{array}\right)\binom{x}{y} \\
& =\left(\begin{array}{cc}
\cos (\beta) & -\sin (\beta) \\
\sin (\beta) & \cos (\beta)
\end{array}\right) \mathbf{p}
\end{aligned}
$$

## 2D Transformations - Translation



$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{l}
x \\
y
\end{array}\right]+\left[\begin{array}{l}
t_{x} \\
t_{y}
\end{array}\right]
$$

## 2D Transformations

Have you noticed we have two types of representations for the basic transformations seen so far?

Scaling, Rotation: Matrix-vector product

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{cc}
s_{x} & 0 \\
0 & s_{y}
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right] \quad\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{rr}
\cos \theta & -\sin \theta \\
\sin \theta & \cos \theta
\end{array}\right]\left[\begin{array}{l}
x \\
y
\end{array}\right]
$$

Translation: Sum of vectors

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime}
\end{array}\right]=\left[\begin{array}{l}
x \\
y
\end{array}\right]+\left[\begin{array}{l}
t_{x} \\
t_{y}
\end{array}\right]
$$

How to achieve a matrix-vector product form for all types of transformations?

## Outline Part I

- 2D Transformations
- Homogeneous Coordinates in 2D
- Hierarchy of Transformations in 2D
- Hierarchy of Transformations in 3D


## Homogeneous Coordinates in 2D

How to achieve a matrix-vector product form for all types of transformations?

- Write a 2D point in homogeneous coordinates:

$$
\mathbf{p}=\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right] \in \mathbb{R}^{3}
$$

> We will see more properties of homogeneous representation later.

- Then, translation can easily be expressed as a matrix-vector product:

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

## Homogeneous Coordinates in 2D

- Rotation in homogeneous representation:

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

- Combined transformations (here: rotation and translation):

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
\cos \theta & -\sin \theta & t_{x} \\
\sin \theta & \cos \theta & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]=\left[\begin{array}{c}
x \cos \theta-y \sin \theta+t_{x} \\
x \sin \theta+y \cos \theta+t_{y} \\
1
\end{array}\right]
$$

## 2D Projective Space

When dealing with points represented in homogeneous coordinates, we are working in a Projective space.

- Points in 2D Euclidean space are represented by Cartesian coordinates

$$
\mathbf{p}=\binom{x}{y} \in \mathbb{R}^{2}
$$

- Points in 2D Projective space are represented by Homogeneous coordinates:

$$
\mathbf{p}=\left(\begin{array}{l}
x \\
y \\
1
\end{array}\right) \in \mathbb{P}^{2}
$$

## 2D Projective Space



Point in Euclidean 2D space


Point in Projective 2D space.
All points on the orange line represent the same point. A homogeneous vector in 2D projective space represents an equivalence class of points.

## Points and Lines in 2D Projective Space

## Properties of points

- A 2D point $(x, y)^{\top}$ has the homogeneous representation $(x, y, 1)^{\top}$.
- The homogeneous vector $(x, y, w)^{\top}$ with $\mathrm{w} \neq 0$ represents the point $\left(\frac{x}{w}, \frac{y}{w}, 1\right)^{\top}$.
- For any non-zero $k,(x, y, w)^{\top}$ and $k(x, y, w)^{\top}$ represent the same point.


## Points and Lines in 2D Projective Space

## Properties of lines

- A line $a x+b y+c=0$ has the homogeneous representation $\mathbf{l}=(a, b, c)^{\top}$.
- A point $\mathbf{x}$ lies on a line $\mathbf{l}$ if and only if $\mathbf{x}^{\top} \mathbf{l}=0$.
- Two lines $\mathbf{l}$ and $\mathbf{l}^{\prime}$ intersect in the point $\mathbf{x}=\mathbf{l} \times \mathbf{l}^{\prime}$
- The line joining the two points $\mathbf{x}$ and $\mathbf{x}^{\prime}$ is $\mathbf{l}=\mathbf{x} \times \mathbf{x}^{\prime}$.


## Points and Lines in 2D Projective Space

## Special points and lines

- Points with $3^{\text {rd }}$ coordinate zero $(x, y, 0)^{\top}$ are called points at infinity.
- All points at infinity lie on the line at infinity $\mathbf{l}_{\infty}=(0,0,1)^{\top}$.

$$
\mathbf{l}_{\infty}^{\top}(x, y, 0)=(0,0,1)^{\top}(x, y, 0)=0
$$

- Parallel lines intersect at infinity.

In projective space, all lines intersect in a point, even parallel lines.

## Outline Part I

- 2D Transformations
- Homogeneous Coordinates in 2D
- Hierarchy of Transformations in 2D
- Hierarchy of Transformations in 3D


## Hierarchy of Transformations in 2D

- Classification of transformations in terms of quantities or properties that are invariant (left unchanged) by the transformation
- Hierarchy: Starting from most specialized type of transformation, successively remove invariants to get more general types of transformations

| Most general: | Projective transformations |
| :---: | :--- |
|  | Affine transformations |
|  | Similarity transformations |
| Most specialized: | Euclidean transformations |

- Containment relation from general to specialized, e.g.: affine transformations have all properties of projective transformations, plus additional ones.

Hierarchy of Transformations in 2D:
Euclidean Transformations

- Transformation:

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
\cos \theta & -\sin \theta & t_{x} \\
\sin \theta & \cos \theta & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {euclidean }}=\left[\begin{array}{cc}
\mathbf{R} & \mathbf{t} \\
\mathbf{0}^{\top} & 1
\end{array}\right] \quad \begin{array}{r}
\mathbf{R}^{\top} \mathbf{R}=\mathbf{I} \\
\operatorname{det}(\mathbf{R})=1
\end{array}
$$

- Degrees of freedom:
- 3 (1 rotation, 2 translation)
- Invariants: length, area, angles



## Hierarchy of Transformations in 2D: Similarity Transformations

- Transformation:

$$
\left[\begin{array}{l}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
s \cos \theta & -s \sin \theta & t_{x} \\
s \sin \theta & s \cos \theta & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {similarity }}=\left[\begin{array}{cc}
s \mathbf{R} & \mathbf{t} \\
\mathbf{0}^{\top} & 1
\end{array}\right] \quad \begin{aligned}
\mathbf{R}^{\top} \mathbf{R} & =\mathbf{I} \\
\operatorname{det}(\mathbf{R}) & =1
\end{aligned}
$$

- Degrees of freedom:
- 4 (1 rotation, 2 translation, 1 isotropic scaling)
- Invariants: ratio of lengths, angles, parallel lines


Hierarchy of Transformations in 2D: Affine Transformations

- Transformation:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
a_{11} & a_{12} & t_{x} \\
a_{21} & a_{22} & t_{y} \\
0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {affinity }}=\left[\begin{array}{cc}
\mathbf{A} & \mathbf{t} \\
\mathbf{0}^{\top} & 1
\end{array}\right] \quad \mathbf{A} \text { non-singular }
$$

- Degrees of freedom:
- 6 (2 translation, 4 entries of $\mathbf{A}$ )
- Invariants: parallel lines, ratios of areas, line at infinity


Hierarchy of Transformations in 2D: Affine Transformations

Affine transformations geometrically consist of two components:


$$
\mathbf{A}=\mathbf{R}(\theta) \mathbf{R}(-\phi)\left[\begin{array}{cc}
\lambda_{1} & 0 \\
0 & \lambda_{2}
\end{array}\right] \mathbf{R}(\phi)
$$



## Hierarchy of Transformations in 2D: <br> Projective Transformations

- Transformation:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{ccc}
a_{11} & a_{12} & t_{x} \\
a_{21} & a_{22} & t_{y} \\
v_{1} & v_{2} & v
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {projectivity }}=\left[\begin{array}{cc}
\mathbf{A} & \mathbf{t} \\
\mathbf{v}^{\top} & v
\end{array}\right] \quad \mathbf{A} \text { non-singular }
$$

- Degrees of freedom:
- 8 (9 arbitrary matrix entries, up to scale)
- Invariants: Intersection points, tangency, cross-ratios



## Hierarchy of Transformations in 2D - Summary

| Projective | 8 DOF |  | $\left[\begin{array}{cc}\mathbf{A} & \mathbf{t} \\ \mathbf{v}^{\top} & v\end{array}\right]$ |
| :---: | :---: | :---: | :---: |
| Affine | 6 DOF |  $\square$ | $\left[\begin{array}{cc}\mathbf{A} & \mathbf{t} \\ \mathbf{0}^{\top} & 1\end{array}\right]$ |
| Similarity | 4 DOF |  | $\left[\begin{array}{ll}s \mathbf{R} & \mathbf{t} \\ \mathbf{0}^{\top} & 1\end{array}\right]$ |
| Euclidean | 3 DOF |  | $\left[\begin{array}{cc}\mathbf{R} & \mathbf{t} \\ \mathbf{0}^{\top} & 1\end{array}\right]$ |

## Action of Transformations in 2D on Points at Infinity

- Affine Transformations:

$$
\left[\begin{array}{ccc}
{\left[\begin{array}{ccc}
a_{11} & a_{12} & t_{x} \\
a_{21} & a_{22} & t_{y} \\
0 & 0 & 1
\end{array}\right]}
\end{array}\left[\begin{array}{c}
x \\
y \\
0
\end{array}\right]=\left[\begin{array}{c}
a_{11} x+a_{12} y \\
a_{21} x+a_{22} y \\
0
\end{array}\right]\right.
$$

## Action of Transformations in 2D on Points at Infinity

- Projective Transformations:

$$
\left[\begin{array}{ccc}
a_{11} & a_{12} & t_{x} \\
a_{21} & a_{22} & t_{y} \\
v_{1} & v_{2} & v
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
0
\end{array}\right]=\left[\begin{array}{c}
a_{11} x+a_{12} y \\
a_{21} x+a_{22} y \\
v_{1} x+v_{2} y
\end{array}\right]
$$

Point at infinity,
Mapped by projective e.g. intersection transformation to a finite of blue lines point (vanishing point)


## Outline Part I

- 2D Transformations
- Homogeneous Coordinates in 2D
- Hierarchy of Transformations in 2D
- Hierarchy of Transformations in 3D


## Homogeneous Coordinates in 3D

How to achieve a matrix-vector product form for all types of transformations?

- Write a 3D point in homogeneous coordinates:

$$
\mathbf{p}=\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right] \in \mathbb{R}^{4}
$$

- 3D projective space: properties analogous to 2D homogeneous representation
- All types of transformations (scaling, translation, rotation, etc.) can be written as a matrix-vector product
- Several transformations can be composed into one transformation matrix

Hierarchy of Transformations in 3D (1): Euclidean Transformations

- Transformation:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
z^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{cccc}
r_{11} & r_{12} & r_{13} & t_{x} \\
r_{21} & r_{22} & r_{23} & t_{y} \\
r_{31} & r_{32} & r_{33} & t_{z} \\
0 & 0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {euclidean }}=\left[\begin{array}{cc}
\mathbf{R} & \mathbf{t} \\
\mathbf{0}^{\top} & 1
\end{array}\right] \quad \begin{array}{r}
\mathbf{R}^{\top} \mathbf{R}=\mathbf{I} \\
\operatorname{det}(\mathbf{R})=1
\end{array}
$$

- Degrees of freedom:
- 6 (3 rotation, 3 translation)
- Invariants: volume, angles

Hierarchy of Transformations in 3D (2): Similarity Transformations

- Transformation:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
z^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{cccc}
s r_{11} & s r_{12} & s r_{13} & t_{x} \\
s r_{21} & s r_{22} & s r_{23} & t_{y} \\
s r_{31} & s r_{32} & s r_{33} & t_{z} \\
0 & 0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {similarity }}=\left[\begin{array}{cc}
s \mathbf{R} & \mathbf{t} \\
\mathbf{0}^{\top} & 1
\end{array}\right] \quad \begin{array}{r}
\mathbf{R}^{\top} \mathbf{R}=\mathbf{I} \\
\operatorname{det}(\mathbf{R})=1
\end{array}
$$

- Degrees of freedom:
- 7 (3 rotation, 3 translation, 1 scaling)
- Invariants: angles

Hierarchy of Transformations in 3D (3): Affine Transformations

- Transformation:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
z^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{cccc}
a_{11} & a_{12} & a_{13} & t_{x} \\
a_{21} & a_{22} & a_{23} & t_{y} \\
a_{31} & a_{32} & a_{33} & t_{z} \\
0 & 0 & 0 & 1
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {affinity }}=\left[\begin{array}{cc}
\mathbf{A} & \mathbf{t} \\
\mathbf{0}^{\top} & 1
\end{array}\right] \quad \mathbf{A} \text { non-singular }
$$

- Degrees of freedom:
- 12 (3 translation, 9 entries of $\mathbf{A}$ )
- Invariants: parallel planes, ratios of volumes, plane at infinity

Hierarchy of Transformations in 3D (4) Projective Transformations

- Transformation:

$$
\left[\begin{array}{c}
x^{\prime} \\
y^{\prime} \\
z^{\prime} \\
1
\end{array}\right]=\left[\begin{array}{cccc}
a_{11} & a_{12} & a_{13} & t_{x} \\
a_{21} & a_{22} & a_{23} & t_{y} \\
a_{31} & a_{32} & a_{33} & t_{z} \\
v_{1} & v_{2} & v_{3} & v
\end{array}\right]\left[\begin{array}{l}
x \\
y \\
z \\
1
\end{array}\right]
$$

- Concise notation for transformation matrix:

$$
\mathbf{T}_{\text {projectivity }}=\left[\begin{array}{cc}
\mathbf{A} & \mathbf{t} \\
\mathbf{v}^{\top} & v
\end{array}\right] \quad \mathbf{A} \text { non-singular }
$$

- Degrees of freedom:
- 15 (16 arbitrary matrix entries, up to scale)
- Invariants: Intersection points, tangency


## Hierarchy of Transformations in 3D - Summary

$\left.\begin{array}{ll}\text { Projective } & 15 \mathrm{DOF} \\ \text { Affine } & 12 \mathrm{DOF} \\ \text { Similarity } & {\left[\begin{array}{cc}\mathbf{A} & \mathbf{t} \\ \mathbf{v}^{\top} & v\end{array}\right]} \\ \mathbf{0}^{\top} & 1\end{array}\right]$

## Lecture Outline

## Part I: Basics of Transformations

## Part II: Basics of Cameras

Pinhole Camera Model
Mapping between 3D world and 2D image by central projection



$$
(X, Y, Z)^{\top} \mapsto(f X / Z, f Y / Z, f)^{\top}
$$

$$
(X, Y, Z)^{\top} \mapsto(f X / Z, f Y / Z)^{\top}=(x, y)^{\top}
$$

## Pinhole Camera Model

In homogeneous coordinates...


$$
\left(\begin{array}{c}
f X \\
f Y \\
Z
\end{array}\right)=\left[\begin{array}{llll}
f & & & 0 \\
& f & & 0 \\
& & 1 & 0
\end{array}\right]\left(\begin{array}{c}
X \\
Y \\
Z \\
1
\end{array}\right)
$$

## Pinhole Camera Model

Mapping between 3D world and 2D image by central projection


$$
\begin{aligned}
\left(\begin{array}{c}
f X \\
f Y \\
Z
\end{array}\right)= & {\left[\begin{array}{lll}
f & & \\
& f & \\
& & 1
\end{array}\right]\left[\begin{array}{llll}
1 & & & 0 \\
& 1 & & 0 \\
& & 1 & 0
\end{array}\right]\left(\begin{array}{c}
X \\
Y \\
Z \\
1
\end{array}\right) } \\
& \mathbf{x}=\operatorname{diag}(f, f, 1)[\mathbf{I} \mid \mathbf{0}] \mathbf{X}
\end{aligned}
$$

## Pinhole Camera Model

Origin of image plane is not necessarily at the camera center


$$
(X, Y, Z) \mapsto\left(f X / Z+p_{x}, f Y / Z+p_{y}\right)
$$

## Pinhole Camera Model

Origin of image plane is not necessarily at the camera center

$$
\begin{gathered}
\left(\begin{array}{c}
f X+Z p_{x} \\
f Y+Z p_{x} \\
Z
\end{array}\right)=\left[\begin{array}{llll}
f & & p_{x} & 0 \\
& f & p_{y} & 0 \\
& & 1 & 0
\end{array}\right]\left(\begin{array}{c}
X \\
Y \\
Z \\
1
\end{array}\right) \\
\mathbf{x}=\mathbf{K}[\mathbf{I} \mid \mathbf{0}] \mathbf{X} \\
\mathbf{K}=\left[\begin{array}{lll}
f & & p_{x} \\
& f & p_{y} \\
& & 1
\end{array}\right] \begin{array}{l}
\text { intrinsic camera } \\
\text { parameters matrix }
\end{array}
\end{gathered}
$$

## Pinhole Camera Model

Camera center is not necessarily at the world coordinate system origin


$$
\begin{gathered}
\tilde{\mathbf{C}} \\
\begin{array}{c}
\text { camera center } \\
\text { in world coordinates }
\end{array} \\
\mathbf{X}_{\text {cam }}=\left[\begin{array}{cc}
\mathrm{R} & -\mathrm{R} \tilde{\mathrm{C}} \\
0 & 1
\end{array}\right] \mathbf{X} \\
\begin{array}{c}
\text { 3D point in camera } \\
\text { coordinates }
\end{array}
\end{gathered}
$$

$\mathbf{x}=\mathrm{K}[\mathrm{I} \mid 0] \mathbf{X}_{\text {cam }}$ projection to image plane from camera coordinates
$\mathbf{x}=\mathrm{KR}[\mathrm{II}-\tilde{\mathrm{C}}] \mathbf{X} \quad$ projection to image plane from world coordinates

## Pinhole Camera Model

Camera center is not necessarily at the world coordinate system origin


## CCD Camera Model

Camera coordinates are mapped to pixels of the CCD chip $m_{x}, m_{y}$ number of pixels per unit distance in camera coordinates


## Finite Projective Camera Model

Both axes in camera coordinates need not have identical scaling
$S$ skew parameter

$$
\begin{aligned}
& \mathbf{K}=\left[\begin{array}{ccc}
\alpha_{x} & s & x_{0} \\
& \alpha_{y} & y_{0} \\
& & 1
\end{array}\right] \quad 5 \mathrm{DOF} \\
& \mathrm{P}=\mathbf{K} \mathbf{R}[\mathrm{I} \mid-\tilde{\mathrm{C}}] \quad 11 \mathrm{DOF}
\end{aligned}
$$

## Camera Calibration

- Relating the ideal camera model to the properties of an actual physical device and determining the position and orientation of the camera w.r.t. the world coordinate frame ${ }^{1}$.
- Estimating the entries of $\mathbf{K}, \mathbf{R}, \tilde{\mathrm{C}}$
- Camera models with increasing generality:
- pinhole camera

9 DOF

- CCD camera

10 DOF

- finite projective camera
- In addition: lens distortion parameters

$$
\begin{aligned}
\mathbf{K}= & {\left[\begin{array}{ccc}
\alpha_{x} & s & x_{0} \\
& \alpha_{y} & y_{0} \\
& & 1
\end{array}\right] } \\
& \mathrm{P}=\mathbf{K} \mathbf{R}[\mathrm{I} \mid-\tilde{\mathrm{C}}]
\end{aligned}
$$

## Camera Calibration



Checkerboard pattern captured from multiple views (the checker sizes in the real world are known)


Semi-automatic measurements to find sizes in each of the images

- Linear and non-linear optimization for estimating camera parameters
- MATLAB calibration toolbox: http://www.vision.caltech.edu/bouguetj/calib_doc


## Camera Calibration

## Calibration results after optimization (with uncertainties):



Extrinsic parameters



## Stereo Calibration

- Find the spatial relation between two cameras observing the same scene (stereo setup)
- For each orientation of the calibration pattern, capture corresponding left and right camera images
- Calibrate each camera separately, as before
- Perform stereo calibration using the result of separate calibration as an input



## Calibrating the Kinect

- Depth and RGB images do not match exactly, there is an offset
- Stereo calibration for getting both images from the same camera perspective


RGB Image


Depth Image

## Calibrating the Kinect

- Kinect has two cameras with standard optics
- RGB camera
- Infrared (IR) camera
- Infrared image is used internally for computing depth image, but is accessible through the API
- Checkerboard pattern is not visible in depth images, but visible in IR images



RGB Image


Infrared Image

## Relating RGB and Depth Images

Stereo calibration gives $\mathbf{R}, \mathbf{t}$ and $\mathbf{K}_{\mathrm{rgb}}, \mathbf{K}_{\mathrm{ir}}$.
Step 1: Back-project every 2D point from depth image into 3D space Step 2: Apply rigid transformation between two cameras to 3D points Step 3: Project every transformed 3D point into the RGB image


## Relating RGB and Depth Images

Step 1: Back-project every 2D point from depth image into 3D space

- ToF data: depth $d$ in meters for every pixel location $\mathbf{x}=(x, y)^{\top}$
- Desired data: 3D coordinates $\mathbf{X}=(X, Y, Z)^{\top}$ for every pixel


Depth Image


3D Point Cloud

## Relating RGB and Depth Images

Step 1: Back-project every 2D point from depth image into 3D space

Apply inverse of IR camera projection to obtain 3D points:

$$
\begin{gathered}
\mathbf{x}_{\text {ir }}=\mathbf{P}_{\mathrm{ir}} \mathbf{X}=\mathbf{K}_{\mathrm{ir}}[\mathbf{I} \mid \mathbf{0}] \mathbf{X} \\
\left(\begin{array}{c}
x \\
y \\
w
\end{array}\right)=\left(\begin{array}{cccc}
f_{x} & 0 & c_{x} & 0 \\
0 & f_{y} & c_{y} & 0 \\
0 & 0 & 1 & 0
\end{array}\right)\left(\begin{array}{c}
X \\
Y \\
Z \\
W
\end{array}\right)=\left(\begin{array}{c}
f_{x} X+c_{x} Z \\
f_{y} Y+c_{y} Z \\
Z
\end{array}\right) \\
X=\frac{\left(x-c_{x}\right) Z}{f_{x}} \quad Y=\frac{\left(y-c_{y}\right) Z}{f_{y}} \\
\text { Inverse relation for } \mathbf{X} \text { and } \mathrm{Y}
\end{gathered}
$$

## Relating RGB and Depth Images

Step 1: Back-project every 2D point from depth image into 3D space
For every depth pixel at location $\mathbf{x}=(x, y)^{\top}$ with depth $d$, compute a 3D point $\mathbf{X}=(X, Y, Z)^{\top}$ with coordinates:

$$
\left(\begin{array}{l}
X \\
Y \\
Z
\end{array}\right)=\left(\begin{array}{c}
\frac{\left(x-c_{x}\right) d}{f_{x}} \\
\frac{\left(y-c_{y}\right) d}{f_{y}} \\
d
\end{array}\right)
$$

$c_{x}, c_{y}, f_{x}, f_{y}$
Intrinsics of the IR (depth) camera


## Relating RGB and Depth Images

Step 2: Apply rigid transformation between two cameras to 3D points Step 3: Project every transformed 3D point into the RGB image

$$
\mathbf{x}_{\mathrm{rgb}}=\mathbf{P}_{\mathrm{rgb}} \mathbf{X}=\mathbf{K}_{\mathrm{rgb}}[\mathbf{R} \mid \mathbf{t}] \mathbf{X}
$$

- $\quad \mathbf{X}_{\mathrm{rgb}}$ are the coordinates of the projection of $X$ into the RGB image
- look up color corrsponding to a given 3D point



## Assignment 3

- Extend your application such that it can read out Kinect IR images
- Extend your application for saving IR and RGB images to bitmap files
- Familiarize yourself with the Matlab Camera Calibration Toolbox http://www.vision.caltech.edu/bougueti/calib doc/
- Read the examples (especially examples 1 and 5)
- Download the toolbox
- Try out the examples
- Perform stereo-calibration for your Kinect device
- Note that you cannot extract IR and RGB images simultaneously...


## Outlook Assignment 4:

- Implement a 3D point cloud visualization for kinect depth data
- Add a feature to color the 3D points with the colors from the RGB camera

