2. Image Formation





2.1 Geometric Primitives & Transformations









- Geometric primitives
- 2D transformations
- 3D transformations
- ✤ 3D rotations
- ✤ 3D to 2D projections
- Lens Distortions





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2D Points



2D point (e.g., a pixel coordinate in an image):

$$oldsymbol{x} = (x,y) \in \mathcal{R}^2$$
 or $oldsymbol{x} = \left[egin{array}{c} x \\ y \end{array}
ight]$

In homogeneous coordinates:

$$\tilde{\boldsymbol{x}} = (\tilde{x}, \tilde{y}, \tilde{w}) \in \mathcal{P}^2$$

 $\mathcal{P}^2 = \mathcal{R}^3 - (0,0,0)$ is called the projective space.

Vectors that differ only by a scale considered equivalent:

 $s\tilde{x} = \tilde{x} \, \forall s \in \mathcal{R}$

Augmented Vectors



A homogenous vector can be converted back to an *inhomogeneous* vector by dividing by the last element:

Why Homogeneous Coordinates?

- Provide a natural representation for points at infinity.
- Allow common geometric transformations (e.g., translation, rotation, scaling, perspective projection) to be effected by matrix multiplication





August Ferdinand Möbius (1790–1868)

2D Lines



2D lines can also be represented using homogeneous coordinates $\tilde{l} = (a, b, c)$.

The corresponding *line equation* is

$$\bar{\boldsymbol{x}}\cdot\tilde{\boldsymbol{l}}=ax+by+c=0.$$

We can normalize the line equation vector so that $\boldsymbol{l} = (\hat{n}_x, \hat{n}_y, -d) = (\hat{\boldsymbol{n}}, -d)$ with $||\hat{\boldsymbol{n}}|| = 1$.

Then:

 \hat{n} is the unit normal to the line, directed toward the line from the origin d is the distance of the line from the origin

$$\hat{\boldsymbol{n}} = (\hat{n}_x, \hat{n}_y) = (\cos\theta, \sin\theta)$$







When using homogeneous coordinates, we can compute the intersection of two lines as

 l_1

 $oldsymbol{ ilde{x}} = oldsymbol{ ilde{l}}_1 imes oldsymbol{ ilde{l}}_2 \ oldsymbol{ ilde{l}}_2 \ oldsymbol{ ilde{l}}_2 \ oldsymbol{ ilde{x}} oldsymbol{ ilde{l}}_2$

Similarly, the line joining two points can be written as

 $egin{aligned} ilde{m{l}} = ilde{m{x}}_1 imes ilde{m{x}}_2 & egin{aligned} ilde{m{x}}_2 & egin{aligned} ilde{m{x}}_2 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_2 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_2 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_2 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_1 & egin{aligned} ilde{m{x}}_2 & egin{aligned} ilde{m{$

3D Points



Straightforward extension from 2D:

Inhomogeneous: $\boldsymbol{x} = (x, y, z) \in \mathcal{R}^3$

Homogeneous: $\tilde{\boldsymbol{x}} = (\tilde{x}, \tilde{y}, \tilde{z}, \tilde{w}) \in \mathcal{P}^3$

Augmented: $\bar{\boldsymbol{x}} = (x, y, z, 1)$

$$\tilde{\boldsymbol{x}} = \tilde{w} \bar{\boldsymbol{x}}$$







3D planes can also be represented as homogeneous coordinates $\tilde{m} = (a, b, c, d)$ with a corresponding plane equation

$$\bar{\boldsymbol{x}}\cdot\tilde{\boldsymbol{m}}=a\boldsymbol{x}+b\boldsymbol{y}+c\boldsymbol{z}+d=0$$

We can also normalize the plane equation as $\boldsymbol{m} = (\hat{n}_x, \hat{n}_y, \hat{n}_z, d) = (\boldsymbol{\hat{n}}, d)$ with $\|\boldsymbol{\hat{n}}\| = 1$.

 \hat{n} is the *normal vector* perpendicular to the plane





3D Lines



A 3D line can be represented using two points p and q that lie on the line. Any point r that also lies on the line can then be represented as

 $\boldsymbol{r} = (1 - \lambda) \boldsymbol{p} + \lambda \boldsymbol{q}$

If we restrict $0 \le \lambda \le 1$, we get the *line segment* joining p and q.

If we use homogeneous coordinates, we can write the line as

$$\tilde{\boldsymbol{r}} = \mu \tilde{\boldsymbol{p}} + \lambda \tilde{\boldsymbol{q}}.$$

A special case of this is when the second point is at infinity, i.e., $\tilde{q} = (\hat{d}_x, \hat{d}_y, \hat{d}_z, 0) = (\hat{d}, 0)$.

where \hat{d} is the direction of the line.

Then:

$$m{r} = m{p} + \lambda m{\hat{d}}$$







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2D translations can be written as x' = x + t or

$$x' = \left[egin{array}{cc} I & t \end{array}
ight]ar{x}$$

where I is the (2×2) identity matrix

or

$$ar{x}' = \left[egin{array}{ccc} I & t \ 0^T & 1 \end{array}
ight]ar{x}$$

Note: Whenever an augmented vector appears on both sides, it can be replaced by a full homogenous vector.



Euclidean Transformation (2D Rotation + Translation)

$$x' = Rx + t$$
 or
 $x' = \begin{bmatrix} R & t \end{bmatrix} \overline{x}$
where
 $R = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

is an orthonormal rotation matrix with $\mathbf{R}\mathbf{R}^T = \mathbf{I}$ and $|\mathbf{R}| = 1$.

Preserves Euclidean distances



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Similarity Transformation



Preserves angles



YOR



Affine Transformation



 $x' = A\bar{x}$, where A is an arbitrary 2×3 matrix

$$m{x}' = \left[egin{array}{cccc} a_{00} & a_{01} & a_{02} \ a_{10} & a_{11} & a_{12} \end{array}
ight] m{ar{x}}$$

Preserves parallelism



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Projective Transformation (Homography)

 $\tilde{x}' = \tilde{H}\tilde{x}$ - NB: Should be an augmented vector, I think. Both Szeliski and Hartley & Zisserman write it as homogeneous.

where \tilde{H} is an arbitrary 3×3 matrix.

 \tilde{H} is homogenous:

Two \tilde{H} matrices that differ only by a scale factor are equivalent.

$$x' = \frac{h_{00}x + h_{01}y + h_{02}}{h_{20}x + h_{21}y + h_{22}} \text{ and } y' = \frac{h_{10}x + h_{11}y + h_{12}}{h_{20}x + h_{21}y + h_{22}}$$

Preserves straight lines



Summary of 2D Transformations



Nested set of groups

- Closed under composition
- Each transformation has an inverse that is a member of the same group

Transformation	Matrix	# DoF	Preserves	Icon
translation	$\left[egin{array}{c c} I & t \end{array} ight]_{2 imes 3}$	2	orientation	
rigid (Euclidean)	$\left[egin{array}{c c} m{R} & t \end{array} ight]_{2 imes 3}$	3	lengths	\bigcirc
similarity	$\left[\begin{array}{c c} s oldsymbol{R} & t \end{array} ight]_{2 imes 3}$	4	angles	\bigcirc
affine	$\left[egin{array}{c} m{A} \end{array} ight]_{2 imes 3}$	6	parallelism	
projective	$\left[egin{array}{c} ilde{m{H}} \end{array} ight]_{3 imes 3}$	8	straight lines	



Co-vectors



We now know how to transform points. How do we transform lines?

$$\tilde{\boldsymbol{l}} \cdot \tilde{\boldsymbol{x}} = 0$$

$$\tilde{\boldsymbol{x}}' = \tilde{\boldsymbol{H}}\tilde{\boldsymbol{x}}$$

$$\tilde{\boldsymbol{l}}' \cdot \tilde{\boldsymbol{x}}' = \tilde{\boldsymbol{l}}^{T}\tilde{\boldsymbol{H}}\tilde{\boldsymbol{x}} = (\tilde{\boldsymbol{H}}^{T}\tilde{\boldsymbol{l}}')^{T}\tilde{\boldsymbol{x}} = \tilde{\boldsymbol{l}} \cdot \tilde{\boldsymbol{x}} = 0$$

Thus

$$\tilde{\boldsymbol{l}}' = \tilde{\boldsymbol{H}}^{-T}\tilde{\boldsymbol{l}}.$$

i.e., the action of a projective transformation on a *co-vector* such as a 2D line can be represented by the transposed inverse of the matrix.





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3D Translation



3D translations can be written as x' = x + t or

$$oldsymbol{x}' = \left[egin{array}{cc} oldsymbol{I} & t \end{array}
ight]oldsymbol{ar{x}}$$

where I is the (3×3) identity matrix



 $egin{aligned} x' &= oldsymbol{R} x + t \ x' &= igg[egin{aligned} R & t \ igg] oldsymbol{ar{x}} \end{aligned}$

where R is a 3×3 orthonormal rotation matrix with $RR^T = I$ and |R| = 1

Preserves Euclidean distances

Similarity Transformation



$$egin{aligned} & x' = s oldsymbol{R} x + t \ & x' = \left[egin{aligned} s oldsymbol{R} & t \end{array}
ight] oldsymbol{ar{x}} \end{aligned}$$

Preserves angles

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Affine Transformation



 $x' = A\bar{x}$, where A is an arbitrary 3×4 matrix

Preserves parallelism

Projective Transformation (Homography)

YORK

 $ilde{x}' = ilde{H} ilde{x}$

where H is an arbitrary 4×4 homogeneous matrix

Preserves straight lines

Summary of 3D Transformations

Transformation	Matrix	# DoF	Preserves	Icon
translation	$\left[egin{array}{c c} oldsymbol{I} & t \end{array} ight]_{3 imes 4}$	3	orientation	
rigid (Euclidean)	$\left[egin{array}{c c} m{R} & t \end{array} ight]_{3 imes 4}$	6	lengths	\bigcirc
similarity	$\left[\begin{array}{c c} s oldsymbol{R} & t \end{array} ight]_{3 imes 4}$	7	angles	\bigcirc
affine	$\left[egin{array}{c} oldsymbol{A} \end{array} ight]_{3 imes 4}$	12	parallelism	
projective	$\left[egin{array}{c} ilde{oldsymbol{H}} \end{array} ight]_{4 imes 4}$	15	straight lines	



YOR



End of Lecture Sept 10, 2018





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Let u be the result of rotating vector v about axis \hat{n} by the angle θ .

First, project the vector v onto the axis \hat{n} :

$$\boldsymbol{v}_{\parallel} = \boldsymbol{\hat{n}}(\boldsymbol{\hat{n}}\cdot\boldsymbol{v}) = (\boldsymbol{\hat{n}}\boldsymbol{\hat{n}}^T)\boldsymbol{v}$$

Next, compute the perpendicular residual:

$$\boldsymbol{v}_{\perp} = \boldsymbol{v} - \boldsymbol{v}_{\parallel} = (\boldsymbol{I} - \boldsymbol{\hat{n}} \boldsymbol{\hat{n}}^T) \boldsymbol{v}$$



We can rotate this vector by 90° using the cross product,

$$\boldsymbol{v}_{\times} = \boldsymbol{\hat{n}} \times \boldsymbol{v} = [\boldsymbol{\hat{n}}]_{\times} \boldsymbol{v},$$

where $[\hat{n}]_{\times}$ is the matrix form of the cross product operator with the vector $\hat{n} = (\hat{n}_x, \hat{n}_y, \hat{n}_z)$,

$$[\hat{\boldsymbol{n}}]_{\times} = \begin{bmatrix} 0 & -\hat{n}_z & \hat{n}_y \\ \hat{n}_z & 0 & -\hat{n}_x \\ -\hat{n}_y & \hat{n}_x & 0 \end{bmatrix}$$

Note that rotating this vector by another 90° is equivalent to taking the cross product again,

$$oldsymbol{v}_{ imes imes} = oldsymbol{\hat{n}} imes oldsymbol{v}_{ imes} = [oldsymbol{\hat{n}}]^2_{ imes} oldsymbol{v} = -oldsymbol{v}_{ot},$$

and hence

$$oldsymbol{v}_{\parallel} = oldsymbol{v} - oldsymbol{v}_{\perp} = oldsymbol{v} + oldsymbol{v}_{ imes imes} = (oldsymbol{I} + [oldsymbol{\hat{n}}]^2_{ imes})oldsymbol{v}.$$



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We can now compute the in-plane component of the rotated vector \boldsymbol{u} as

$$oldsymbol{u}_{\perp} = \cos heta oldsymbol{v}_{\perp} + \sin heta oldsymbol{v}_{ imes} = (\sin heta [oldsymbol{\hat{n}}]_{ imes} - \cos heta [oldsymbol{\hat{n}}]_{ imes}^2) oldsymbol{v}.$$

Putting all these terms together, we obtain the final rotated vector as

$$oldsymbol{u} = oldsymbol{u}_{\perp} + oldsymbol{v}_{\parallel} = (oldsymbol{I} + \sin heta [oldsymbol{\hat{n}}]_{ imes} + (1 - \cos heta) [oldsymbol{\hat{n}}]_{ imes}^2) oldsymbol{v}.$$

We can therefore write the rotation matrix corresponding to a rotation by θ around an axis \hat{n} as

 $R(\hat{n}, \theta) = I + \sin \theta [\hat{n}]_{\times} + (1 - \cos \theta) [\hat{n}]_{\times}^2$ (Rodriquez' formula)



 $\boldsymbol{R}(\boldsymbol{\hat{n}}, \theta) = \boldsymbol{I} + \sin \theta [\boldsymbol{\hat{n}}]_{\times} + (1 - \cos \theta) [\boldsymbol{\hat{n}}]_{\times}^2$ (Rodriquez' formula)

For small rotations:

$$oldsymbol{R}(oldsymbol{\omega}) pprox oldsymbol{I} + \sin heta [oldsymbol{\hat{n}}]_{ imes} pprox oldsymbol{I} + [heta oldsymbol{\hat{n}}]_{ imes} = egin{bmatrix} 1 & -\omega_z & \omega_y \ \omega_z & 1 & -\omega_x \ -\omega_y & \omega_x & 1 \end{bmatrix}$$





where \hat{n} and θ are the rotation axis and angle.

Rodriguez' formula now becomes (see textbook):

$$\begin{aligned} \boldsymbol{R}(\boldsymbol{\hat{n}}, \theta) &= \boldsymbol{I} + \sin \theta [\boldsymbol{\hat{n}}]_{\times} + (1 - \cos \theta) [\boldsymbol{\hat{n}}]_{\times}^2 \\ &= \boldsymbol{I} + 2w [\boldsymbol{v}]_{\times} + 2 [\boldsymbol{v}]_{\times}^2. \end{aligned}$$



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Quaternion Algebra



$$\boldsymbol{q} = (\boldsymbol{v}, w) = (\sin \frac{\theta}{2} \boldsymbol{\hat{n}}, \cos \frac{\theta}{2})$$

Composition (multiplication): $\boldsymbol{q}_2 = \boldsymbol{q}_0 \boldsymbol{q}_1 = (\boldsymbol{v}_0 \times \boldsymbol{v}_1 + w_0 \boldsymbol{v}_1 + w_1 \boldsymbol{v}_0, w_0 w_1 - \boldsymbol{v}_0 \cdot \boldsymbol{v}_1)$

$$\boldsymbol{R}(\boldsymbol{q}_2) = \boldsymbol{R}(\boldsymbol{q}_0)\boldsymbol{R}(\boldsymbol{q}_1)$$

Inverse: flip the sign of v or w (but not both). i.e., if q = (v, w), then $q^{-1} = (-v, w) = (v, -w)$.







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Orthographic (Parallel) Projection

- Reasonable approximation to perspective projection when % depth variation within field of view is small.
- This is often the case for telephoto lenses (long viewing distances, small field of view)
- Given camera-aligned world coordinate frame, simply drop the z component!
- In inhomogeneous (Euclidean) coordinates:

$$oldsymbol{x} = [oldsymbol{I}_{2 imes 2}|oldsymbol{0}] oldsymbol{p}$$

where p is the 3D point and x is the projected 2D image point

◆ In practice, we also need to scale the x and y coordinates from metres to pixels:

 $\boldsymbol{x} = [s\boldsymbol{I}_{2\times 2}|\boldsymbol{0}] \boldsymbol{p}.$



Perspective Projection



Points projected onto image plane by dividing them by their z component.

$$ar{oldsymbol{x}} = \mathcal{P}_z(oldsymbol{p}) = \left[egin{array}{c} x/z \ y/z \ 1 \end{array}
ight]$$

In homogeneous coordinates:

$$(x_0, y_0, z_0)$$

$$(x, y_1, z)$$

$$(a, b, c)$$

$$ilde{m{x}} = \left[egin{array}{ccccc} 1 & 0 & 0 & 0 \ 0 & 1 & 0 & 0 \ 0 & 0 & 1 & 0 \end{array}
ight] ilde{m{p}},$$



$$\boldsymbol{K} = \begin{bmatrix} f_x & 5 & c_x \\ 0 & f_y & c_y \\ 0 & 0 & 1 \end{bmatrix}$$

 f_x and f_y : encode focal length and pixel spacing, which may be slightly different in *x* and *y* dimensions.

 c_x and c_y : encode principal point (intersection of optic axis with sensor plane) - usually very close to centre of image

s: encodes possible skew between sensor axes (usually close to 0).



Focal Lengths



✤ Focal length can be measured either in pixels or in mm.



Example: Consider the FLIR BlackFly S BFS-PGE-122S6C-C paired with a 10mm lens:

- Resolution: 4096 x 3000 pixels
- Sensor width: 1.1" = 27.94mm







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Lens Distortions



- In perspective projection, straight lines project to straight lines.
- This is not true in real cameras, due to lens distortions.
- Wide-angle lenses produce noticeable radial distortion
- Let (x_c, y_c) be image coordinates after perspective projection but before scaling by focal length and shifting by the optical centre.
- \clubsuit Then without distortion, we should have

$$x_c = \frac{\boldsymbol{r}_x \cdot \boldsymbol{p} + t_x}{\boldsymbol{r}_z \cdot \boldsymbol{p} + t_z}$$
$$y_c = \frac{\boldsymbol{r}_y \cdot \boldsymbol{p} + t_y}{\boldsymbol{r}_z \cdot \boldsymbol{p} + t_z}$$

where r_x , r_y , and r_z are the three rows of R.

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Radial Distortion



- In radial distortion, points are displaced radially by an amount that increases with their distance from the image centre
 - Barrel distortion: points are displaced away from the image centre
 - Pincushion distortion: points are displaced towards the image centre
 - Radial distortion can be modelled by a 4th-order perturbation on these coordinates:

$$\hat{x}_{c} = x_{c}(1 + \kappa_{1}r_{c}^{2} + \kappa_{2}r_{c}^{4})$$
$$\hat{y}_{c} = y_{c}(1 + \kappa_{1}r_{c}^{2} + \kappa_{2}r_{c}^{4}),$$

where
$$r_c^2 = x_c^2 + y_c^2$$





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Pincushion Distortion







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