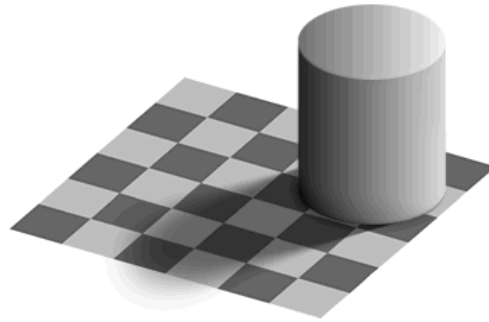


Image Formation



Goal: Introduce the elements of camera models, optics, and image formation.

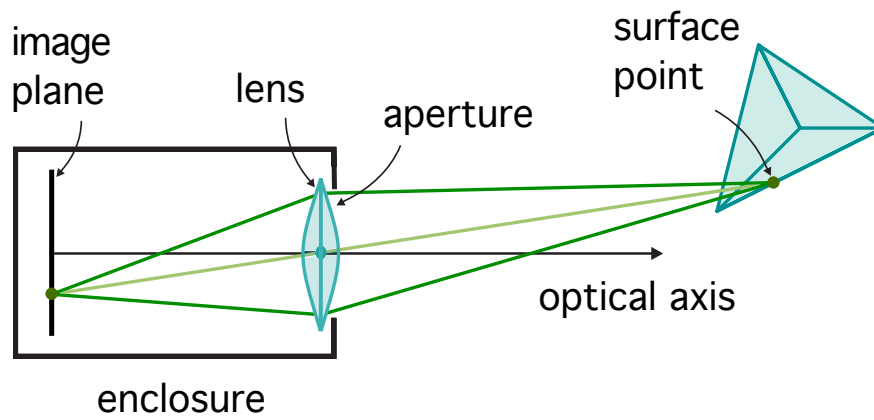
Motivation: Camera models, together with radiometry and reflectance models, allow us to formulate the dependence of image color/intensity on material reflectance, surface geometry, and lighting.

- Based on these models, we can formulate the inference of scene properties such as surface shape, reflectance, and scene lighting, from image data.
- Here we consider the “forward” model. That is, assuming various scene and camera properties, what should we observe in an image?

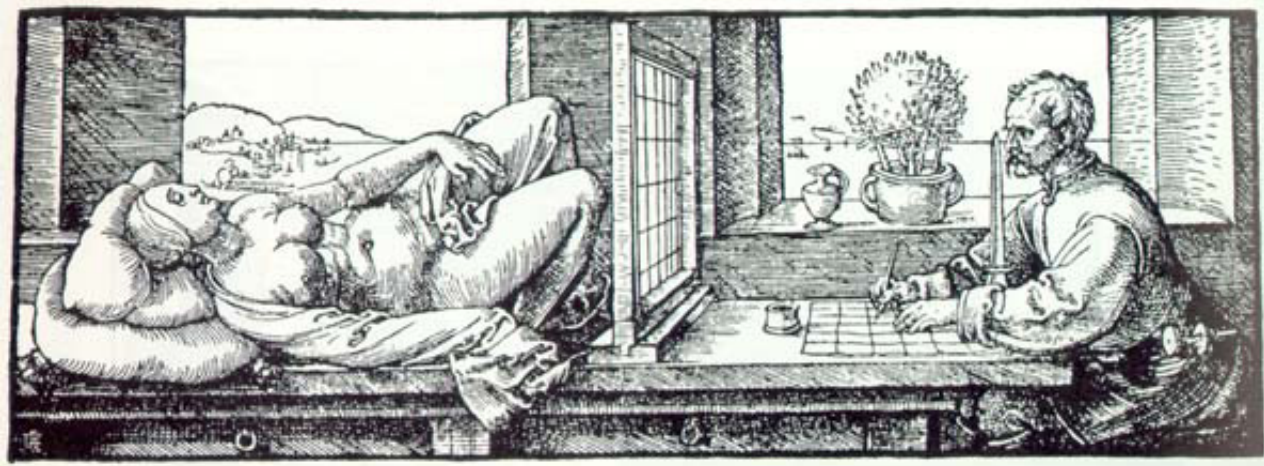
Readings: Part I (Image Formation and Image Models) of Forsyth and Ponce.

Matlab Tutorials: colourTutorial.m (in UTVis)

Camera Elements

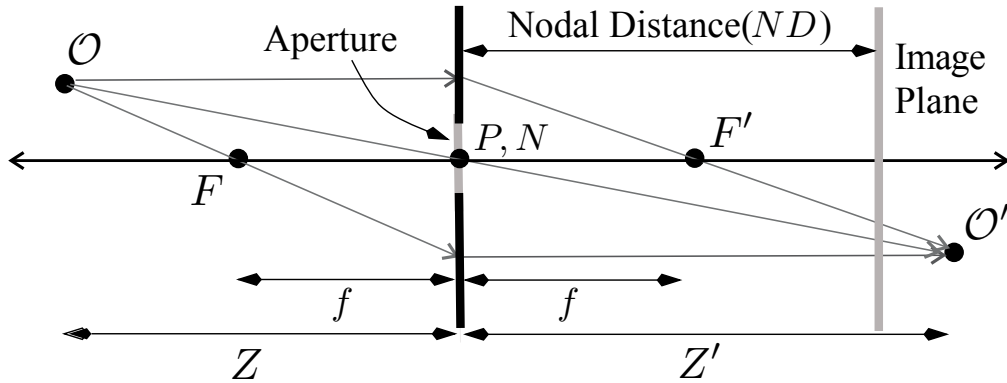


Albrecht Durer (1471-1528):



Thin Lens

Suppose we neglect the thickness of the lens and assume the medium (e.g. air) is the same on both sides of the lens:



A world point \vec{O} is focussed at \vec{O}' , at the intersection of three rays:

- A ray from \vec{O} through N (the nodal point, or center of projection)
- Two rays, parallel to the optical axis, one in front of and one behind the lens (aka principal plane), which pass through the rear/front focal points (F' and F) on the opposite side of the lens.

Remarks:

- Lens *Aperture* is modeled as an occluder (in the principal plane)
- The point \vec{O} is not in focus on the image plane, but rather at \vec{O}' .
- Thin lens model:

$$\frac{1}{f} = \frac{1}{Z} + \frac{1}{Z'}$$

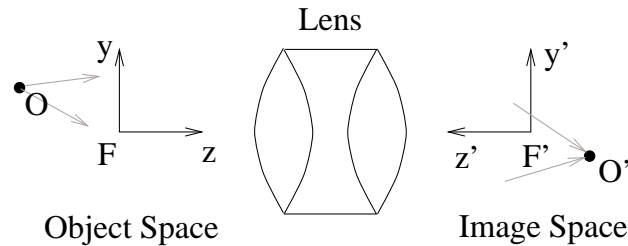
- The *focal length* f is the distance of the point behind the lens, F' , to which rays of light from infinity converge.

General Lens Model

Given a general lens, a point at \vec{O} is imaged to \vec{O}' , where the locations of \vec{O} and \vec{O}' are given by the **lens formula**:

$$\vec{O}' \equiv \begin{pmatrix} z' \\ y' \end{pmatrix} = \frac{f}{z} \begin{pmatrix} f' \\ y \end{pmatrix}, \quad \vec{O} \equiv \begin{pmatrix} z \\ y \end{pmatrix} = \frac{f'}{z'} \begin{pmatrix} f \\ y' \end{pmatrix}.$$

Here F, F' are focal points, and f, f' are focal lengths.



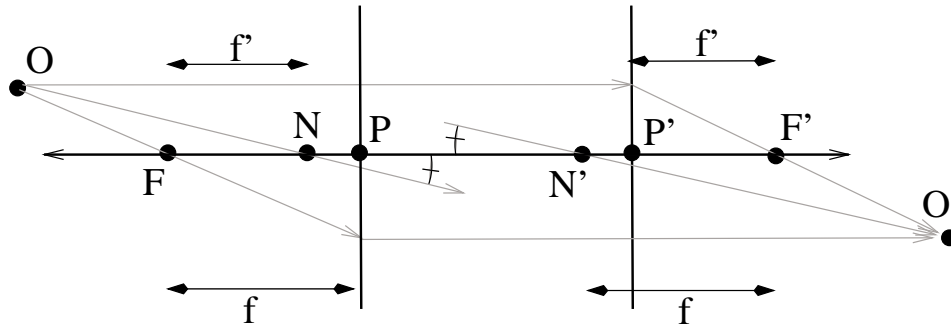
In general the ratio of focal lengths equals the ratio of the indices of refraction of the pre- and post-lens material, that is $f/f' = n/n'$ (eg. $f \neq f'$ for eyes, but $f = f'$ for most cameras). The index of refraction of a material is the ratio of the speed of light in a vacuum over the speed of light in that medium.

As for a thin lens, the formation of the image of \vec{O} can be interpreted geometrically as the intersection of three canonical rays, which are determined by the **cardinal points** of the lens. The cardinal points are:

Focal Points F, F' provide origins for the object and image spaces.

Nodal Points N, N' , are defined using the lens axis, F, F' , and focal lengths, f, f' .

Principal Points P, P' are also defined using the lens axis, F, F' , and focal lengths, f, f' .



Lens Formula

An alternative coordinate system which is sometimes used to write the lens formula is to place the origins of the coordinates in the object and image space at the principal points P and P' , and flip both the z -axes to be pointing away from the lens. These new z -coordinates are:

$$\hat{z} = f - z,$$
$$\hat{z}' = f' - z'.$$

Solving for z and z' and substituting into the previous lens formula, we obtain:

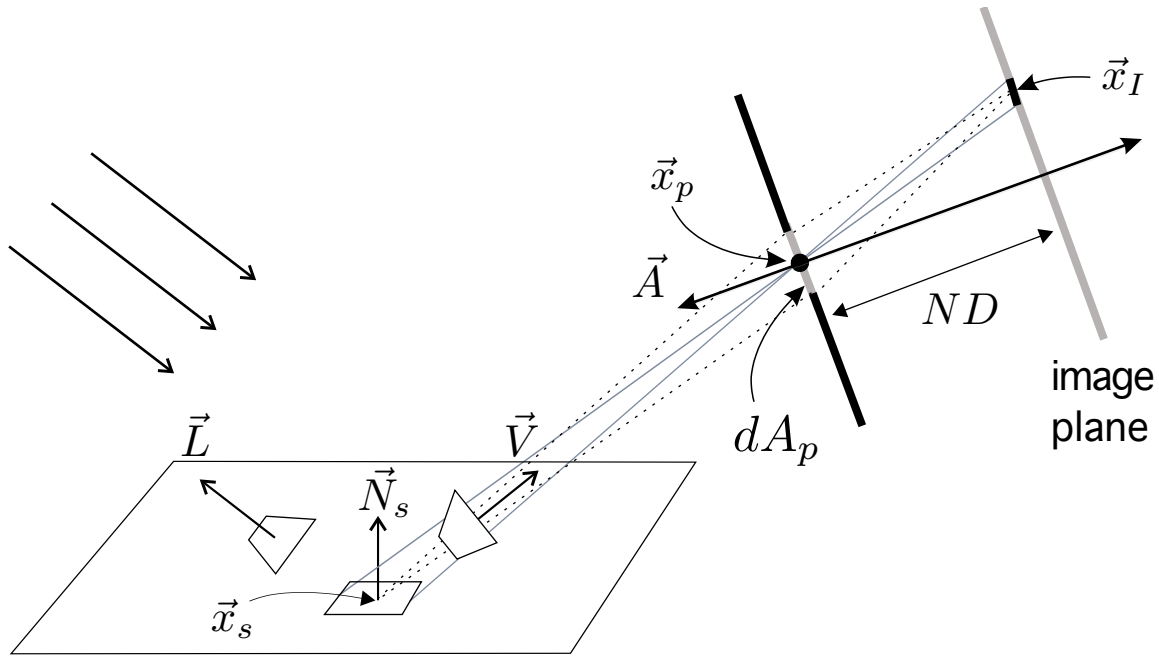
$$(f' - \hat{z}) = f f' / (f - \hat{z}),$$
$$f f' = (f' - \hat{z}')(f - \hat{z})$$
$$\hat{z}' \hat{z} = \hat{z}' f + \hat{z} f'$$
$$1 = \frac{f}{\hat{z}} + \frac{f'}{\hat{z}'}$$

The last line above is also known as the lens formula. As we have seen, it is equivalent to the one on the previous page, only with a change in the definition of the coordinates.

For cameras with air both in front of and behind the lens, we have $f = f'$. This simplifies the lens formula above. Moreover, the nodal and principal points coincide in both the object and scene spaces (i.e., $N = P$ and $N' = P'$ in the previous figure).

Finally it is worth noting that, in terms of image formation, the difference between this general lens model and the thin lens approximation is only in the displacement of the cardinal points along the optical axis. That is, effectively, the change in the imaging geometry from a thin lens model to the general lens model is simply the introduction of an absolute displacement in the image space coordinates. For the purpose of modelling the image for a given scene, we can safely ignore this displacement and use the thin lens model. When we talk about the center of projection of a camera in a world coordinate frame, however, it should be understood we are talking about the location of the nodal point N in the object space (and not N' in the image space). Similarly, when we talk about the nodal distance to the image plane, we mean the distance from N' to the image plane.

Image of a Lambertian Surface



The irradiance on the image plane is

$$I(\lambda, \vec{x}_I) = T_l \frac{d\Omega_p dA_V}{dA_I} r(\lambda) I(\lambda, \vec{x}_s)$$

Here

- \vec{N}_I is normal to the image plane, and \vec{A} is the optical axis;
- $T_l \in (0, 1]$ is the transmittance of the lens;
- dA_I is the area of each pixel;
- dA_p is the area of the aperture;
- $d\Omega_p$ is the solid angle of the aperture from the surface point \vec{x}_s ;
- dA_V is the cross-sectional area, perpendicular to the viewing direction, of the portion of the surface imaged to the pixel at \vec{x}_I .

Derivation of the Image of a Lambertian Surface

From our notes on Lambertian reflection, the radiance (spectral density) of the surface is

$$R(\lambda, \vec{x}_s; \vec{V}) = r(\lambda) I(\lambda, \vec{x}_s; \vec{N}_s) = r(\lambda) [\vec{N} \cdot \vec{L}] I(\lambda, \vec{x}_s; \vec{L}) .$$

The reflected radiance is measured in Watts per unit wavelength, per unit cross-sectional area perpendicular to the viewer, per unit steradian.

The total power (per unit wavelength) from the patch dA_V , arriving on the aperature, is therefore

$$P(\lambda) = R(\lambda, \vec{x}_s; \vec{V}) d\Omega_p dA_V .$$

A fraction T_l of this is transmitted through the lens, and ends up on a pixel of area dA_I . Therefore, the pixel irradiance spectral density is

$$I(\lambda, \vec{x}_I, \vec{n}_I) = \frac{T_l P(\lambda)}{dA_I} ,$$

which is the expression on the previous page.

To simplify this, first compute the solid angle of the lens aperature, with respect to the surface point \vec{x}_s . Given the area of the aperature, dA_p , and the optical axis, \vec{A} , which is assumed to be perpendicular to the aperature, we have

$$d\Omega_p = \frac{|\vec{V} \cdot \vec{A}| dA_p}{\|\vec{x}_p - \vec{x}_s\|^2} .$$

Here the numerator is the cross-sectional area of the aperature viewed from the direction \vec{V} . The denominator scales this foreshortened patch back to the unit sphere to provide the desired measure of solid angle. Secondly, we need the foreshortened surface area dA_V which projects to the individual pixel at \vec{x}_I having area dA_I . These two patches are related by rays passing through the center of projection \vec{x}_p ; they have the same solid angle with respect to \vec{x}_p . As a result,

$$dA_V = \|\vec{x}_p - \vec{x}_s\|^2 \frac{|\vec{V} \cdot \vec{A}| dA_I}{\|\vec{x}_p - \vec{x}_I\|^2}$$

The distance in the denominator here can be replaced by

$$\|\vec{x}_p - \vec{x}_I\| = \frac{ND}{|\vec{V} \cdot \vec{A}|} .$$

Substituting these expressions for $d\Omega_p$, dA_V , and $\|\vec{x}_p - \vec{x}_I\|$ gives the equation for the image irradiance due to a Lambertian surface on the following page.

Image of a Lambertian Surface (cont.)

This expression for the irradiance due to a Lambertian surface simplifies to

$$I(\lambda, \vec{x}_I; \vec{N}_I) = T_l \frac{dA_p}{|ND|^2} |\vec{A} \cdot \vec{V}|^4 r(\lambda) [\vec{N} \cdot \vec{L}] I(\lambda, \vec{x}_s; \vec{L})$$

where dA_p is the area of the aperture.

Note that image irradiance

- does not depend on the distance to the surface $\|\vec{x}_s - \vec{x}_p\|$ (as the distance to the surface increases, the surface area "seen" by a pixel also increases to compensate for the distance change);
- falls off like $\cos(\theta)^4$ in the corners of the image where θ is the angle between the viewing direction \vec{V} and the optical axis \vec{A} . For wide angle images, there is a significant roll-off in the image intensity towards the corners.

The fall off of the brightness in the corners of the image is called **vignetting**. The actual vignetting obtained depends on the internal structure of the lens, and will deviate from the above $\cos(\theta)^4$ term.

Image Irradiance to Absorbed Energy

A pixel response is a function of the energy absorbed by that pixel (i.e., the integral of irradiance over pixel area, the duration of the shutter opening, and the pixel *spectral sensitivity*)

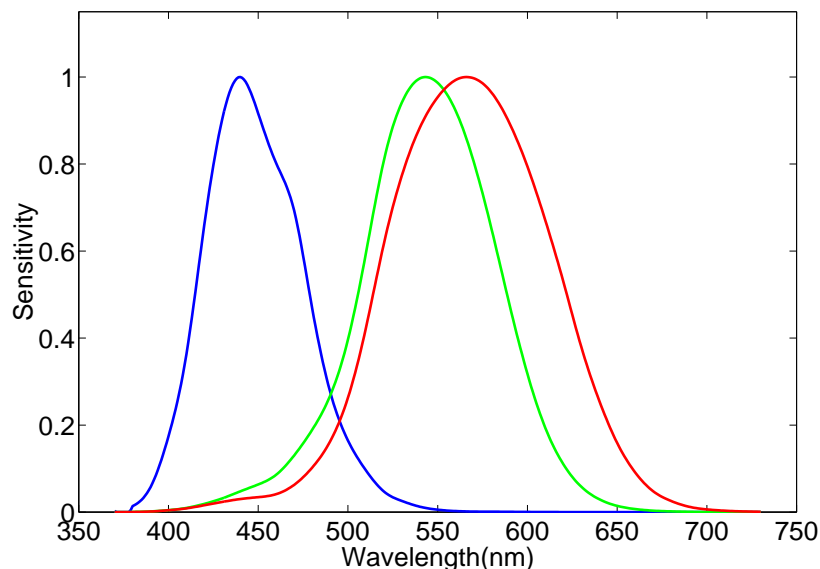
For a steady image, not changing in time, the absorbed energy at pixel \vec{x}_I can be approximated by

$$e_{\mu}(\vec{x}_I) = C_T A_I \int_0^{\infty} S_{\mu}(\lambda) I(\lambda, \vec{x}_I) d\lambda .$$

Here $I(\lambda, \vec{x}_I)$ is the image irradiance, $S_{\mu}(\lambda)$ is the spectral sensitivity of the μ^{th} colour sensor, A_I is the area of the pixel, and C_T is the temporal integration time (eg. 1/(shutter speed)).

Colour images are formed (typically) using three spectral sensitivities, say $\mu = R, G, B$ for the ‘red’, ‘green’ and ‘blue’ channel.

Normalized spectral sensitivities in the human retina:



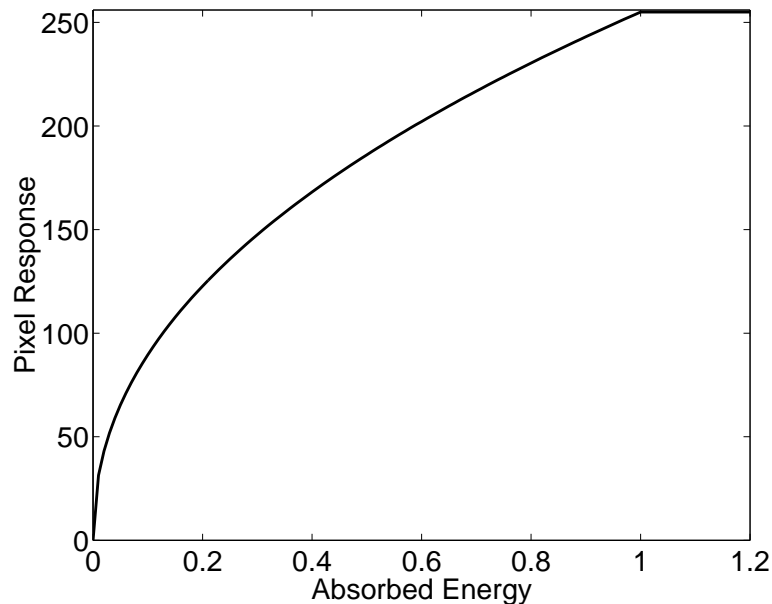
Absorbed Energy to Pixel Response

Gamma Correction. Finally, the absorbed energy e_μ is converted to a quantized pixel response, say r_μ , through a nonlinear function called a gamma correction, for example,

$$r_\mu = \beta [e_\mu]^\gamma .$$

The value of γ can vary; values between 2 and 3 are common.

This response r_μ is quantized, typically to 8 bits.

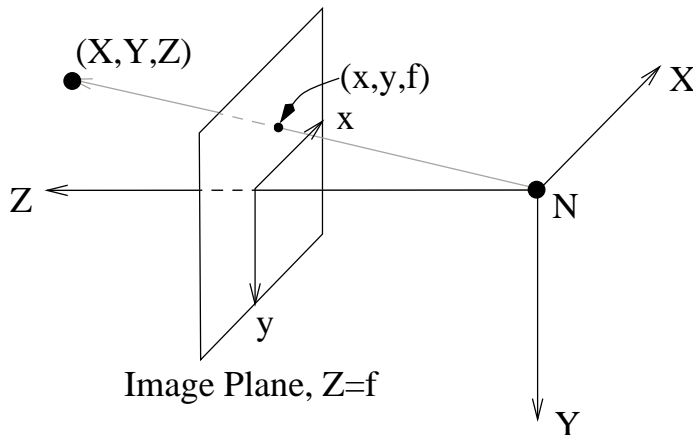


This completes the basic scene and image formation models.

Next we consider approximations and simplifications of the model.

Perspective Projection (The Pinhole Camera)

The image formation of both thick and thin lenses can be approximated with a pinhole camera,



The image position for a 3D point (X, Y, Z) is given by *perspective projection*:

$$\begin{pmatrix} x \\ y \\ f \end{pmatrix} = \frac{f}{Z} \begin{pmatrix} X \\ Y \\ Z \end{pmatrix}$$

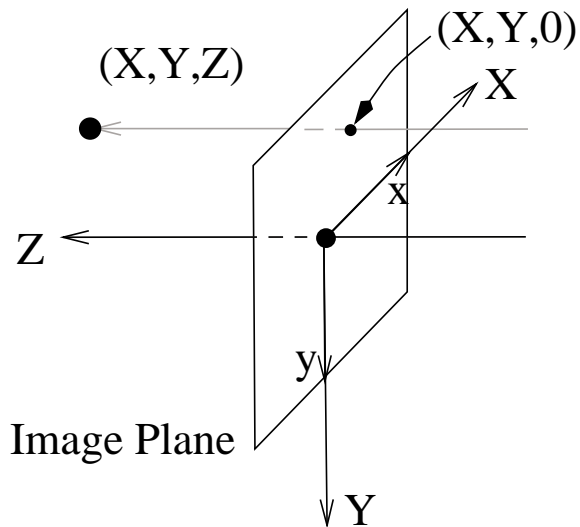
By convention, the nodal distance $|ND|$ is labelled f , the *focal length*. The key approximation here is that all depths are taken to be in focus.

Remarks:

- We put the image plane in front of the nodal point to avoid flipping the image about the origin (for mathematical convenience).
- image coordinate x increase to the right, and y downwards.

Orthographic Projection

Alternative projections onto an image plane are given by orthographic projection and scaled orthographic projection.



Given a 3D point (X, Y, Z) , the corresponding image location under scaled orthographic projection is

$$\begin{pmatrix} x \\ y \end{pmatrix} = s_0 \begin{pmatrix} X \\ Y \end{pmatrix}$$

Here s_0 is a constant scale factor; orthographic projection uses $s_0 = 1$.

Scaled orthographic projection provides a linear approximation to perspective projection, which is applicable for a small object far from the viewer and close to the optical axis.

Coordinate Frames

Consider the three coordinate frames:

- a world coordinate frame \vec{X}_w ,
- a camera coordinate frame, \vec{X}_c ,
- an image coordinate frame, \vec{p} .

The world and camera frames provide standard 3D orthogonal coordinates. The image coordinates are written as a 3-vector, $\vec{p} = (p_1, p_2, 1)^T$, with p_1 and p_2 the pixel coordinates of the image point.

Camera Coordinate Frame. The origin of the camera coordinates is at the nodal point of the camera (say at \vec{d}_w in world coords). The z -axis is taken to be the optical axis of the camera (with points in front of the camera having a positive z value).

Image formation requires that we specify two mappings, i.e., from *world to camera* coordinates, and from *camera to image* coordinates.

External Calibration Matrix

The external calibration parameters specify the 3D coordinate transformation from world to camera:

$$\vec{X}_c = R(\vec{X}_w - \vec{d}_w) , \quad \vec{X}_w = R^T \vec{X}_c + \vec{d}_w ,$$

where R is a 3×3 rotation matrix, and \vec{d}_w is the location of the nodal point for the camera, in world coordinates.

It is common to write the external (aka extrinsic) calibration parameters in terms of a 3×4 matrix M_{ex} :

$$\vec{X}_c = M_{ex} \begin{pmatrix} \vec{X}_w \\ 1 \end{pmatrix} \quad (1)$$

where

$$M_{ex} = \begin{pmatrix} R & -R\vec{d}_w \end{pmatrix} . \quad (2)$$

In the camera coordinate frame, the perspective transformation of the 3D point \vec{X}_c (in the camera's coordinates) to the image plane is

$$\vec{x}_c = \frac{f}{X_{3,c}} \vec{X}_c = \begin{pmatrix} x_{1,c} \\ x_{2,c} \\ f \end{pmatrix} . \quad (3)$$

Here f is the nodal distance (focal length) for the camera.

Internal Calibration Matrix

The internal (aka intrinsic) calibration matrix transforms the 3D image position \vec{x}_c to pixel coordinates,

$$\vec{p} = M_{in} \vec{x}_c, \quad (4)$$

where M_{in} is a 3×3 matrix.

For example, a camera with rectangular pixels of size $1/s_x$ by $1/s_y$, with focal length f , and piercing point (o_x, o_y) (i.e., the intersection of the optical axis with the image plane, in pixel coordinates) has

$$M_{in} = \begin{pmatrix} s_x & 0 & o_x/f \\ 0 & s_y & o_y/f \\ 0 & 0 & 1/f \end{pmatrix}. \quad (5)$$

Note that for a 3D point \vec{x}_c on the image plane, the third coordinate of the pixel coordinate vector, \vec{p} , is $p_3 = 1$. As we see next, this redundancy is useful.

Together, Eqns (1), (3) and (4) define the mapping from a 3D point in world coordinates, \vec{X}_w , to \vec{p} , the pixel coordinates of the image of \vec{X}_w . The mapping is nonlinear, due to the scaling by $X_{3,c}$ in Eqn (3).

Homogeneous Coordinates

It is useful to express this transformation in terms of homogeneous coordinates,

$$\vec{X}_w^h = a \begin{pmatrix} \vec{X}_w \\ 1 \end{pmatrix}, \quad \vec{p}^h = b \vec{p} = b \begin{pmatrix} p_1 \\ p_2 \\ 1 \end{pmatrix},$$

for arbitrary nonzero constants a, b . The last element of a homogeneous vector provides the scale factor which allows one to convert back and forth between the homogeneous form and the standard form.

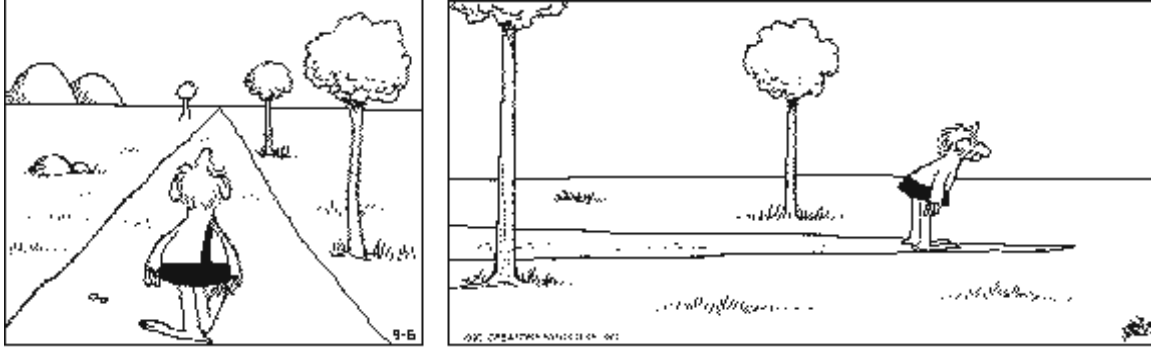
In homogeneous coordinates the mapping from world to pixel coordinates is a *linear* transformation,

$$\vec{p}^h = M_{in} M_{ex} \vec{X}_w^h. \quad (6)$$

The division operation in perspective projection is now implicit in the homogeneous vector \vec{p}^h . It is simply postponed until \vec{p}^h is rescaled by its third coordinate to form the pixel coordinate vector \vec{p} .

Due to its linearity, Eqn (6) is useful in formulating many computer vision problems.

Parallel Lines Project to Intersecting Lines



As an application of (6), consider a set of parallel lines in 3D, say

$$\vec{X}_k^h(s) = \begin{pmatrix} \vec{X}_k^0 \\ 1 \end{pmatrix} + s \begin{pmatrix} \vec{t} \\ 0 \end{pmatrix}.$$

Here \vec{X}_k^0 , for $k = 1, \dots, K$, and \vec{t} are 3D vectors in the world coordinate frame. Here \vec{t} is the common 3D tangent direction for all the lines, and \vec{X}_k^0 is an arbitrary point on the k^{th} line.

Then, according to equation (6), the images of these points in homogeneous coordinates are given by

$$\vec{p}_k^h(s) = M\vec{X}_k^h(s) = \vec{p}_k^h(0) + s\vec{p}_t^h,$$

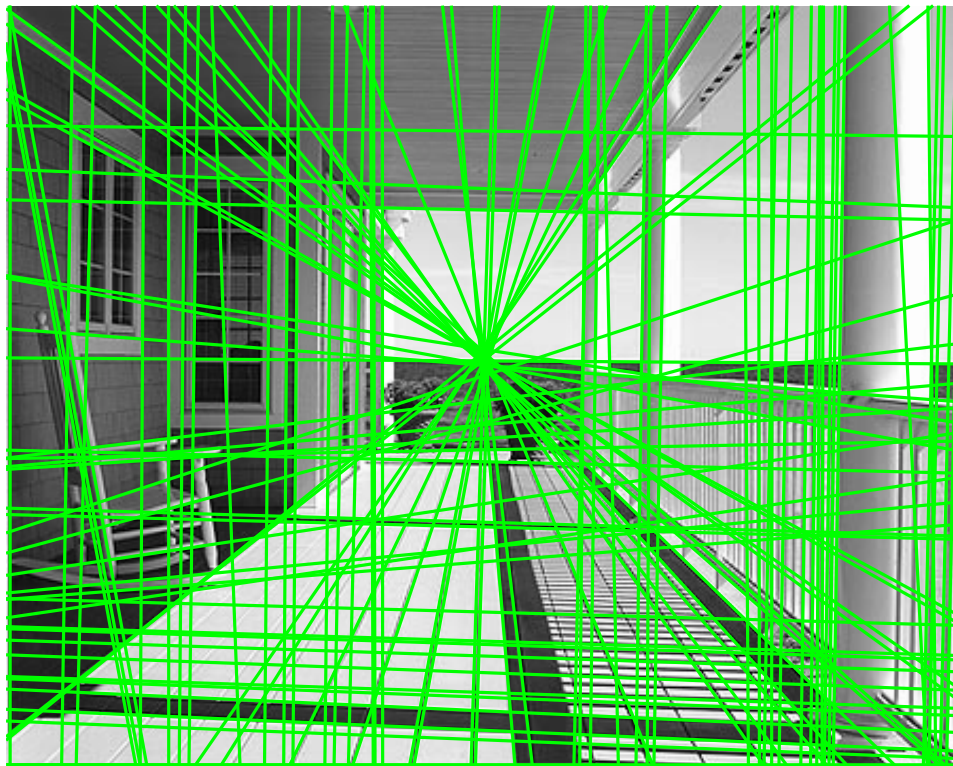
where $M = M_{in}M_{ex}$ is a 3×4 matrix, $\vec{p}_t^h = M(\vec{t}^T, 0)^T$ and $\vec{p}_k^h(0) = M((\vec{X}_k^0)^T, 1)^T$. Note \vec{p}_t^h and $\vec{p}_k^h(0)$ are both constant vectors, independent of s . Converting to standard pixel coordinates, we have

$$\vec{p}_k(s) = \frac{1}{\alpha(s)}\vec{p}_k^h(0) + \frac{s}{\alpha(s)}\vec{p}_t^h,$$

where $\alpha(s) = p_{k,3}^h(s)$ is third component of $\vec{p}_k^h(s)$. Therefore we have shown $\vec{p}_k(s)$ is in the subspace spanned by two constant 3D vectors. It is also in the image plane, $p_{k,3} = 1$. Therefore it is in the intersection of these two planes, which is a line in the image. That is, lines in 3D are imaged as lines in 2D. (Although, in practice, some lenses introduce “radial distortion”, which causes the image of a 3D line to be bent. However, this distortion can be removed with careful calibration.)

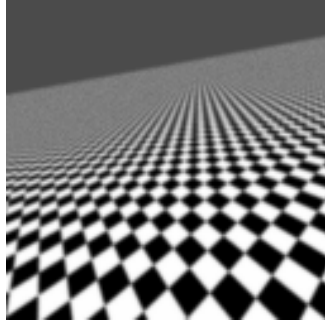
In addition it follows that $\alpha(s) = p_{k,3}^h(0) + \beta s$ where $\beta = p_{t,3}^h = (0, 0, 1)M(\vec{t}^T, 0)^T$. Assuming $\beta \neq 0$, we have $1/\alpha(s) \rightarrow 0$ and $s/\alpha(s) \rightarrow 1/\beta$ as $s \rightarrow \infty$. Therefore the image points $\vec{p}_k(s) \rightarrow (1/\beta)\vec{p}_t^h$, which is a constant image point dependent only on the tangent direction of the 3D lines. This shows that the images of the parallel 3D lines $\vec{X}_k^h(s)$ all intersect at the image point $(1/\beta)\vec{p}_t^h$.

Example of Parallel Lines



The Horizon Line

Suppose the parallel lines discussed above are coplanar. Then analysis above shows that the images of such lines intersect at the horizon (i.e., the image of points on the plane infinitely far from the camera). This property is depicted in the left panel of the previous cartoon. As another exercise in projective geometry, let's show that the horizon of a planar surface is a straight line in the image.



Consider multiple families of parallel lines in the plane. Let the j^{th} family have the tangent direction \vec{t}_j in 3D. From the previous analysis, the j^{th} family must co-intersect at the image point (in homogeneous coordinates)

$$\vec{p}_j^h = M \begin{pmatrix} \vec{t}_j \\ 0 \end{pmatrix},$$

and these points \vec{p}_j^h must be on the horizon.

Because the tangent directions are coplanar in 3D, two distinct directions provide a basis. That is, assuming the first two directions are linearly independent, we can write

$$\vec{t}_j = a_j \vec{t}_1 + b_j \vec{t}_2,$$

for some constants a_j and b_j . As a consequence, follows that

$$\vec{p}_j^h = M \begin{pmatrix} a_j \vec{t}_1 + b_j \vec{t}_2 \\ 0 \end{pmatrix} = a_j \vec{p}_1^h + b_j \vec{p}_2^h$$

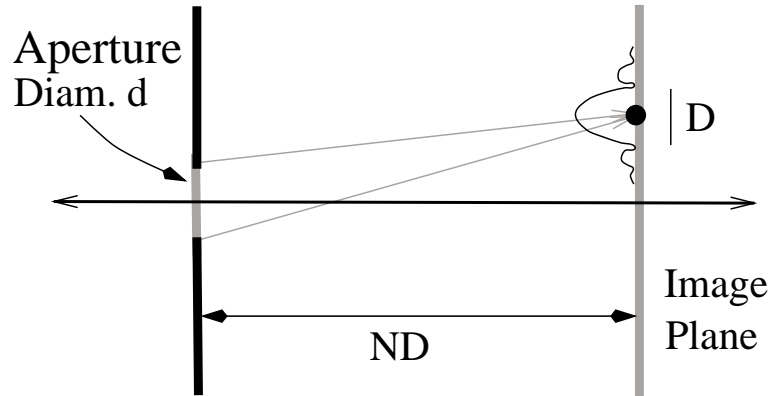
Dividing through by the third coordinate, $p_{j,3}^h$, we find the points of intersection of the j^{th} family of lines is at the image point

$$\vec{p}_j = \left(\frac{1}{p_{j,3}^h} \right) \vec{p}_j^h = \left(\frac{a_j p_{1,3}^h}{p_{j,3}^h} \right) \vec{p}_1 + \left(\frac{b_j p_{2,3}^h}{p_{j,3}^h} \right) \vec{p}_2 = \alpha_j \vec{p}_1 + \beta_j \vec{p}_2.$$

From the third coefficient of this equation it follows that $\alpha_j + \beta_j = 1$. Hence the image point \vec{p}_j is an affine combination of the image points \vec{p}_1 and \vec{p}_2 (e.g., we can write $\vec{p}_j = \vec{p}_1 + \beta_j (\vec{p}_2 - \vec{p}_1)$). Therefore the horizon must be the line in the image passing through \vec{p}_1 and \vec{p}_2 .

Physical Limitations to Image Resolution

1. Diffraction



Even a properly focussed point is *not* imaged to a point. Rather, there is a *point spread function* (PSF).

For diffraction alone, this PSF can be modelled using the ‘Airy disk’, which has diameter

$$D \approx \frac{1.22\lambda}{n'} \frac{|ND|}{d},$$

where d is the aperture diameter and n' is the index of refraction inside the eye. Lens imperfections and imperfect focusing lead to larger blur diameters.

Diffraction Limit (cont.)

For human eyes (see Wyszecki & Stiles, Color Science, 1982):

- the index of refraction within the eye is $n' = 1.33$;
- the nodal distance is $|ND| \approx 16.7mm$ (accommodated at ∞);
- the pupil diameter is $d \approx 2mm$ (adapted to bright conditions);
- a typical wavelength is $\lambda \approx 500nm$.

Therefore the diameter of the Airy disk is

$$D \approx 4\mu = 4 \times 10^{-6}m$$

This compares closely to the diameter of a foveal cone (i.e. the smallest pixel), which is between 1μ and 4μ . So, human vision operates at the diffraction limit.

By the way, a 2μ pixel spacing in the human eye corresponds to having a 300×300 pixel resolution of the image of your thumbnail at arm's length. Compare this to the typical sizes of images used by machine vision systems, usually about 1000×1000 .

2. Photon Noise

The average photon flux (spectral density) at the image (in units of photons per sec, per unit wavelength, per image area) is

$$I(\lambda, \vec{x}_I; \vec{N}_I) \frac{\lambda}{\hbar c}$$

Here \hbar is Planck's constant and c is the speed of light.

The photon arrivals can be modelled with Poisson statistics, so the variance is equal to the mean photon catch.

Even in bright conditions, foveal cones have a significant photon noise component (a std. dev. $\approx 10\%$ of the signal, for unshaded scenes).

3. Defocus

An improperly focussed lens causes the PSF to broaden. Geometrical optics can be used to get a rough estimate of the size.

4. Motion Blur

Given temporal averaging, the image of a moving point forms a streak in the image, causing further blur.

Conclude: There is a limit to how small standard cameras and eyes can be made (but note multi-faceted insect eyes). Human vision operates close to the physical limits of resolution (ditto for insects).