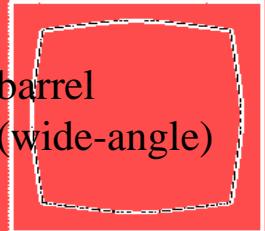
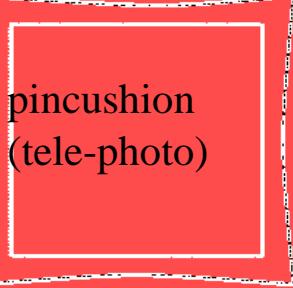




Radial Distortion

magnification/focal length different
for different angles of inclination

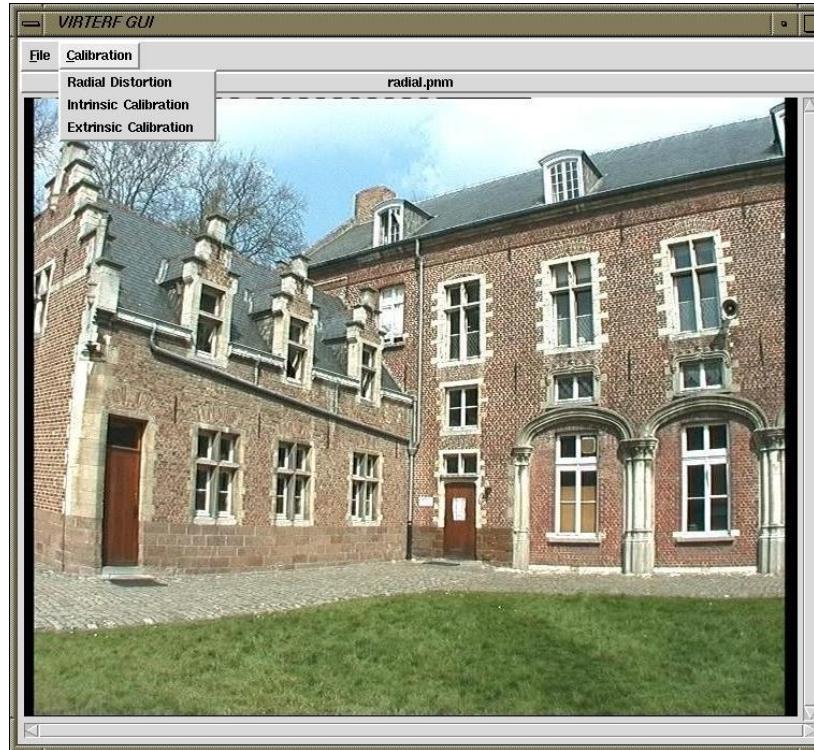


Can be corrected! (if parameters are known)



Radial Distortion

magnification/focal length different
for different angles of inclination



pincushion
(tele-photo)

barrel
(wide-angle)



Radial Distortion

magnification/focal length different
for different angles of inclination



pincushion
(tele-photo)

barrel
(wide-angle)

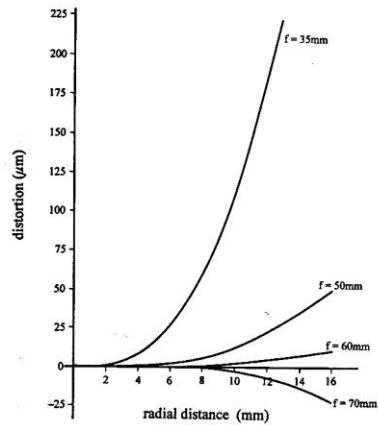
Can be corrected! (if parameters are known)



Radial Distortion



straight lines are not straight anymore



barrel dist.

pincushion dist.

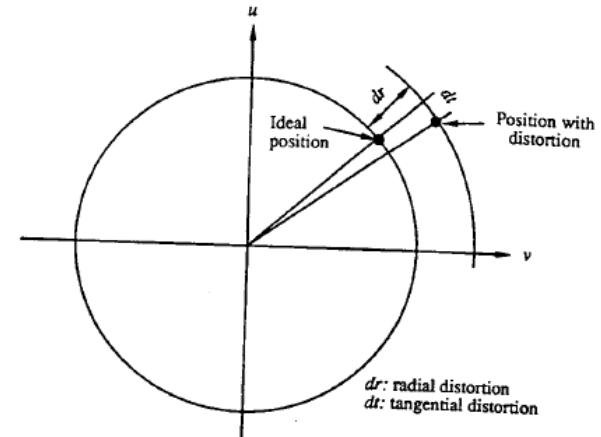


Fig. 2. Radial and tangential distortions.

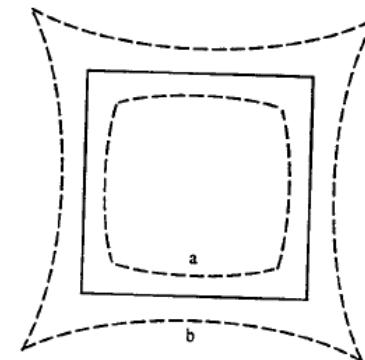


Fig. 3. Effect of radial distortion. Solid lines: no distortion; dashed lines: with radial distortion (a: negative, b: positive).



Radial distortion

- Due to spherical lenses (cheap/wide angle)
- Model:(following Tsai 1987 et al.):

$$\vec{p} = \boxed{R^{-1}} * \frac{1}{z} K \begin{pmatrix} {}^C {}_W R & {}^C \vec{t} \\ 0,0,0 & 1 \end{pmatrix} {}^W \vec{p}$$

$$\mathbf{R}(x, y) = (1 + K_1(x^2 + y^2) + K_2(x^4 + y^4) + \dots) \begin{bmatrix} x^{rad} \\ y^{rad} \end{bmatrix}$$

$$\mathbf{p} = \begin{pmatrix} 1/\lambda & 0 & 0 \\ 0 & 1/\lambda & 0 \\ 0 & 0 & 1 \end{pmatrix} \mathcal{M} \mathbf{P}$$

λ is a polynomial function of $\hat{r}^2 \stackrel{\text{def}}{=} \hat{u}^2 + \hat{v}^2$, i.e., $\lambda = 1 + \kappa_1 \hat{r}^2 + \kappa_2 \hat{r}^4 + \dots$

Radial distortion example





Radial distortion example



Radial distortion example





3.3.1 Estimation of Projection Matrix

Geometrically, radial distortion changes the distance between the image center and the image point \mathbf{p} but it does not affect the direction of the vector joining these two points. This is called the *radial alignment constraint* by Tsai, and it can be expressed algebraically by writing

$$\lambda \begin{pmatrix} u \\ v \end{pmatrix} = \begin{pmatrix} \frac{\mathbf{m}_1 \cdot \mathbf{P}}{\mathbf{m}_3 \cdot \mathbf{P}} \\ \frac{\mathbf{m}_2 \cdot \mathbf{P}}{\mathbf{m}_3 \cdot \mathbf{P}} \end{pmatrix} \implies v(\mathbf{m}_1 \cdot \mathbf{P}) - u(\mathbf{m}_2 \cdot \mathbf{P}) = 0.$$

This is a linear constraint on the vectors \mathbf{m}_1 and \mathbf{m}_2 . Given n fiducial points we obtain n equations in the eight coefficients of the vectors \mathbf{m}_1 and \mathbf{m}_2 , namely

$$\mathcal{Q}\mathbf{n} = 0, \quad \text{where} \quad \mathcal{Q} \stackrel{\text{def}}{=} \begin{pmatrix} v_1 \mathbf{P}_1^T & -u_1 \mathbf{P}_1^T \\ \dots & \dots \\ v_n \mathbf{P}_n^T & -u_n \mathbf{P}_n^T \end{pmatrix} \quad \text{and} \quad \mathbf{n} = \begin{pmatrix} \mathbf{m}_1 \\ \mathbf{m}_2 \end{pmatrix}. \quad (6.3.9)$$

Note the similarity with the previous case. When $n \geq 8$, the system of equations (6.3.9) is in general overconstrained, and a solution with unit norm can be found using linear least squares.



Useful Links

Demo calibration (some links broken):

- <http://mitpress.mit.edu/e-journals/Videre/001/articles/Zhang/CalibEnv/CalibEnv.html>

Bouget camera calibration SW:

- http://www.vision.caltech.edu/bouguetj/calib_doc/

CVonline: Monocular Camera calibration:

- <http://homepages.inf.ed.ac.uk/cgi/rbf/CVONLINE/entries.pl?TAG250>