

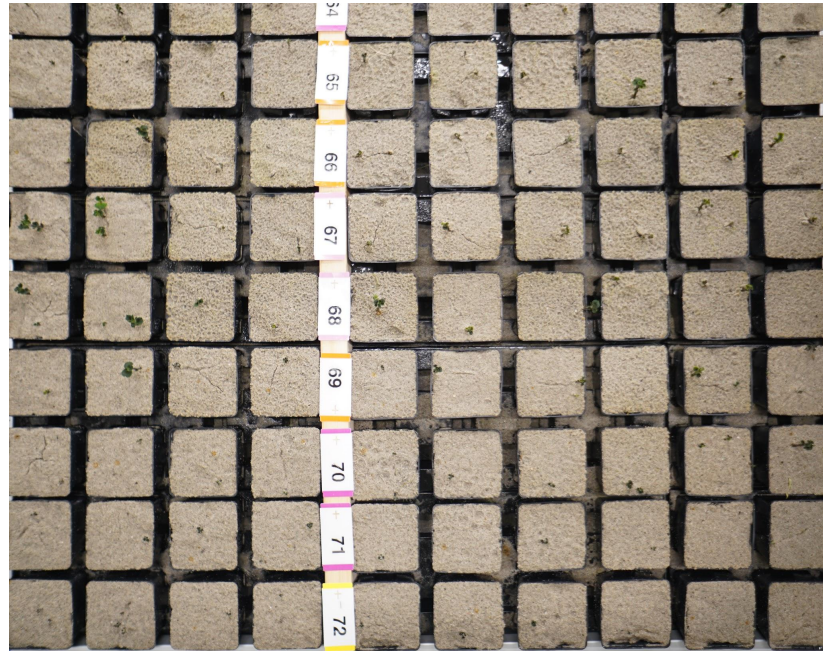
Modelling and imaging of growing plants

Merel te Hofsté

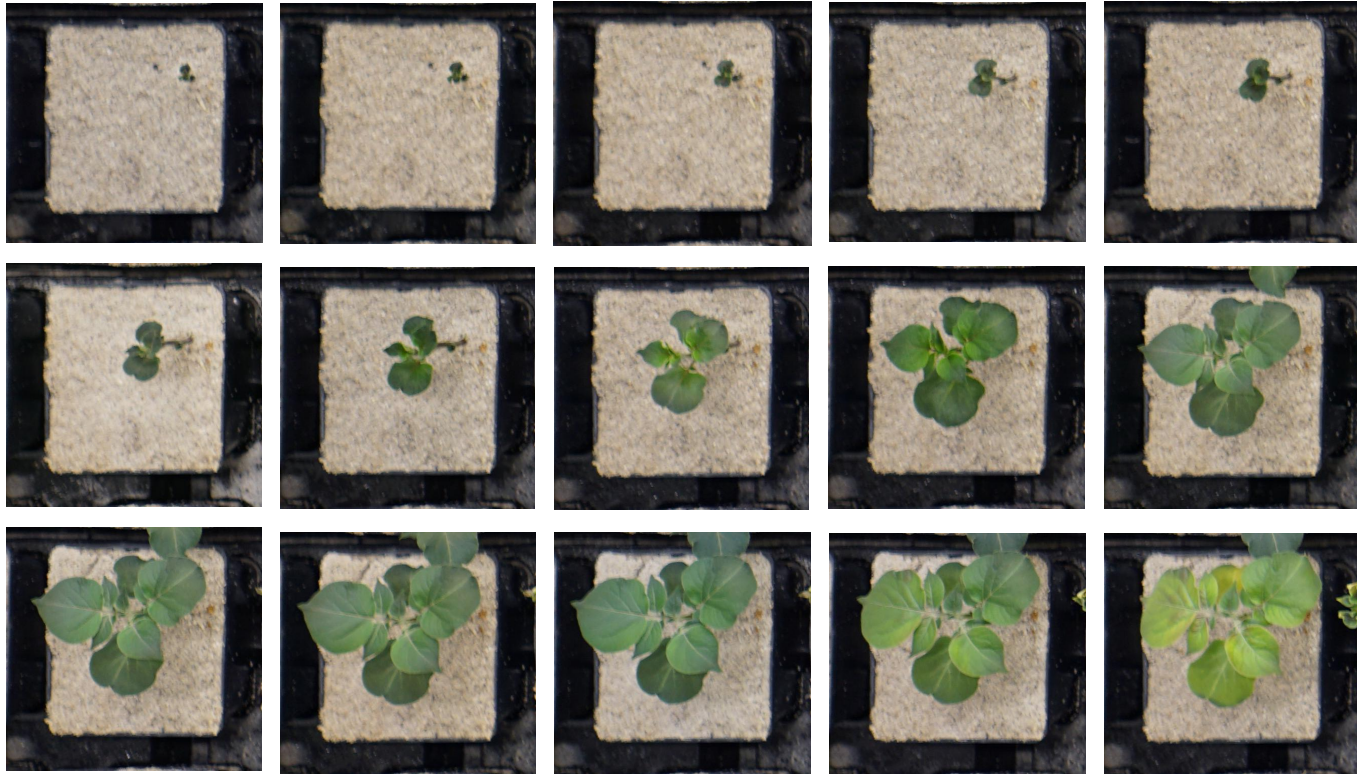
Daily supervisor: N.V. Budko

Problem

- The problem of quantifying the growth dynamics of plants from time-lapse images is important for determining variety-specific characteristics and subsequent breeding.
- Amount of data:
 - Six potato varieties
 - Two climate rooms
 - Dry and wet sections

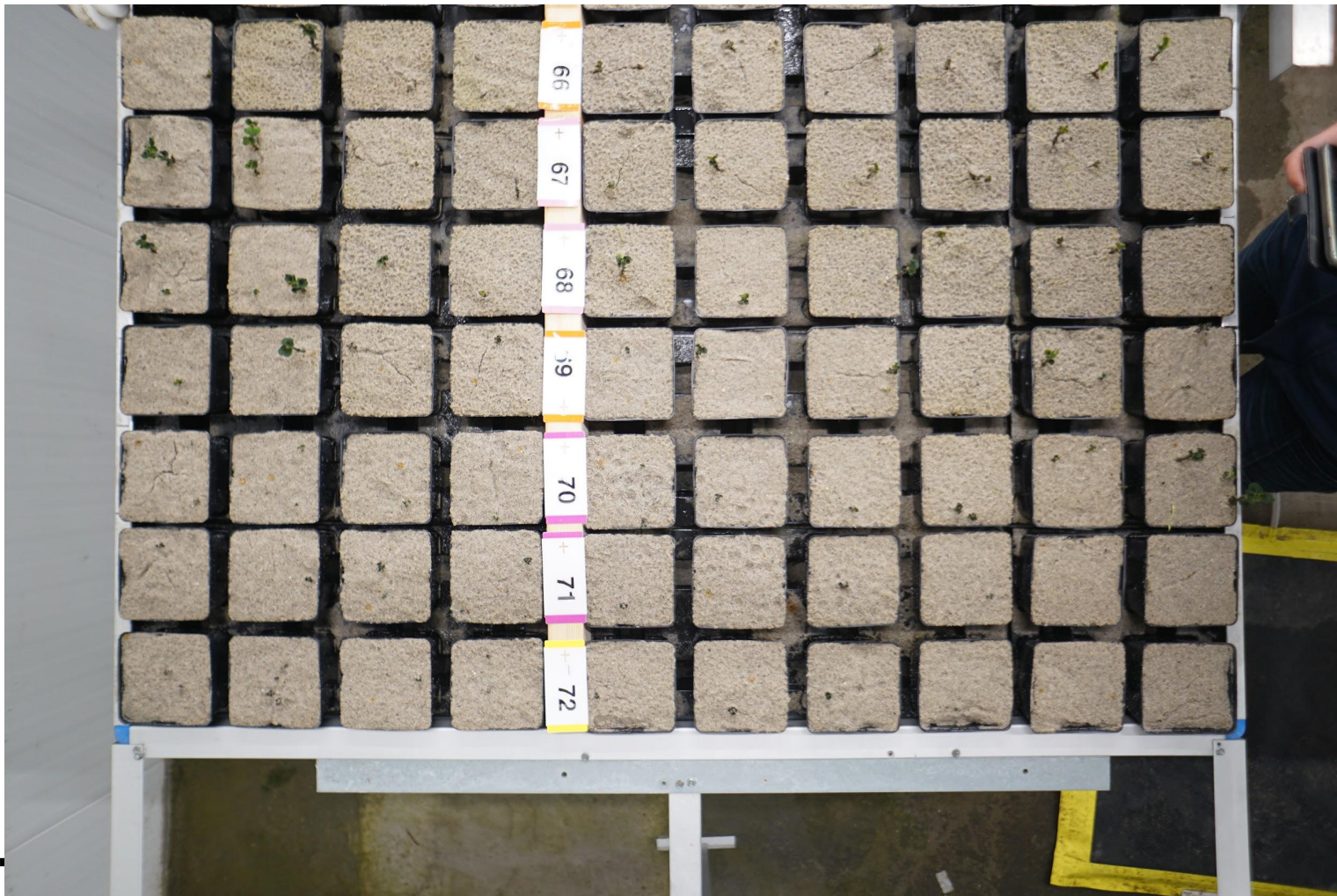


Problem

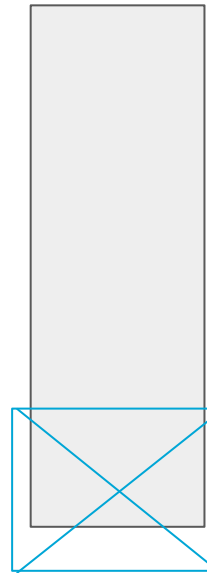


Features of interest

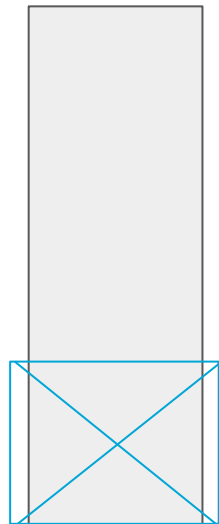
- Green
- Height
- Day of appearance
- Number of leaves
- Contour of leaves
- Number of plants

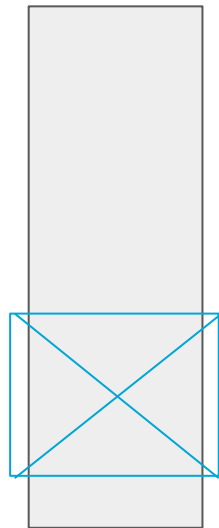


table

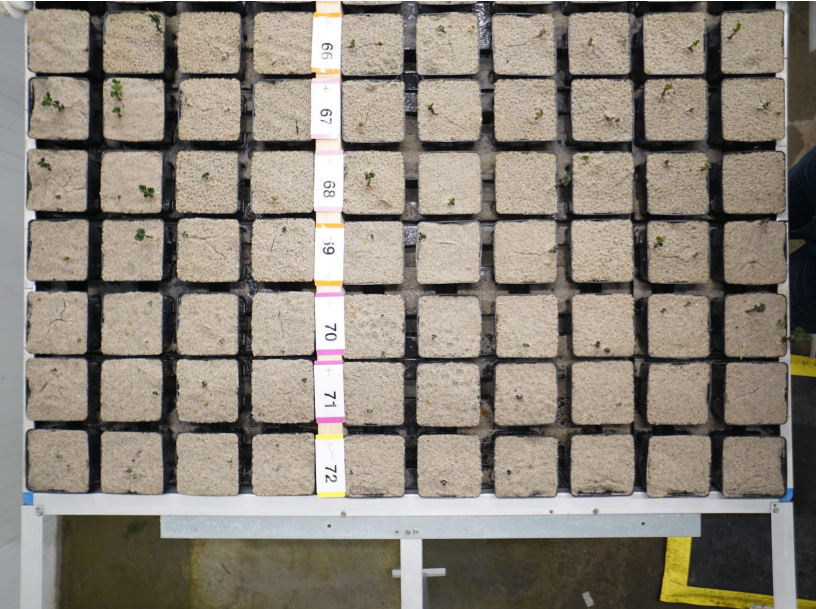


moving camera

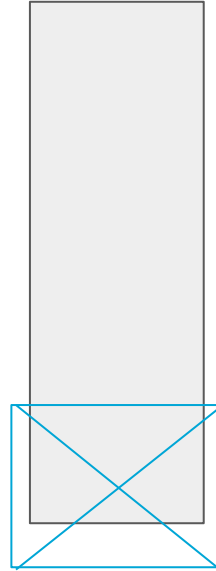




Snapshot n



2-4 days between
snapshots



Snapshot n+1



Focus

- Plant outline detection
- Matching points for 3 different camera views over time
- 3D point-reconstruction

Plant outline

- Traditional Snakes algorithm with Gradient Vector Flow force field
- Kass, Michael, Andrew Witkin, and Demetri Terzopoulos. "Snakes: Active contour models." *International journal of computer vision* 1.4 (1988): 321-331.
- Prince, Jerry L., and Chenyang Xu. "A new external force model for snakes." *Proc. 1996 Image and Multidimensional Signal Processing Workshop*. Vol. 3. No. 31. 1996.

Traditional Snakes

- Energy functional of a traditional snake parameterized by $\mathbf{x}(s) = (x(s), y(s))$

$$\begin{aligned} E_{\text{snake}}^* &= \int_0^1 E_{\text{snake}}(\mathbf{x}(s)) \, ds \\ &= \int_0^1 E_{\text{int}}(\mathbf{x}(s)) + E_{\text{ext}}(\mathbf{x}(s)) \, ds \end{aligned}$$

- α and β control parameters for amount of stretch and curvature

$$E[\mathbf{x}] = \int_0^1 \frac{1}{2} \left(\alpha \|\mathbf{x}'(s)\|^2 + \beta \|\mathbf{x}''(s)\|^2 \right) + E_{\text{ext}}(\mathbf{x}(s)) \, ds$$

- Assume local minimum of E in \mathbf{x} to derive Euler-Lagrange equations

$$\alpha \mathbf{x}'' - \beta \mathbf{x}'''' - \nabla E_{\text{ext}}(\mathbf{x}) = 0$$

Gradient Vector Flow

- Define new external force field $\mathbf{v}(x, y) = (u(x, y), v(x, y))$ which minimizes the functional

$$\mathcal{E}[\mathbf{v}] = \iint \mu (u_x^2 + u_y^2 + v_x^2 + v_y^2) + |\nabla f|^2 |\mathbf{v} - \nabla f|^2 \, dx dy$$

μ smoothing term and f edge map derived from image.

- Again assume minimum in \mathcal{E} and derive Euler-Lagrange equations

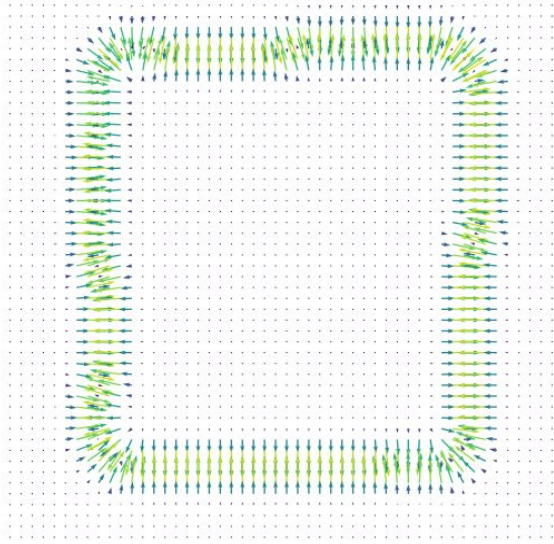
$$\mu \nabla^2 u - (f_x^2 + f_y^2) (u - f_x) = 0$$

$$\mu \nabla^2 v - (f_x^2 + f_y^2) (v - f_y) = 0$$

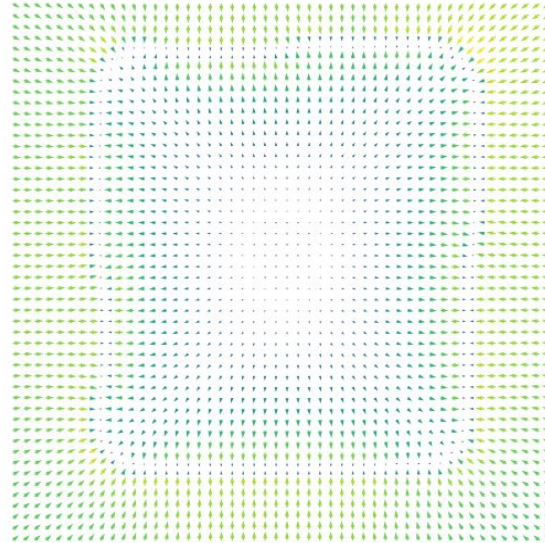
- GVF snake

$$\alpha \mathbf{x}'' - \beta \mathbf{x}'''' + \mathbf{v} = 0$$

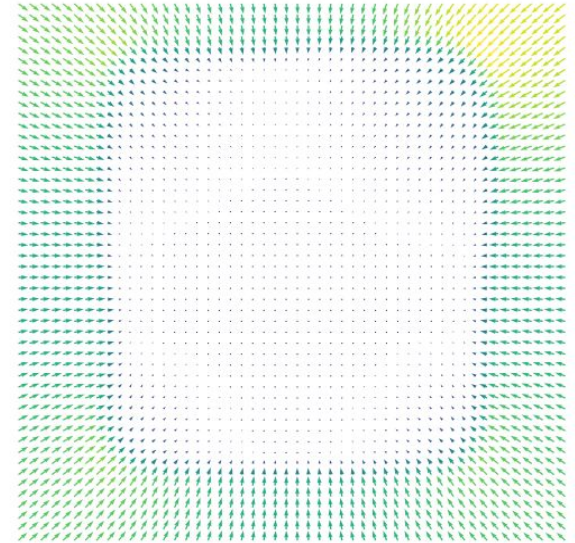
External force field



$\mu=0$



$\mu=1e-3$



$\mu=0.1$

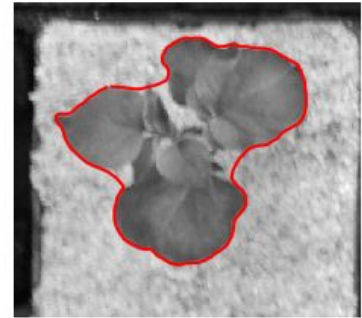
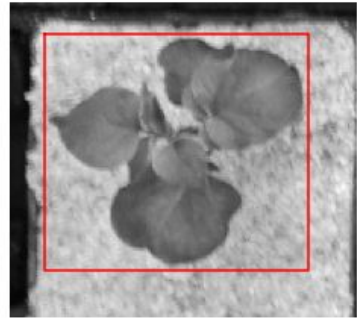
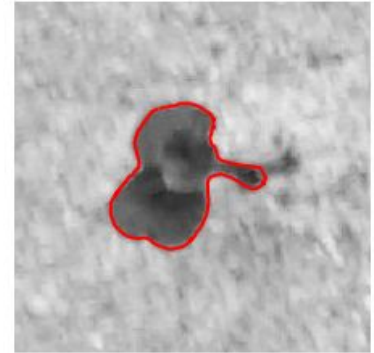
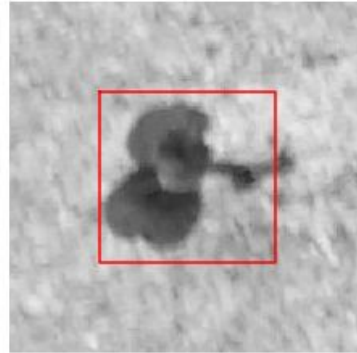
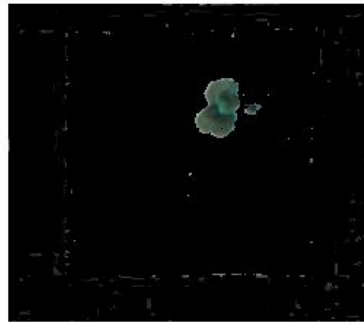
Discretization

- Dynamic snake equation $\frac{\partial \mathbf{x}}{\partial t} = \alpha \frac{\partial^2 \mathbf{x}}{\partial s^2} - \beta \frac{\partial^4 \mathbf{x}}{\partial s^4} + \mathbf{v}(\mathbf{x})$
discretize with central differences and solve with Euler Forward
- Solve for \mathbf{v} by considering BVP

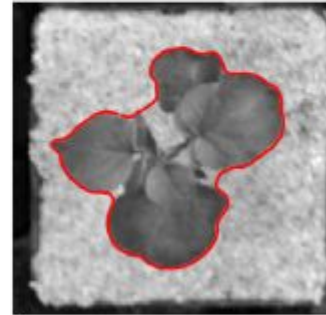
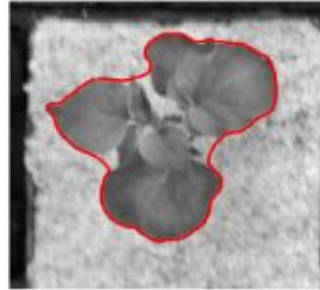
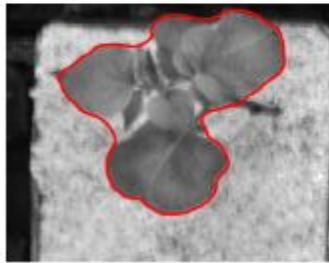
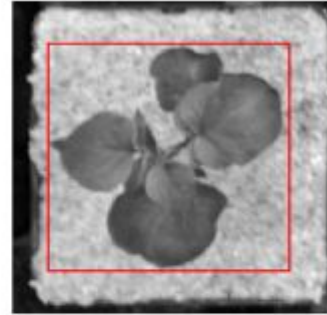
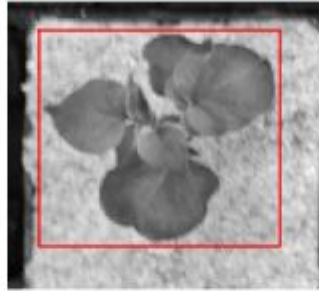
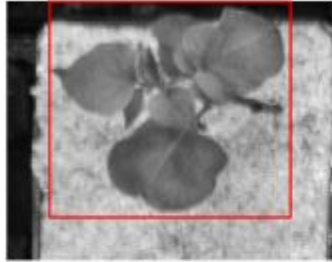
$$\begin{cases} \mu \nabla^2 u - gu = -gf_x, & \text{in } \Omega \\ \frac{\partial u}{\partial n} = 0, & \text{on } \partial\Omega \end{cases}, \quad \text{where } g = f_x^2 + f_y^2$$

discretize with central differences and solve with MINRES

Initial and final snake



Plant outline for 3 camera views



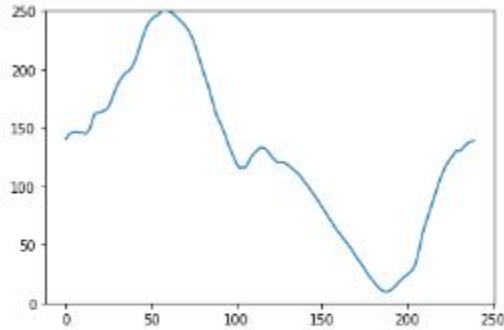
(a) Angle -1

(b) Angle 0

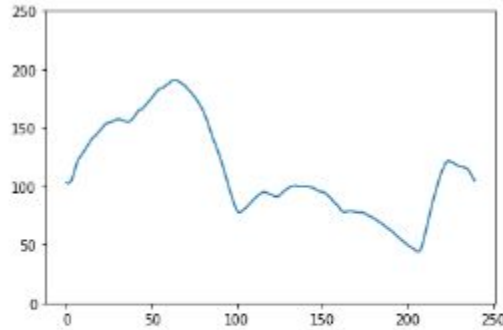
(c) Angle +1

Matching points on curve

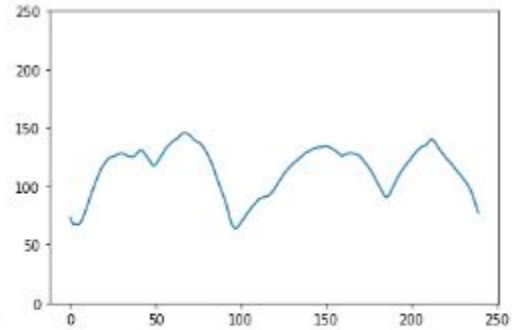
- Radius with respect to centre point plotted against curve



(a) Angle -1

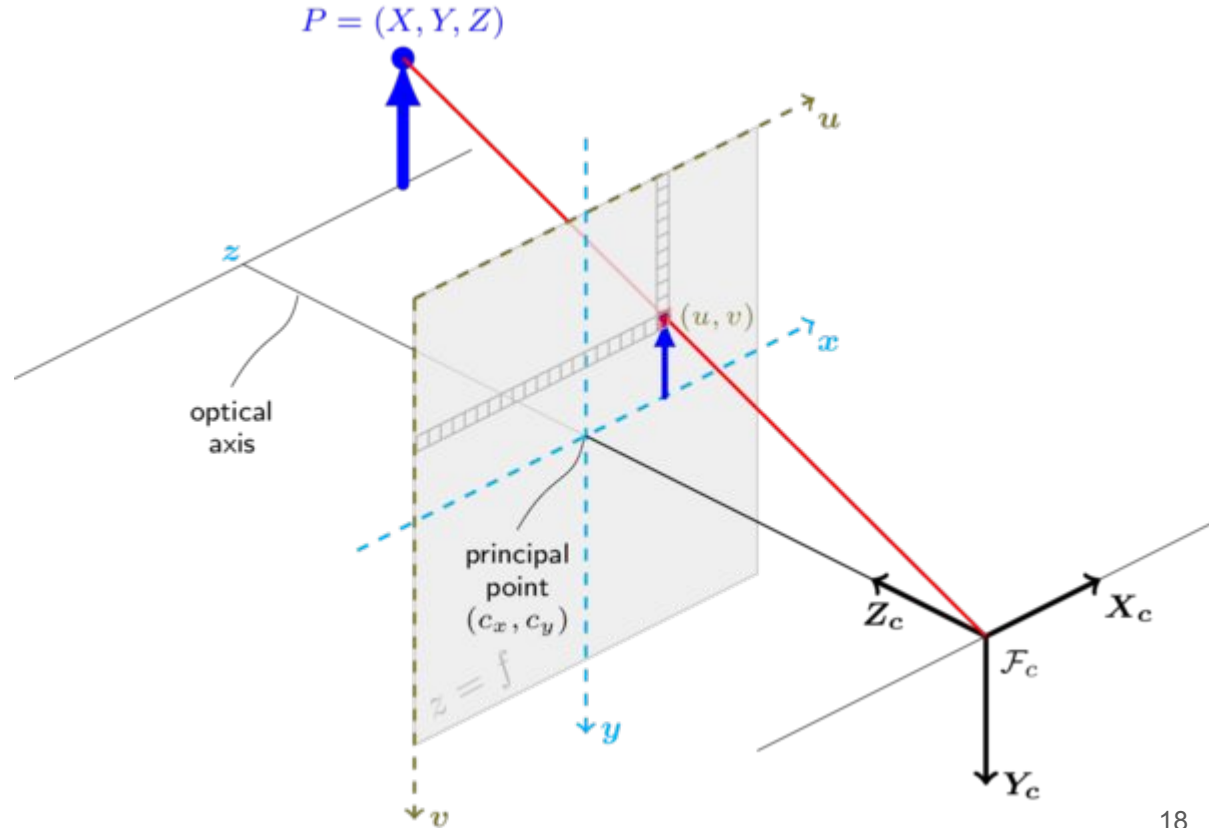
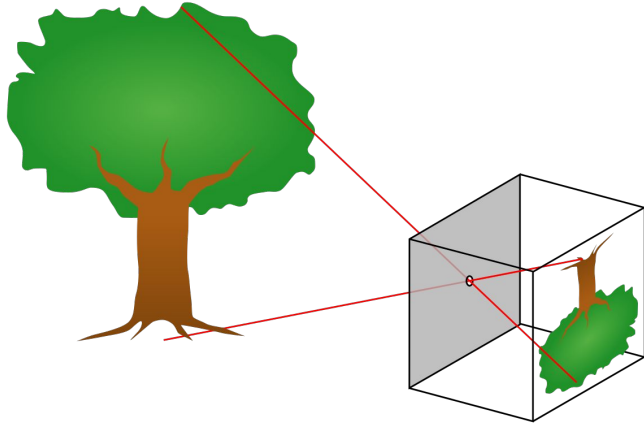


(b) Angle 0



(c) Angle +1

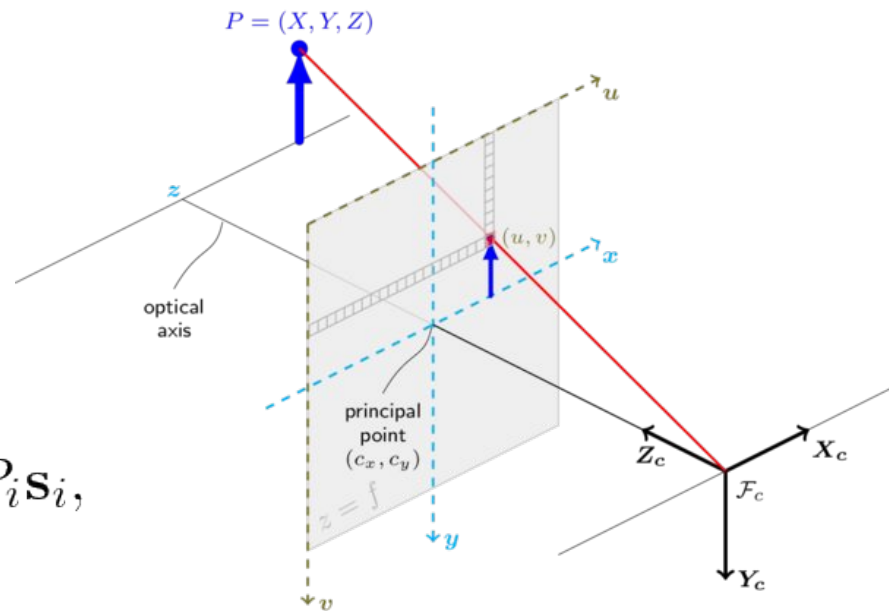
Pinhole camera model



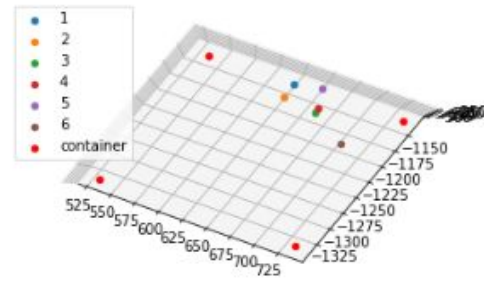
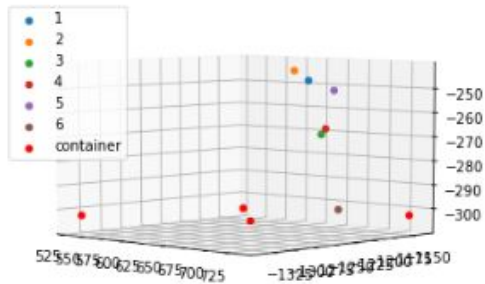
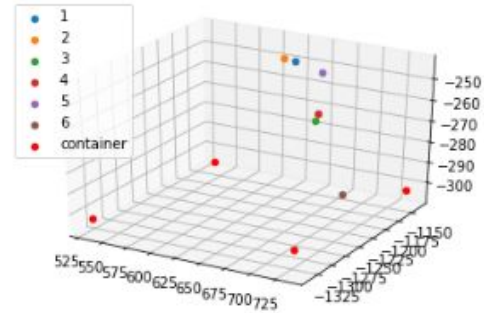
Structure from motion

- $\begin{bmatrix} u \\ v \end{bmatrix} = \frac{f}{Z} \begin{bmatrix} X \\ Y \end{bmatrix}$
- $P\mathbf{r}' = \mathbf{0}$
- $\mathbf{r}' = \mathbf{s} + R\mathbf{r}$
- if we assume translation \mathbf{s} and rotation \mathbf{R} to be known

$$R^T \left(\sum_{i=1}^3 P_i^T P_i \right) R\mathbf{r} = -R^T \sum_{i=1}^3 P_i^T P_i \mathbf{s}_i,$$

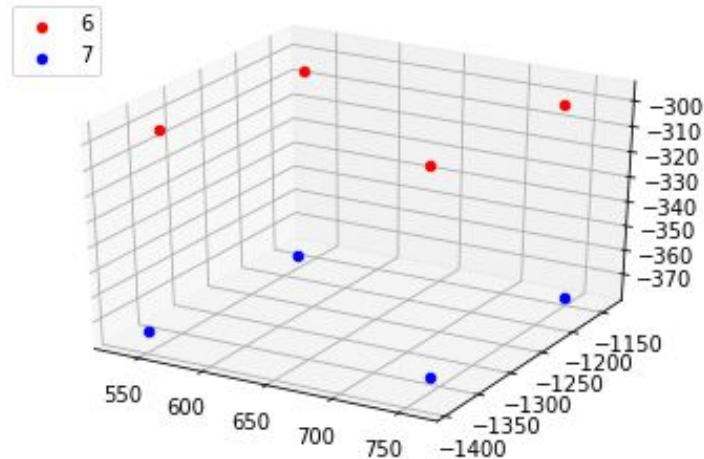


3D point-reconstruction



3D point-reconstruction

- Error in invariant points like container
- Caused by small perturbations in camera's positioning
- Translation and rotation not as assumed



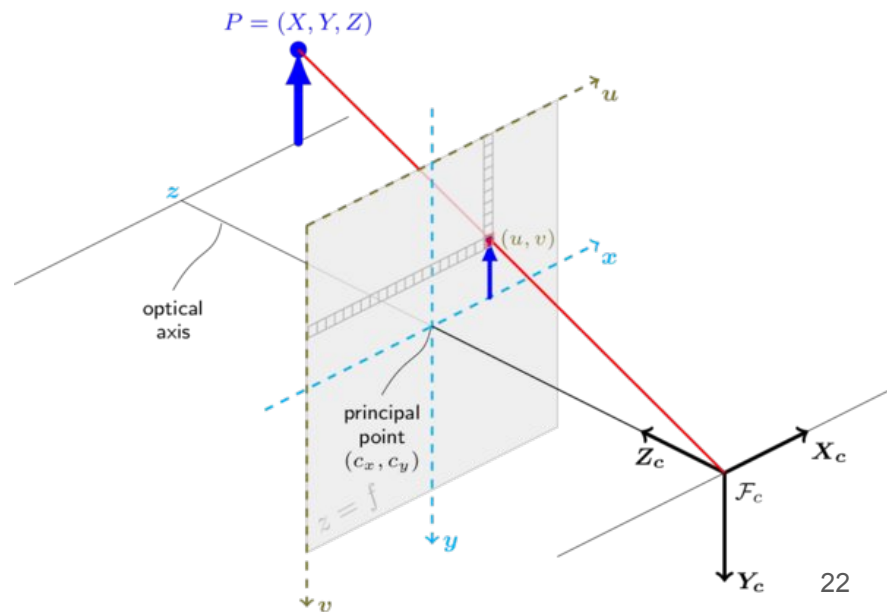
Calibration

- A. Dermanis - Mathematical Foundations of Analytical Photogrammetry
- Estimate camera's internal and external parameters
- Collinearity condition
- Collinearity equations:

$$u = u_0 - f \frac{R_{11}(x - x_0) + R_{12}(y - y_0) + R_{13}(z - z_0)}{R_{31}(x - x_0) + R_{32}(y - y_0) + R_{33}(z - z_0)}$$

$$v = v_0 - f \frac{R_{21}(x - x_0) + R_{22}(y - y_0) + R_{23}(z - z_0)}{R_{31}(x - x_0) + R_{32}(y - y_0) + R_{33}(z - z_0)}$$

- Set ground control points



Calibration

- Direct Linear Transformation method

$$u = \frac{L_1x + L_2y + L_3z + L_4}{L_9x + L_{10}y + L_{11}z + 1}$$

$$v = \frac{L_5x + L_6y + L_7z + L_8}{L_9xL_{10}y + L_{11}z + 1}$$

- 11 unknowns, 6 ground control points

Calibration

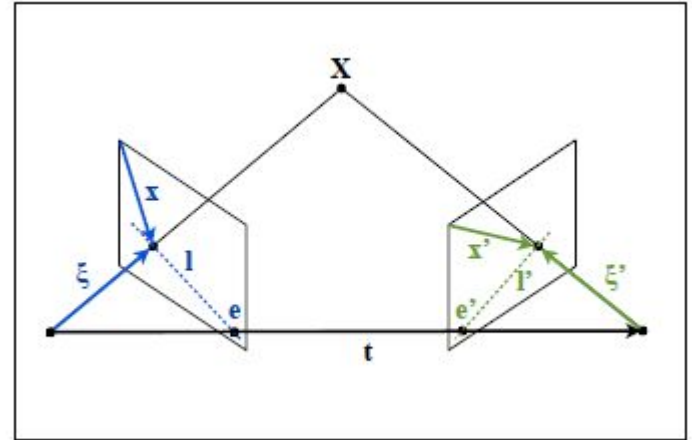
- $A\mathbf{l} = \mathbf{0}$

$$A = \begin{bmatrix} x_1 & y_1 & z_1 & 1 & 0 & 0 & 0 & 0 & -u_1 x_1 & -u_1 y_1 & -u_1 z_1 & -u_1 \\ 0 & 0 & 0 & 0 & x_1 & y_1 & z_1 & 1 & -v_1 x_1 & -v_1 y_1 & -v_1 z_1 & -v_1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x_n & y_n & z_n & 1 & 0 & 0 & 0 & 0 & -u_n x_n & -u_n y_n & -u_n z_n & -u_n \\ 0 & 0 & 0 & 0 & x_n & y_n & z_n & 1 & -v_n x_n & -v_n y_n & -v_n z_n & -v_n \end{bmatrix}$$

- find null space of A using SVD
- edges of container don't contain info about depth, resulting in 3 zero columns of A and a non-unique solution \mathbf{l}

Coplanarity condition

- Schindler, Konrad. "Mathematical foundations of photogrammetry." *Handbook of Geomathematics*. Springer, 2015. 3087-3103.
- $\xi \cdot (\mathbf{t} \times \xi') = 0$
- this relates to the fundamental matrix
 $\mathbf{x}^T F \mathbf{x}' = 0$
- camera matrices can be obtained
 $[I \mid \mathbf{0}]$, $[[\mathbf{e}']_{\times} F \mid \mathbf{e}']$



Fundamental matrix

- F 3x3 matrix with rank 2
- $B\mathbf{f} = \mathbf{0}$

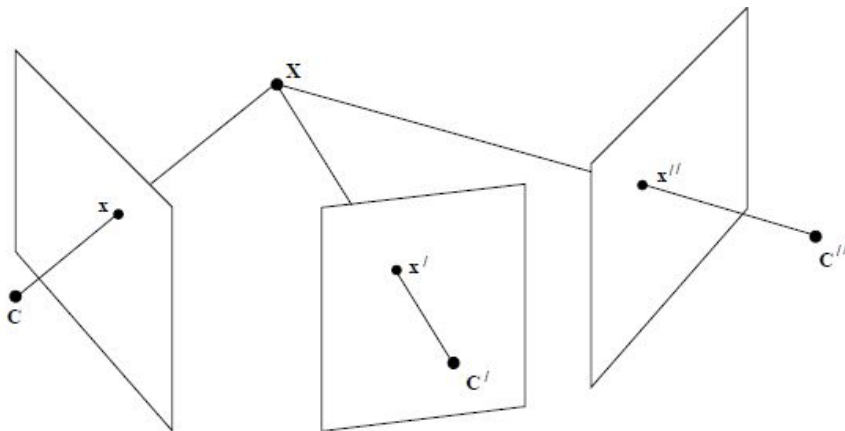
$$B = \begin{bmatrix} x'_1 x_1 & x'_1 y_1 & y'_1 x_1 & y'_1 y_1 & x'_1 & y'_1 & x_1 & y_1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ x'_n x_n & x'_n y_n & y'_n x_n & y'_n y_n & x'_n & y'_n & x_n & y_n & 1 \end{bmatrix}$$

- find null space of B using SVD
- solve with 8 point correspondences

Trifocal tensor

- Hartley, Richard, and Andrew Zisserman. *Multiple view geometry in computer vision*. Cambridge university press, 2003.
- Extension to three camera views
- Three 3x3 matrices T_i

$$[\mathbf{x}']_{\times} \left(\sum_i x^i \mathbf{T}_i \right) [\mathbf{x}'']_{\times} = \mathbf{0}_{3 \times 3}.$$



Next steps

- Explore trifocal tensor
- From fundamental matrix/trifocal tensor to more accurate 3D reconstruction in the reference system of object
- Matching points