

# Contour Detection in Multi-Angle Time-Lapse Images of Growing Plants

Merel te Hofsté

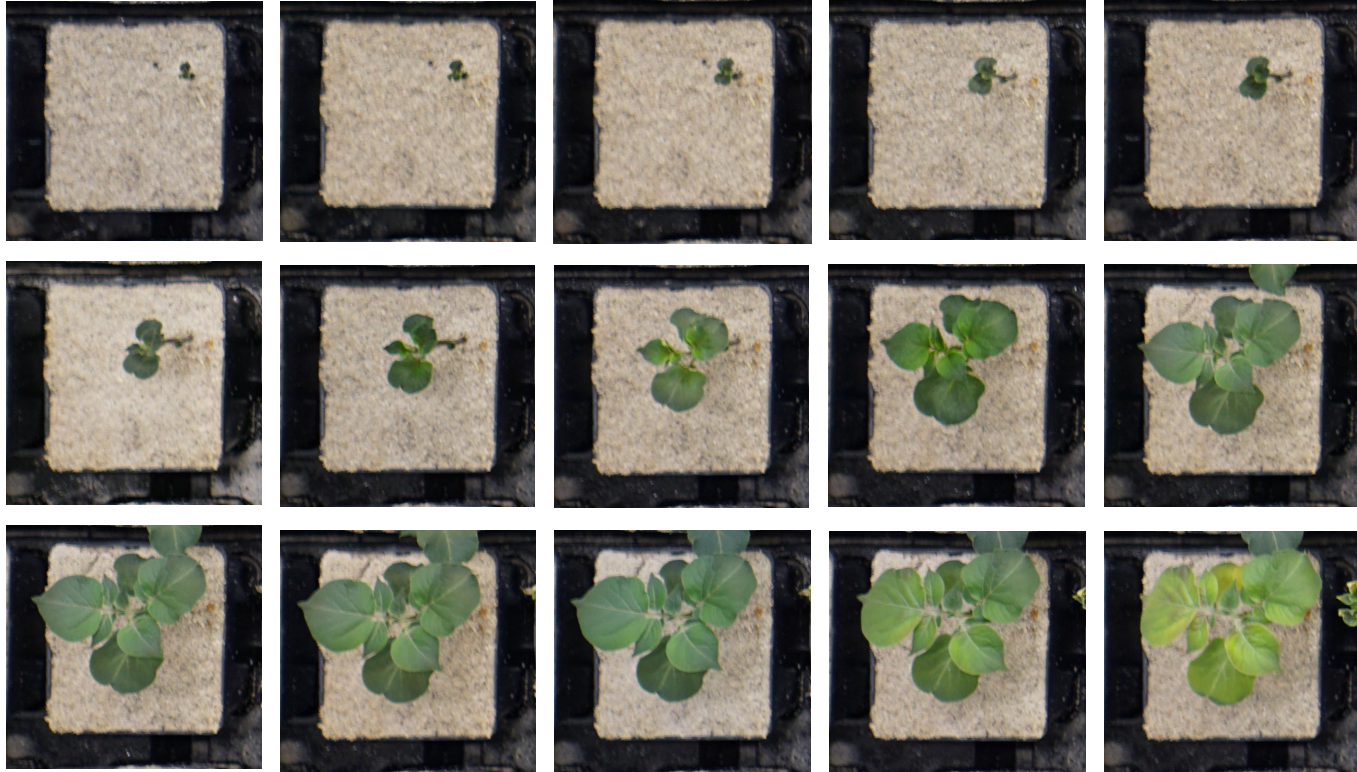
Supervisor: Neil Budko

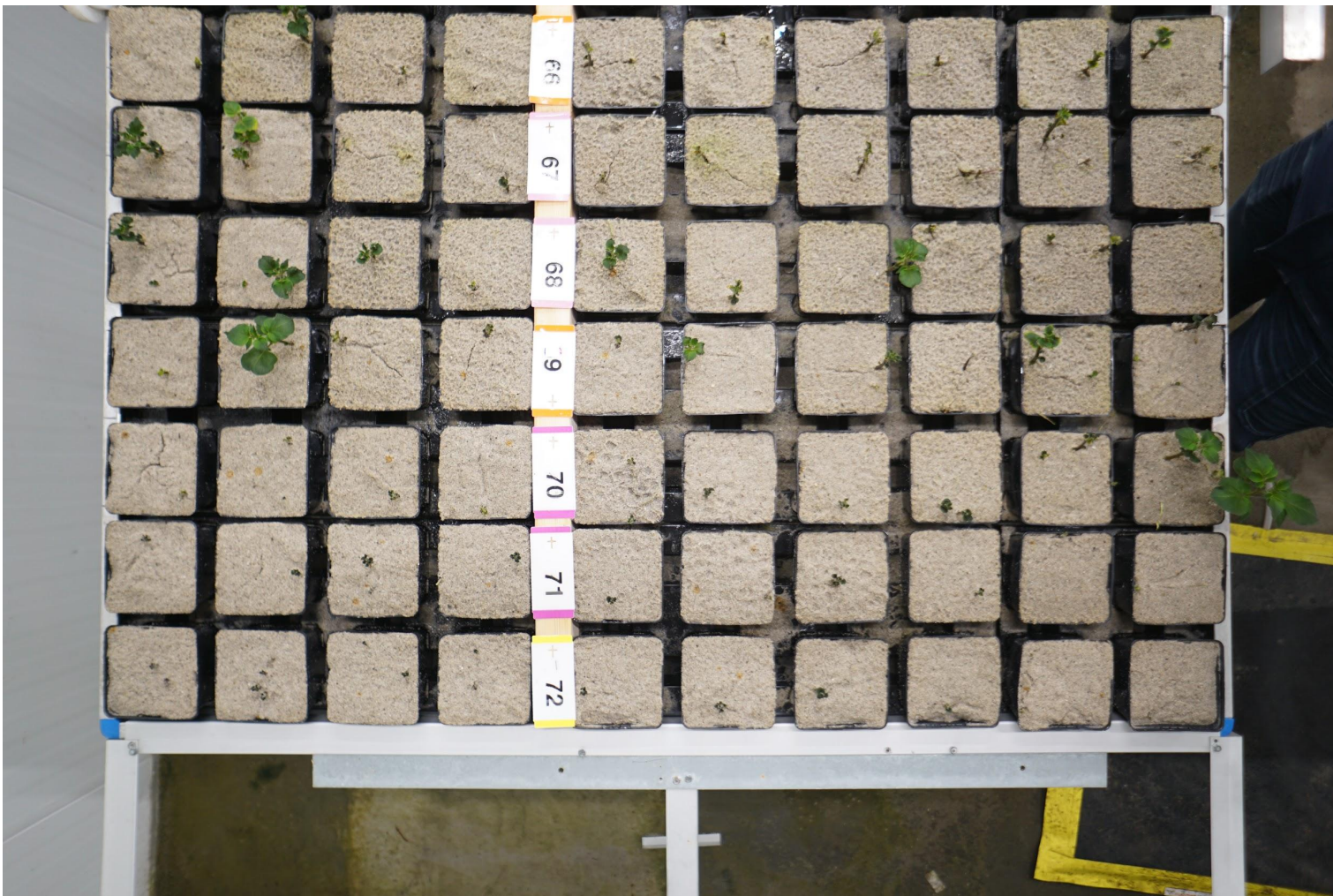


# Quantifying growth

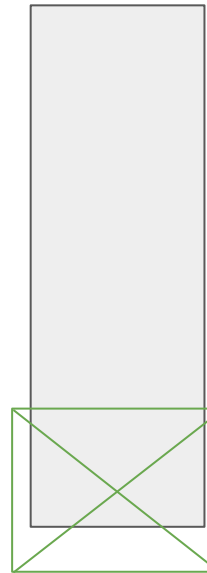
- Six potato varieties
- 2 Climate rooms
- Dry and wet sections

# Problem



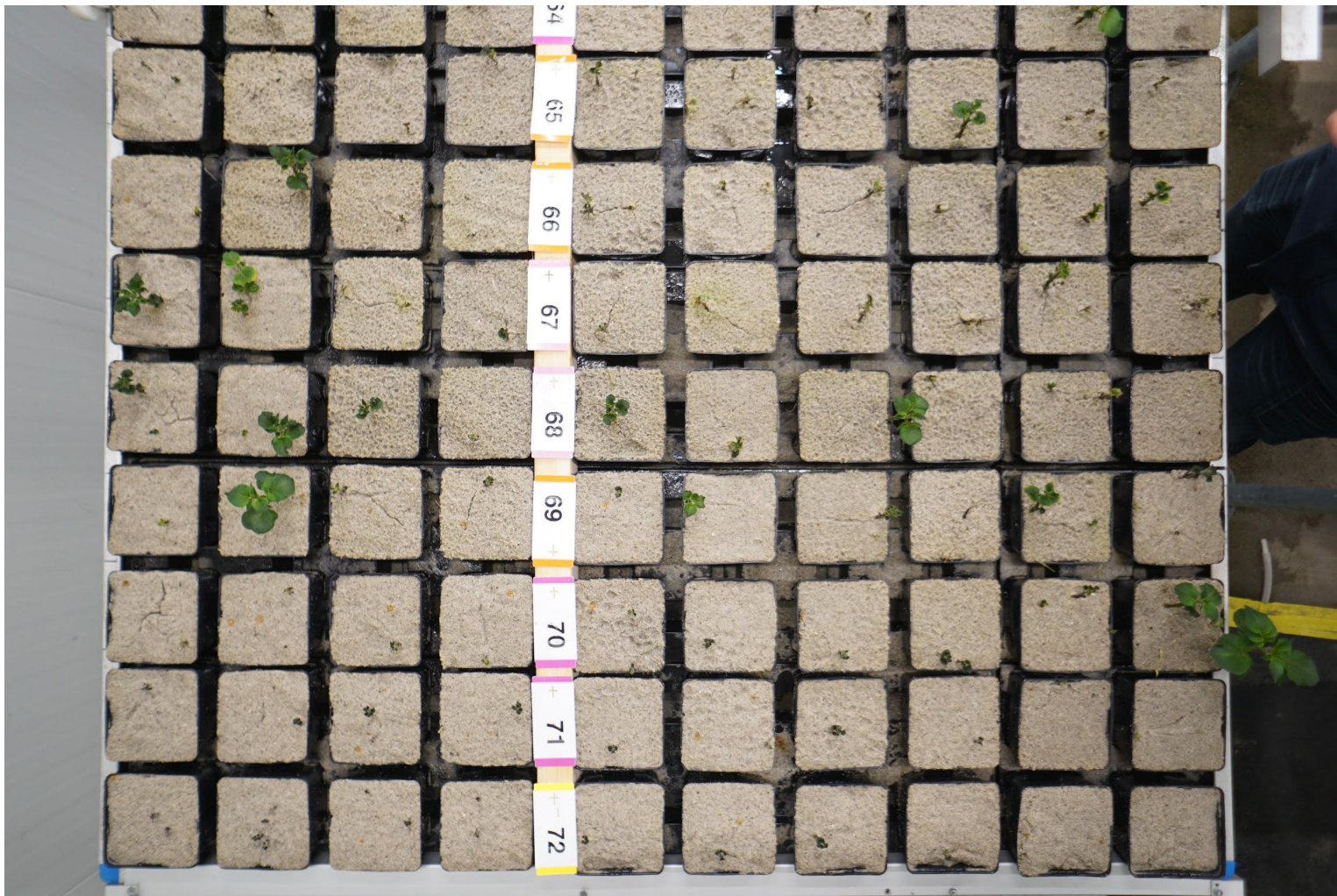


table

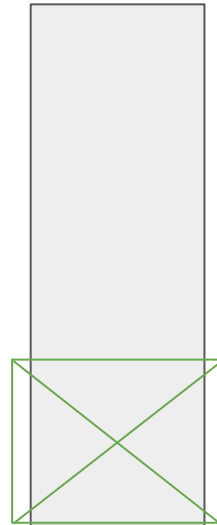


moving camera





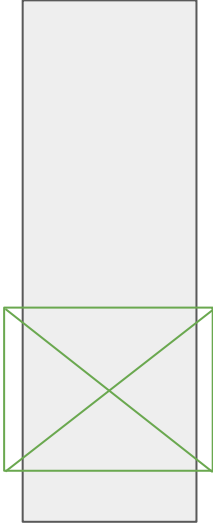
table







table



# Overview



## Contour detection

To detect the outline of the plant



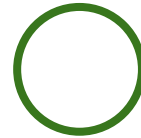
## 3D Contour Reconstruction

Using a pinhole camera model



## Double Snakes

Combination of previous problems

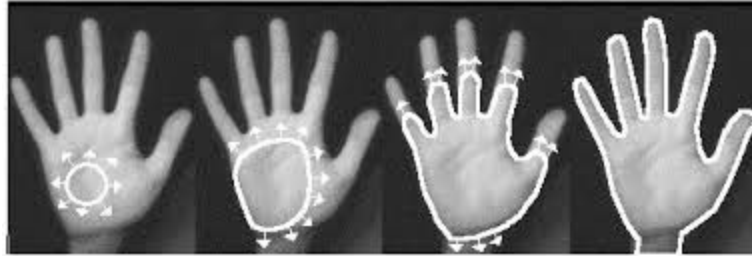


## Conclusion

Conclusion and recommendations

# Traditional Snakes

- Active contours to detect outline in image



Contour  
detection





# Traditional Snakes

Contour  
detection

- Energy functional of a traditional snake parameterized by  $\mathbf{q}(s) = [u(s), v(s)]$

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

- $\alpha$  and  $\beta$  control parameters for amount of stretch and curvature

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

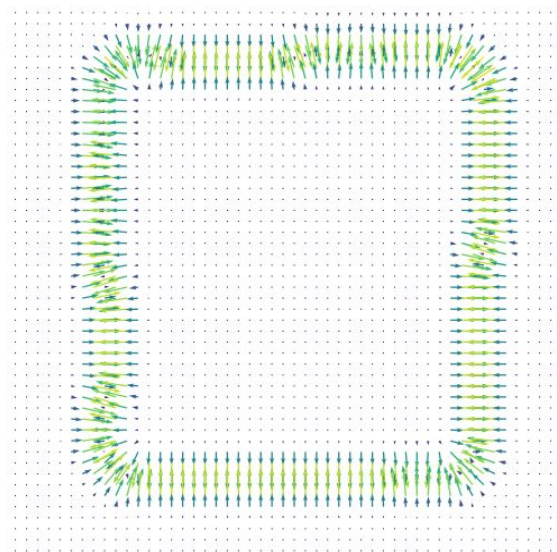
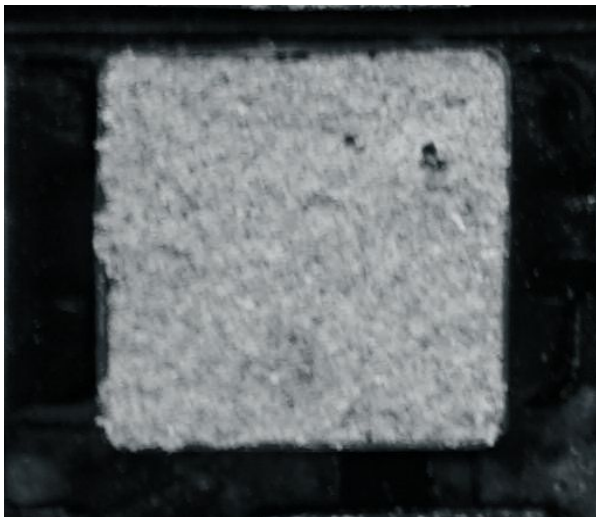
- Assume local minimum of  $E$  in  $\mathbf{q}$  to derive the Euler-Lagrange equation

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



# Traditional Snakes

- Typical external energy  $E_{\text{ext}}(\mathbf{q}) = -|\nabla I(\mathbf{q})|^2$   
which points towards regions of interest in image  $I$



Contour  
detection



# Gradient Vector Flow

Contour  
detection



- New external force field  $\mathbf{w}(u, v) = [\phi(u, v), \psi(u, v)]$   
which minimizes the functional

$$\mathcal{E}[\mathbf{w}] = \iint \mu (\phi_u^2 + \phi_v^2 + \psi_u^2 + \psi_v^2) + |\nabla f|^2 |\mathbf{w} - \nabla f|^2 \, du \, dv$$

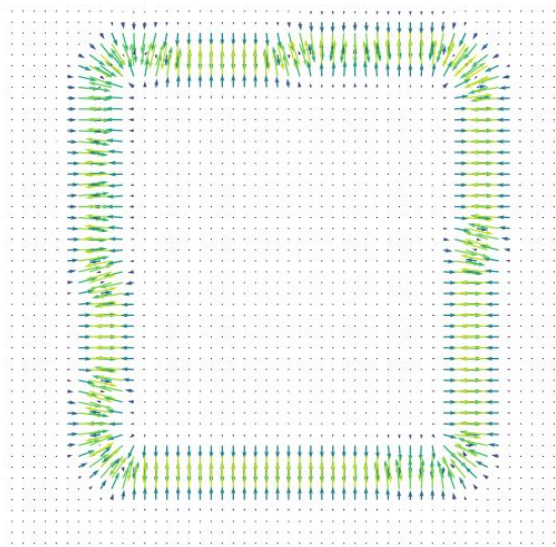
$\mu$  smoothing term and edge map  $f$  derived from image

- GVF Snake

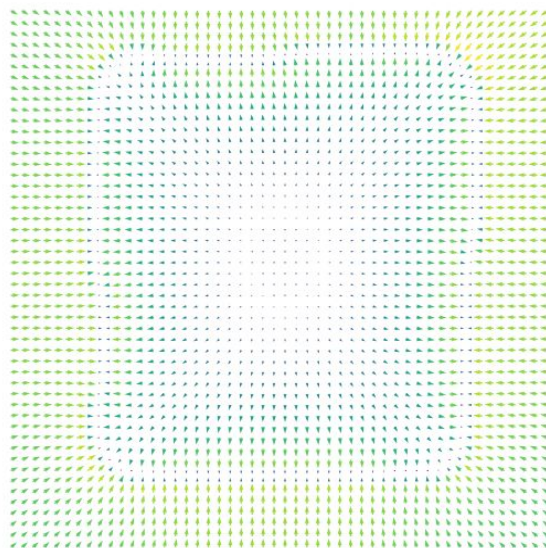
$$\alpha \mathbf{q}'' - \beta \mathbf{q}'''' + \mathbf{w} = 0$$



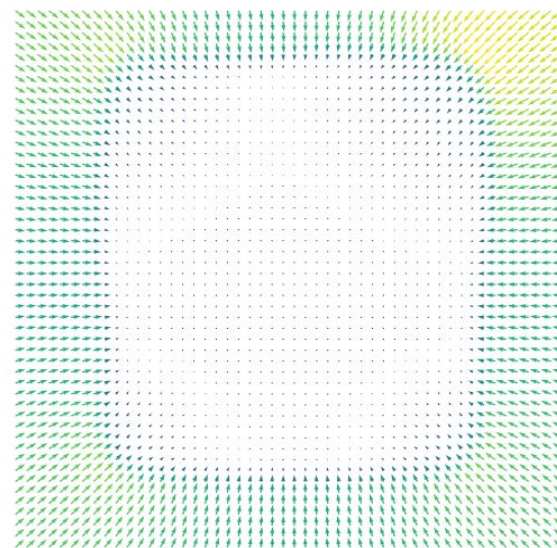
# Gradient Vector Flow



$\mu=0$



$\mu=1e-3$



$\mu=0.1$

Contour  
detection

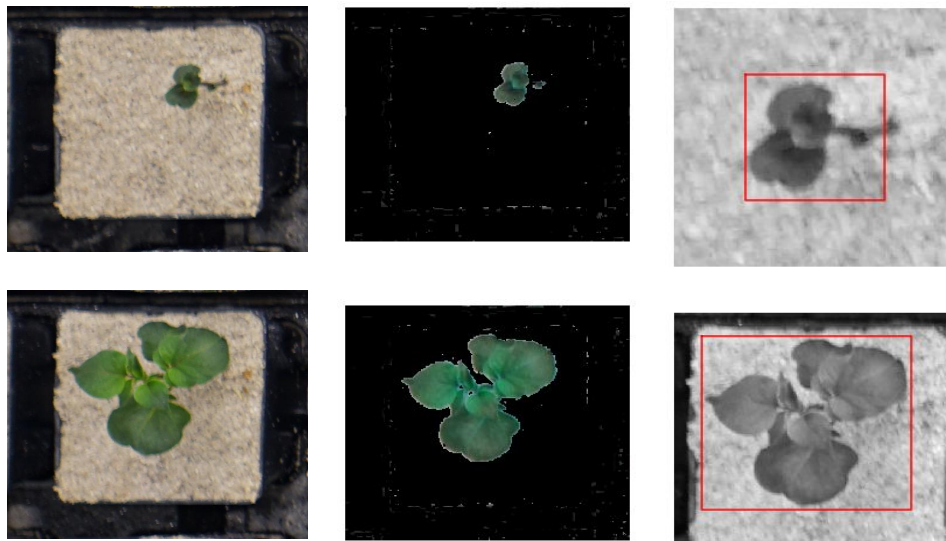


# GVF-Snakes

- Solve for  $\frac{\partial \mathbf{q}}{\partial t} = \alpha \frac{\partial^2 \mathbf{q}}{\partial s^2} - \beta \frac{\partial^4 \mathbf{q}}{\partial s^4} + \mathbf{w}(\mathbf{q})$

with time-integration method

- Initial contour



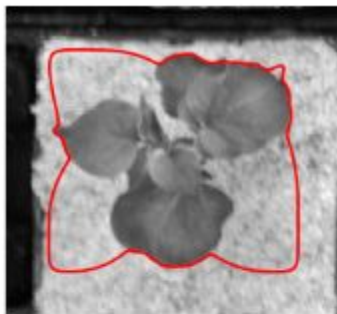
Contour  
detection



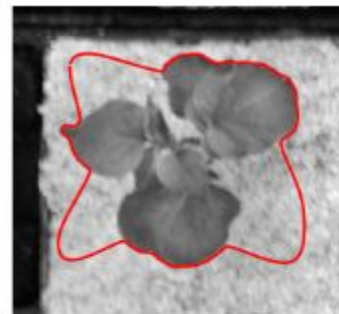
# Results



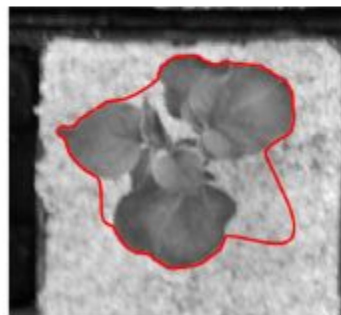
(a)  $t = 0$



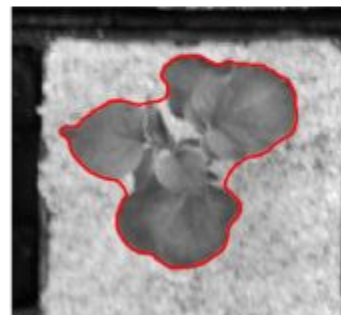
(b)  $t = 1$



(c)  $t = 5$



(d)  $t = 10$



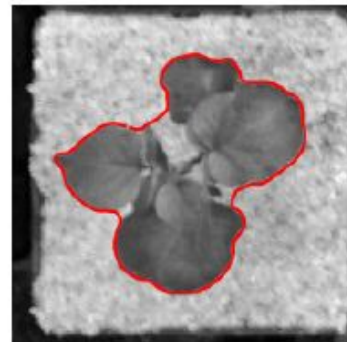
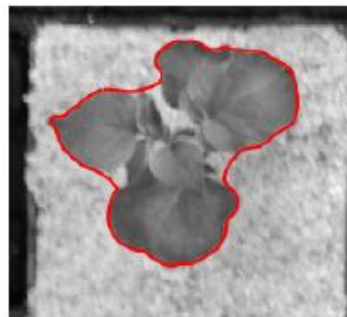
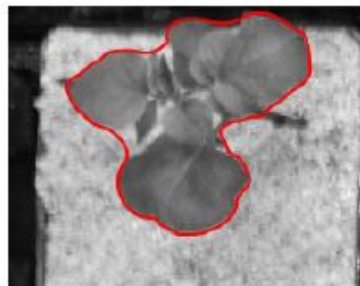
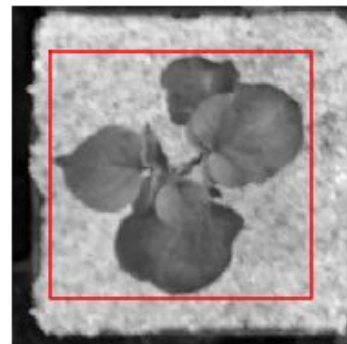
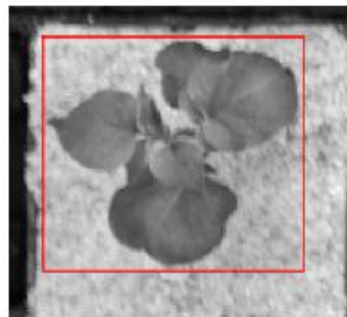
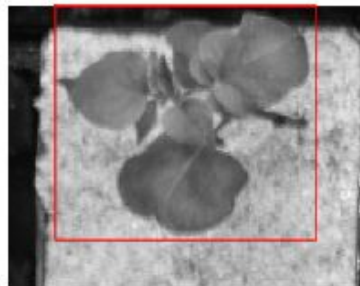
(e)  $t = 50$

Contour  
detection





# Results



(a) Angle -1

(b) Angle 0

(c) Angle +1

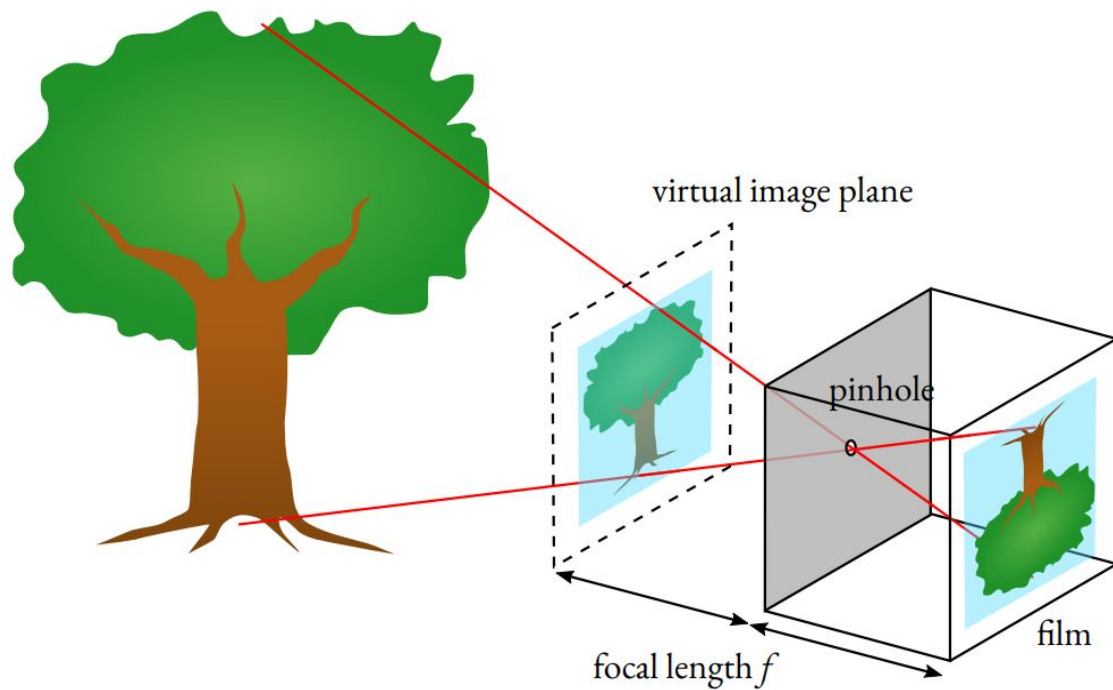
Contour  
detection



# GVF-Snakes

- Algorithm to detect outline
- Success ensured by initial contour and preprocessing steps

# Pinhole camera model



3D Contour  
Reconstruction





# Pinhole camera model

- Point in camera reference frame:

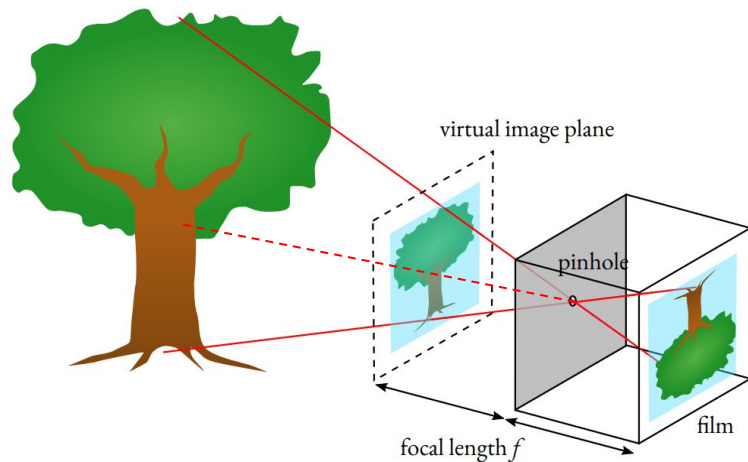
$$\mathbf{r}' = [x', y', z']$$

- Projection equations:

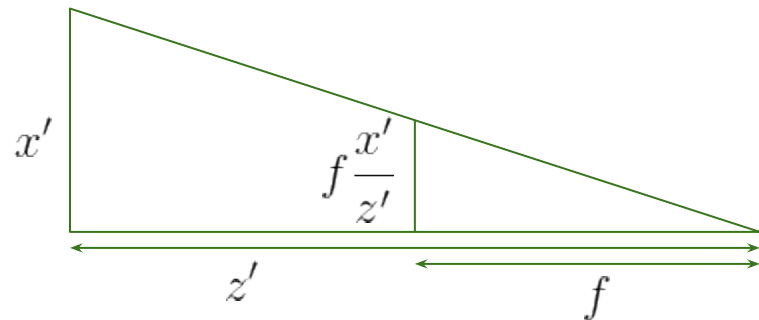
$$f \frac{x'}{z'} = u - u_0$$

$$f \frac{y'}{z'} = v - v_0$$

$(u_0, v_0)$  center of image plane



3D Contour  
Reconstruction



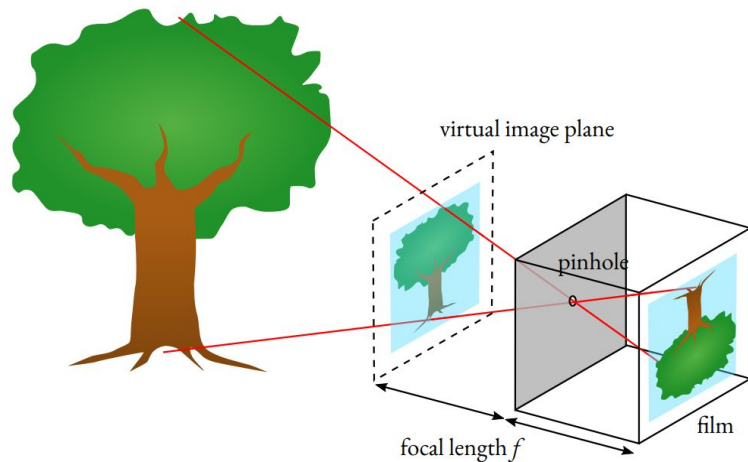
# Pinhole camera model

- Point in camera reference frame:

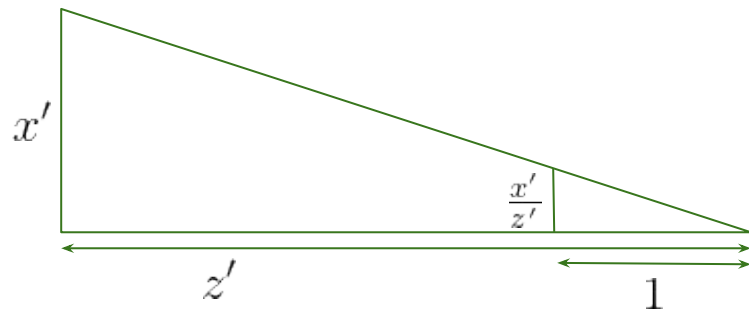
$$\mathbf{r}' = [x', y', z']$$

- Normalized image coordinates:

$$\frac{1}{z'} \mathbf{r}' = \begin{bmatrix} x'/z' \\ y'/z' \\ z'/z' \end{bmatrix} = \mathbf{y}$$



3D Contour  
Reconstruction



# Two-view geometry

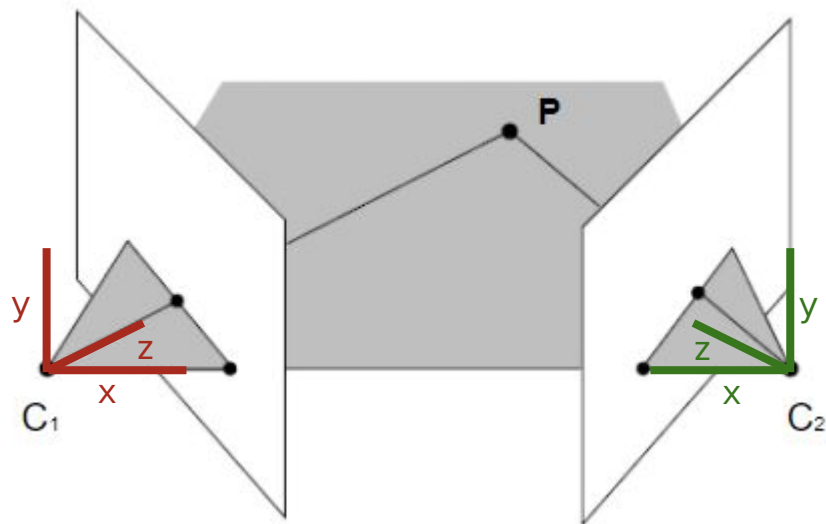
- Two camera's aiming at the same point  $\mathbf{P}$

$$\mathbf{r}' = [x', y', z']$$

$$\mathbf{r}'' = [x'', y'', z'']$$

$$\mathbf{t} = [t_1, t_2, t_3]$$

$$\mathbf{r}'' = R\mathbf{r}' + \mathbf{t}$$



3D Contour  
Reconstruction



# Essential Matrix

- $\mathbf{r}'' = R\mathbf{r}' + \mathbf{t}$   
 $z''\mathbf{y}'' = z'R\mathbf{y}' + \mathbf{t}$   
 $0 = (\mathbf{y}'')^T E\mathbf{y}'$

3D Contour  
Reconstruction





# Essential Matrix

- $\mathbf{r}'' = R\mathbf{r}' + \mathbf{t}$   
 $z''\mathbf{y}'' = z'R\mathbf{y}' + \mathbf{t}$   
 $0 = (\mathbf{y}'')^T E\mathbf{y}'$

- Where  $E$  is a 3x3 matrix called the Essential Matrix defined by  $E = [\mathbf{t}]_{\times} R$

$$\mathbf{t} \times R = [\mathbf{t}]_{\times} R, \quad [\mathbf{t}]_{\times} = \begin{bmatrix} 0 & -t_3 & t_2 \\ t_3 & 0 & -t_1 \\ -t_2 & t_1 & 0 \end{bmatrix}$$



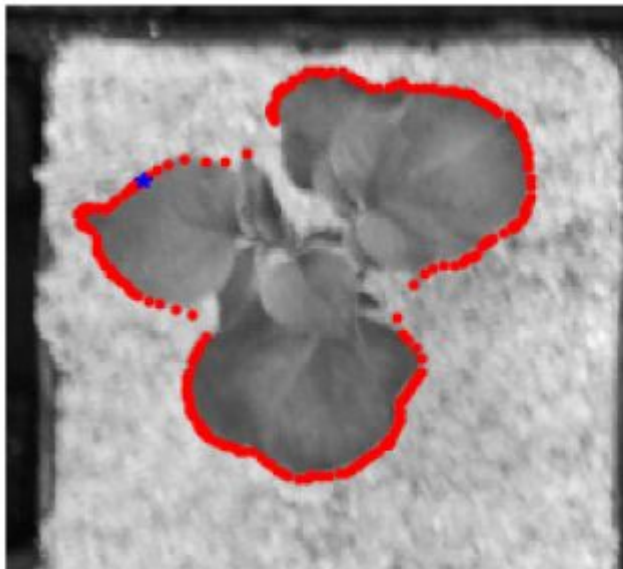
# Essential Matrix

- Determine  $E$  from  $(\mathbf{y}'')^T E \mathbf{y}' = 0$
- $Y \mathbf{e} = \mathbf{0}$ ,  $n \times 9$  matrix  $Y$
- Minimum of 8 point correspondences
- Find nullspace of  $Y$  using Singular Value Decomposition  
 $U, D, V^T = \text{svd}(Y)$   
in column of  $V$

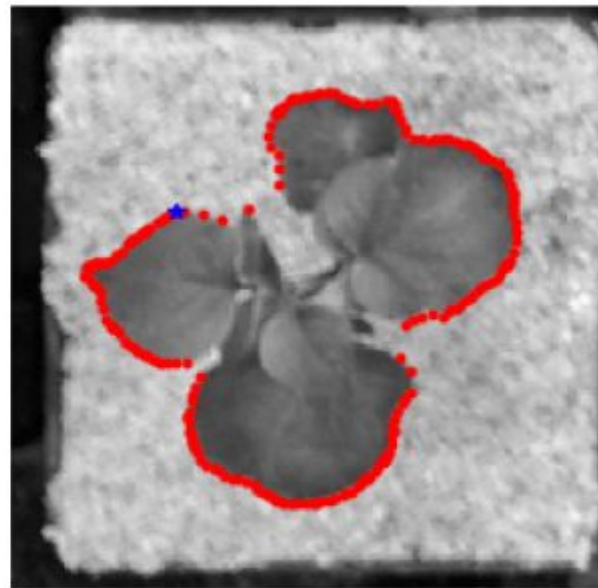
3D Contour  
Reconstruction



# Point matching



(a) Image and contour 1



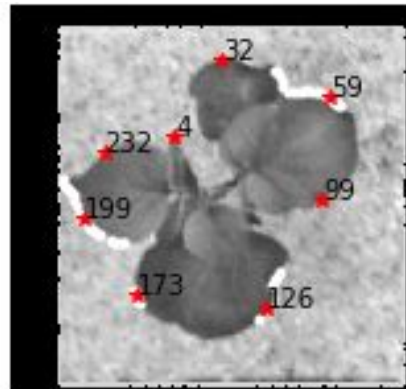
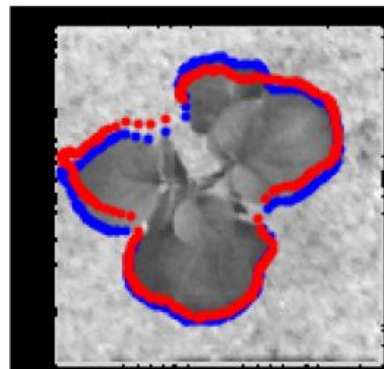
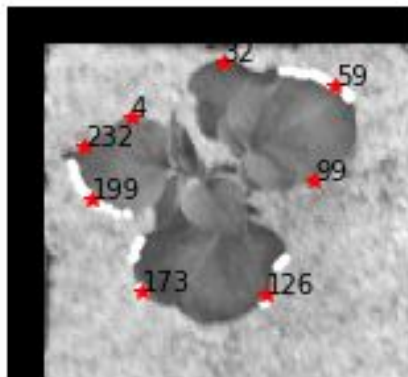
(b) Image and contour 2

3D Contour  
Reconstruction



# Point matching

- Rough alignment
- Choose 8 random points on contour and find the corresponding closest points
- Compute  $E$  and check  $(\mathbf{y}''')^T E \mathbf{y}' < \varepsilon$
- Continue until satisfied



3D Contour  
Reconstruction





# 3D Contour reconstruction

- Obtain rotation and translation via another Singular Value Decomposition of  $E$
- Retrieve depth coordinate

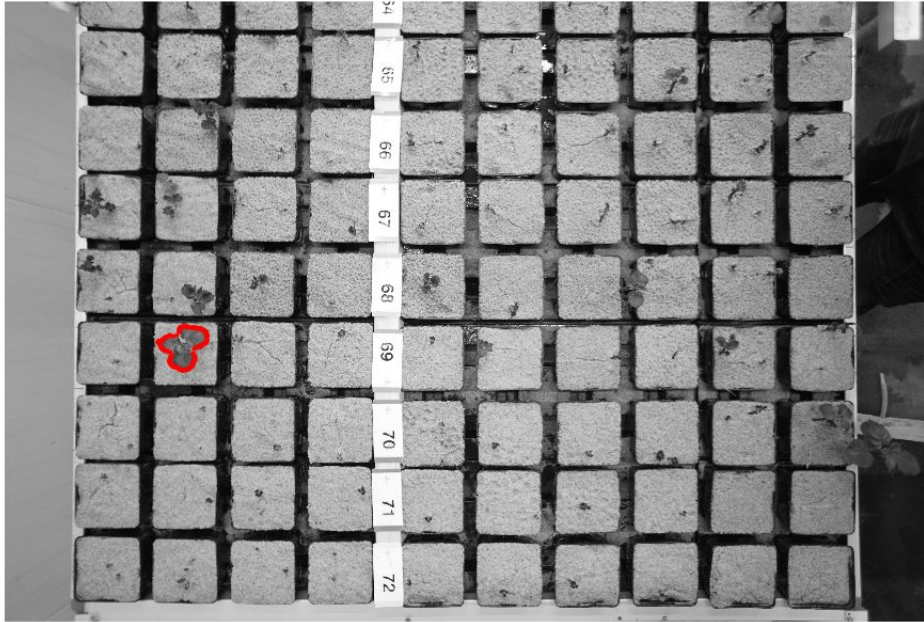
$$\mathbf{y}'' = \frac{1}{z''} \mathbf{r}'' = \frac{R\mathbf{r}' + \mathbf{t}}{\mathbf{e}_3^T (R\mathbf{r}' + \mathbf{t})}$$

$$\mathbf{y}'' = \frac{Rz'\mathbf{y}' + \mathbf{t}}{\mathbf{e}_3^T (Rz'\mathbf{y}' + \mathbf{t})}$$

$$z' = -\frac{\mathbf{y}'^T R^T (\mathbf{y}'' \mathbf{e}_3^T - I)^T (\mathbf{y}'' \mathbf{e}_3^T - I) \mathbf{t}}{\mathbf{y}'^T R^T (\mathbf{y}'' \mathbf{e}_3^T - I)^T (\mathbf{y}'' \mathbf{e}_3^T - I) R\mathbf{y}'}$$



# 3D Contour reconstruction



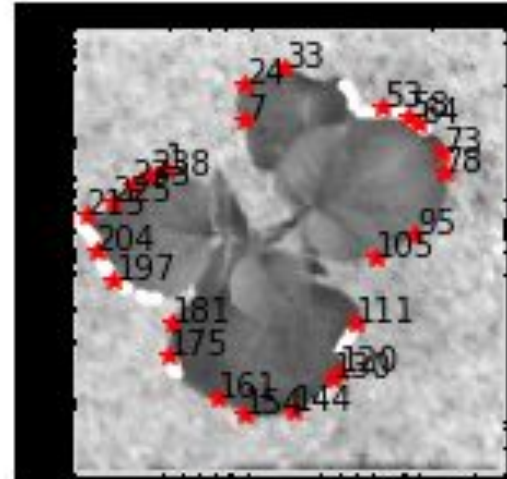
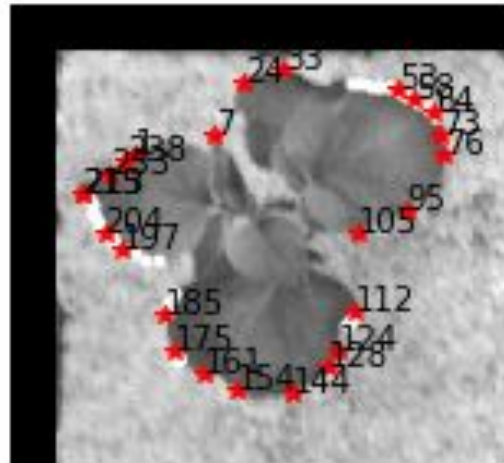
3D Contour  
Reconstruction



# 3D Contour reconstruction

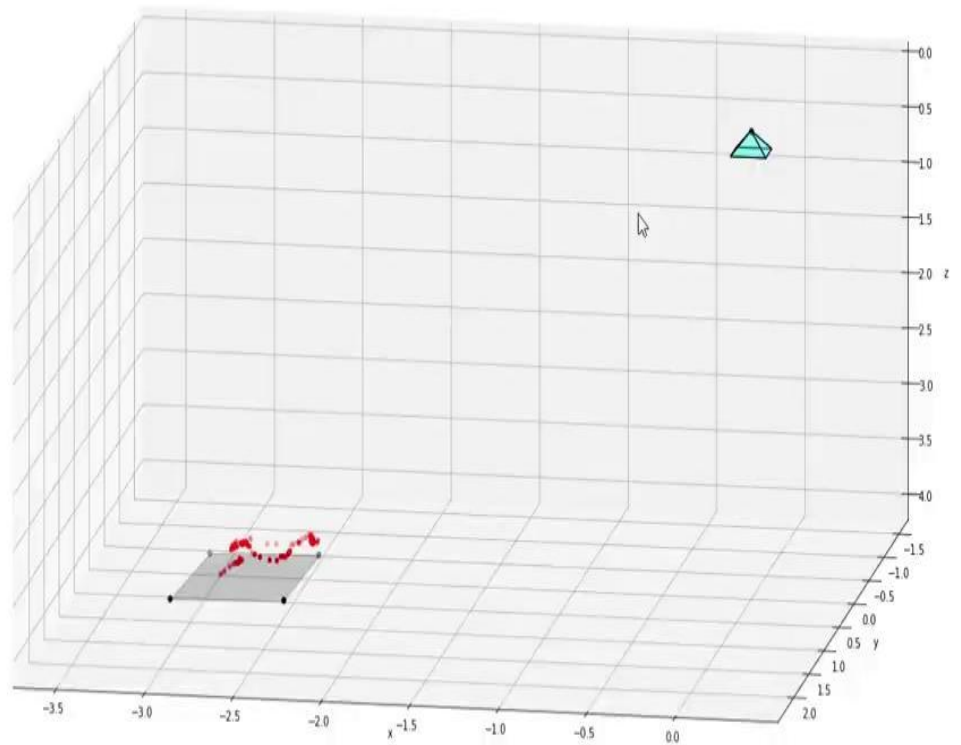
$$R = \begin{bmatrix} 1 & 0.015 & -0.010 \\ -0.014 & 1 & 0.025 \\ 0.011 & -0.025 & 1 \end{bmatrix}$$

$$t = \begin{bmatrix} 0.08 \\ 0.99 \\ -0.11 \end{bmatrix}$$

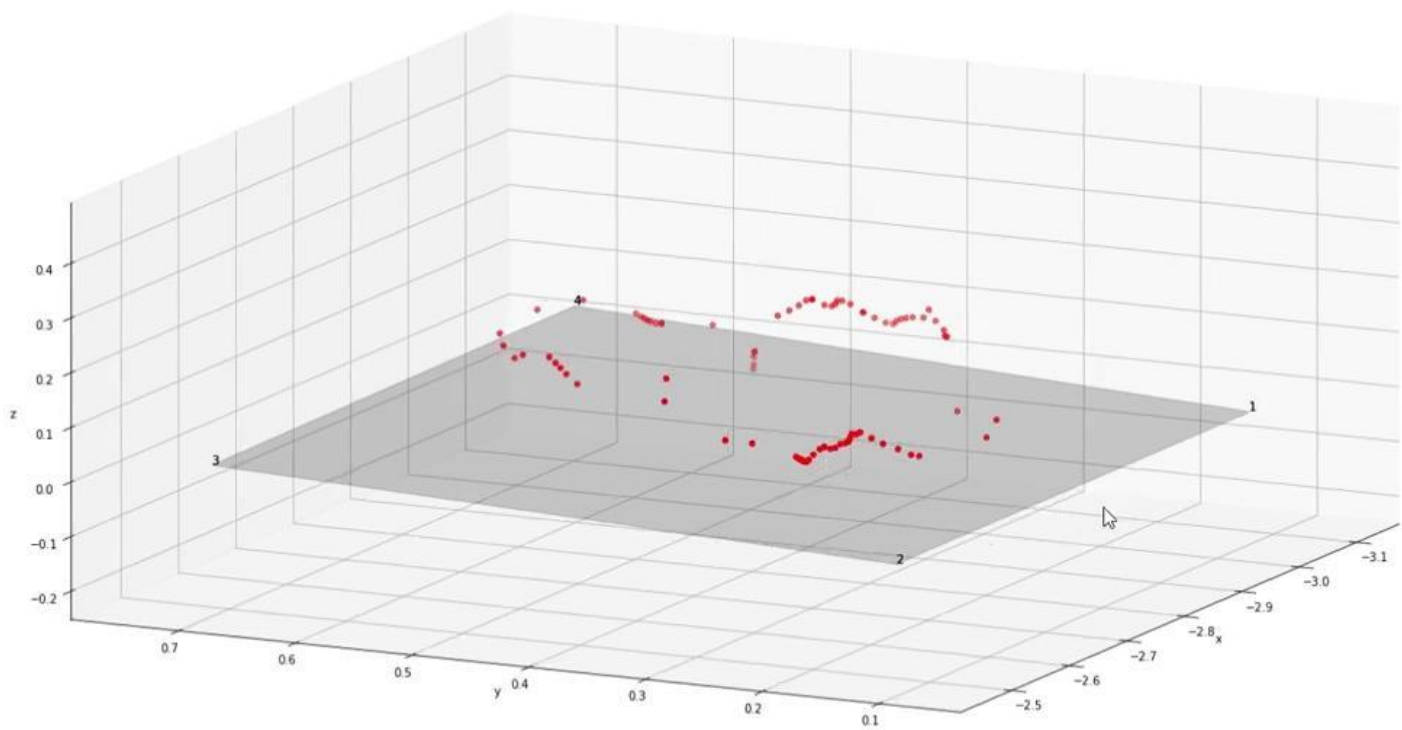


3D Contour  
Reconstruction



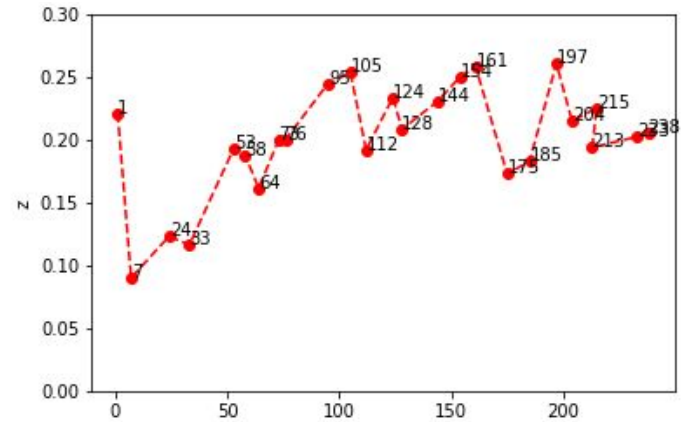
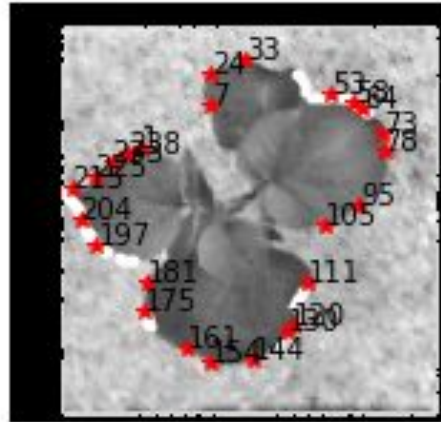
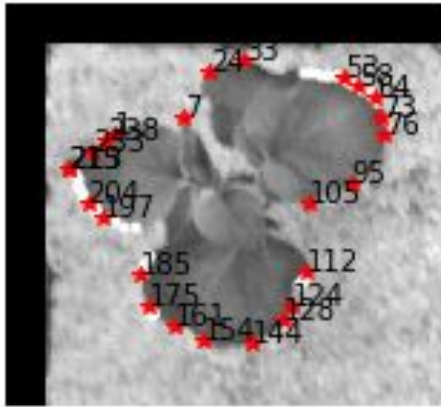






x=-3.14622 , y=0.269611 , z=-0.259866

# 3D Contour reconstruction



# Observation of growth

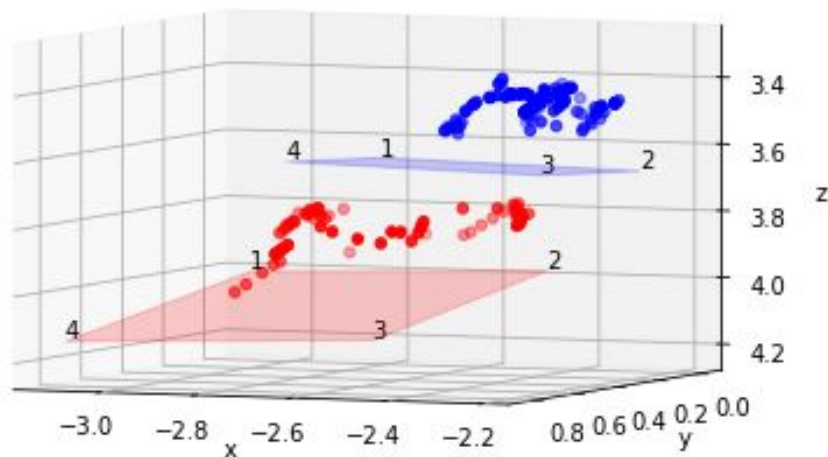


3D Contour  
Reconstruction



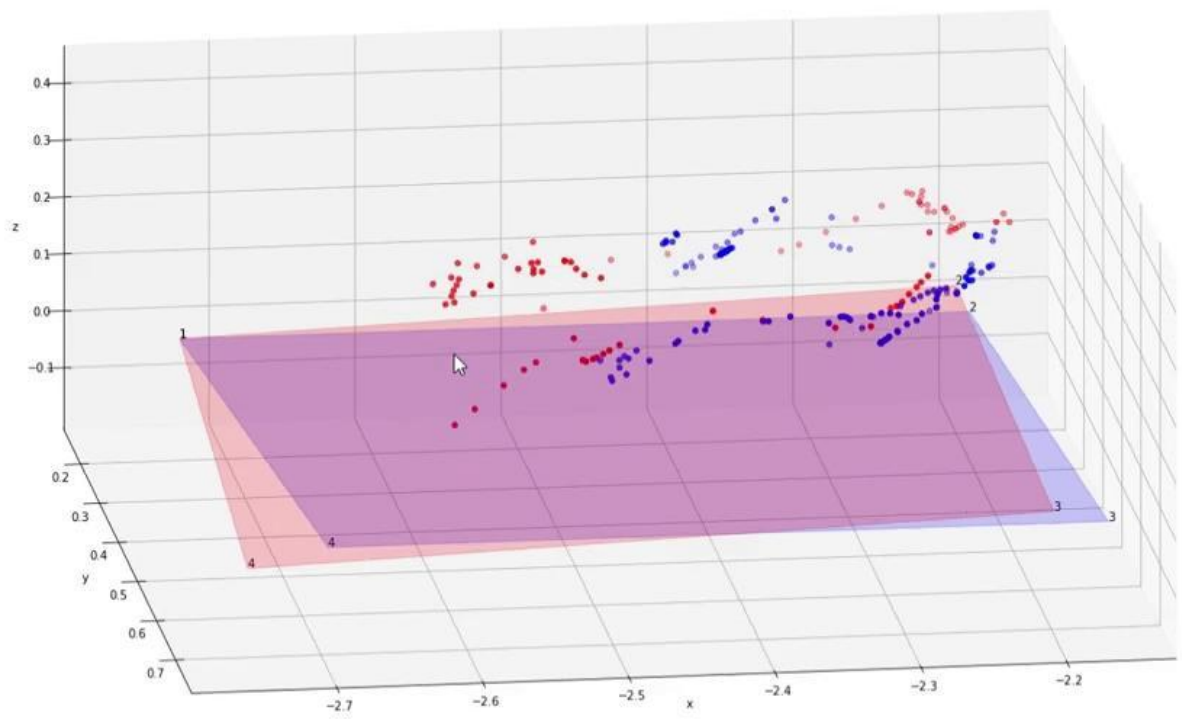
# Misalignment

- $\mathbf{r}'(\tau_1) = aR\mathbf{r}'(\tau_2) + \mathbf{t}$
- Alignment based solely on corner points



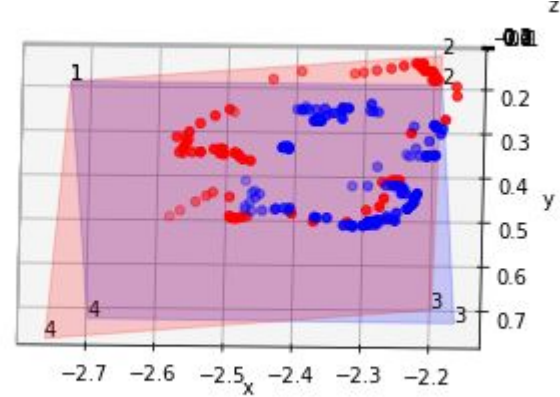
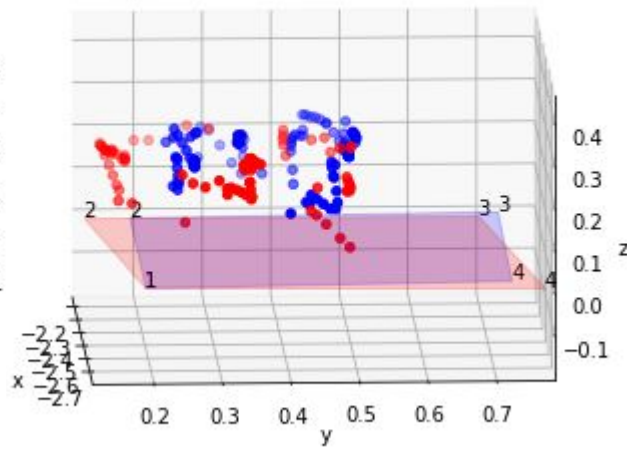
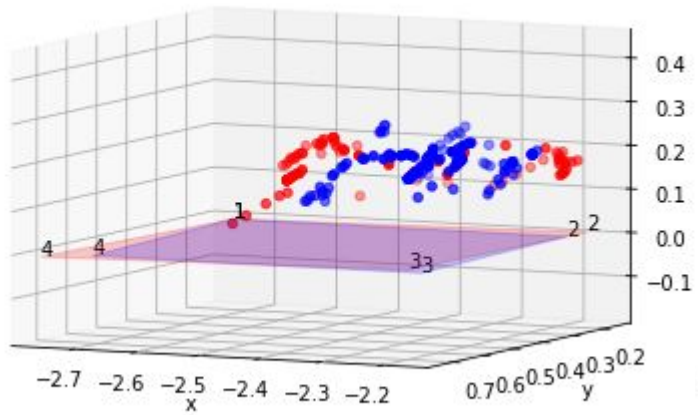
3D Contour  
Reconstruction







# Observation of growth



# 3D Reconstruction

- 3D reconstruction up to scale
- In camera reference frame
- SVD
- Alignment of plant's container

# Double Snakes

- Combine previous problems
- Evolve both snakes simultaneously and match points on contour
- Minimize the functional

$$\Phi[\mathbf{y}_1, \mathbf{y}_2] = \int_0^1 \left[ \frac{\alpha}{2} (\|\mathbf{y}'_1\|^2 + \|\mathbf{y}'_2\|^2) + \frac{\beta}{2} (\|\mathbf{y}''_1\|^2 + \|\mathbf{y}''_2\|^2) + \frac{1}{2} (\mathbf{y}_2^T E \mathbf{y}_1)^2 \right] ds$$

$$\frac{\partial \mathbf{y}_1}{\partial t} = \alpha \frac{\partial^2 \mathbf{y}_1}{\partial s^2} - \beta \frac{\partial^4 \mathbf{y}_1}{\partial s^4} - E^T \mathbf{y}_2 \mathbf{y}_2^T E \mathbf{y}_1 + \mathbf{F}(\mathbf{y}_1),$$

$$\frac{\partial \mathbf{y}_2}{\partial t} = \alpha \frac{\partial^2 \mathbf{y}_2}{\partial s^2} - \beta \frac{\partial^4 \mathbf{y}_2}{\partial s^4} - E \mathbf{y}_1 \mathbf{y}_1^T E^T \mathbf{y}_2 + \mathbf{F}(\mathbf{y}_2).$$

Double Snakes



# Convexity

- When does  $(\mathbf{y}'')^T E \mathbf{y}' = 0$  hold?
- $(\mathbf{y}'')^T E = \mathbf{0}$   
 $E \mathbf{y}' = \mathbf{0}$

The left and right nullspace of  $E$

Double Snakes



# Double Snakes

- Simultaneously updating both contours
- Convexity of the essential matrix

# Conclusion

- GVF-Snakes algorithm to detect contour
- 3D contour reconstruction up to scale
- Observation of growth due to alignment of plant's container
- Introduction of Double Snakes

Conclusion





# Recommendations

- Stabilizing the estimation of the essential matrix
- More advanced camera model
- Improve alignment of plant's container
- Explore convexity of the coplanarity constraint

Conclusion



Questions?