

How to optimize a polarization imager for target detection?

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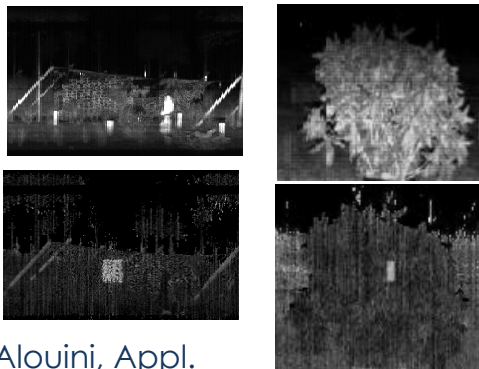
PT&T 2014, March 24th-28th

Our approach of polarimetric imaging

Objective: to reveal contrasts that do not appear in classical images by using **polarization properties** of scenes

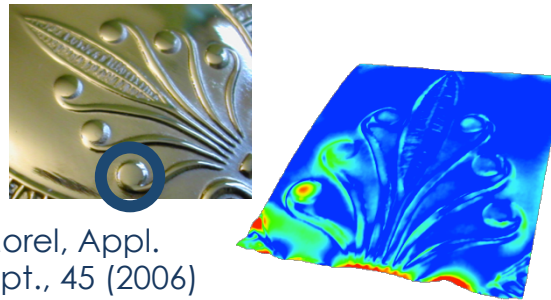
Transform **differences of polarimetric properties** into **differences of graylevel**

Decamouflaging



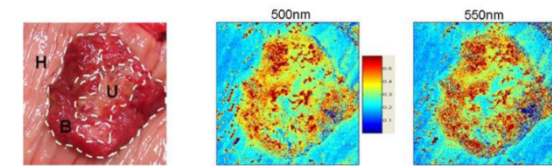
Alouini, Appl. Opt., 48 (2009)

Industrial inspection

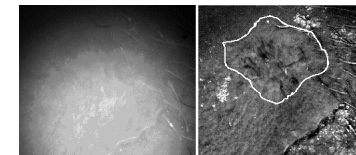


Morel, Appl. Opt., 45 (2006)

Tumor detection



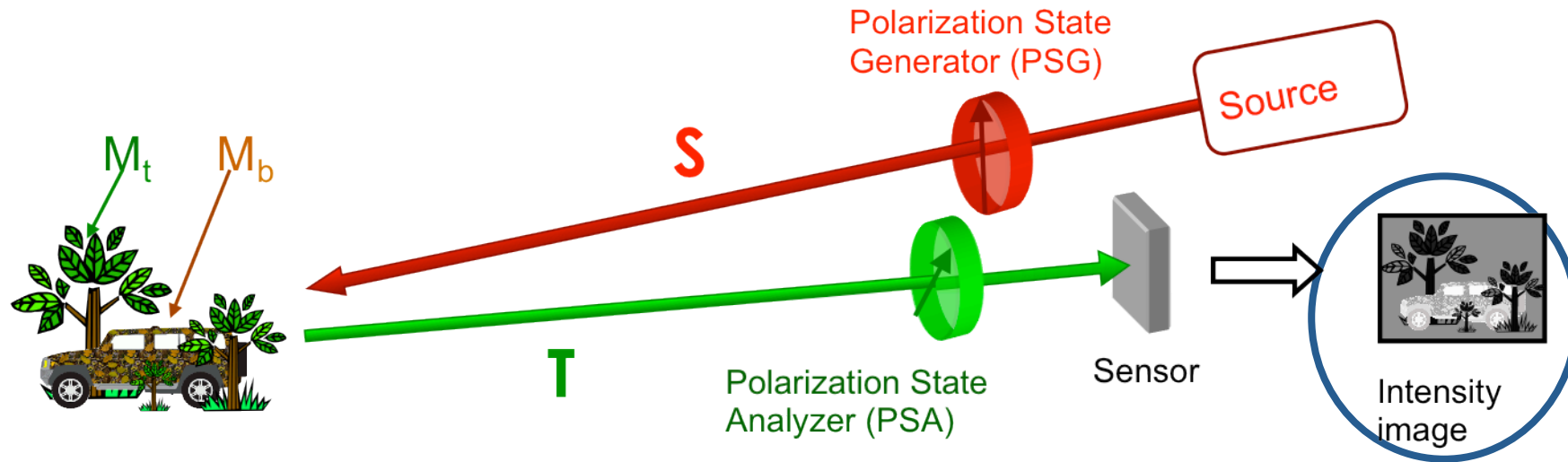
Pierangelo, Opt. Express, 19 (2011)



Jacques, J. Biomed. Opt., 7 (2002)

- our goal is not to measure polarimetric properties but to **optimize a contrast**
- **Adaptive polarimetric imaging:** adapt the polarization states to the observed scene and its variations

Active polarimetric imaging system



Intensity collected by one pixel seeing **Mueller matrix** M_u : $i_u \propto \mathbf{T}^t \mathbf{M}_u \mathbf{S}$

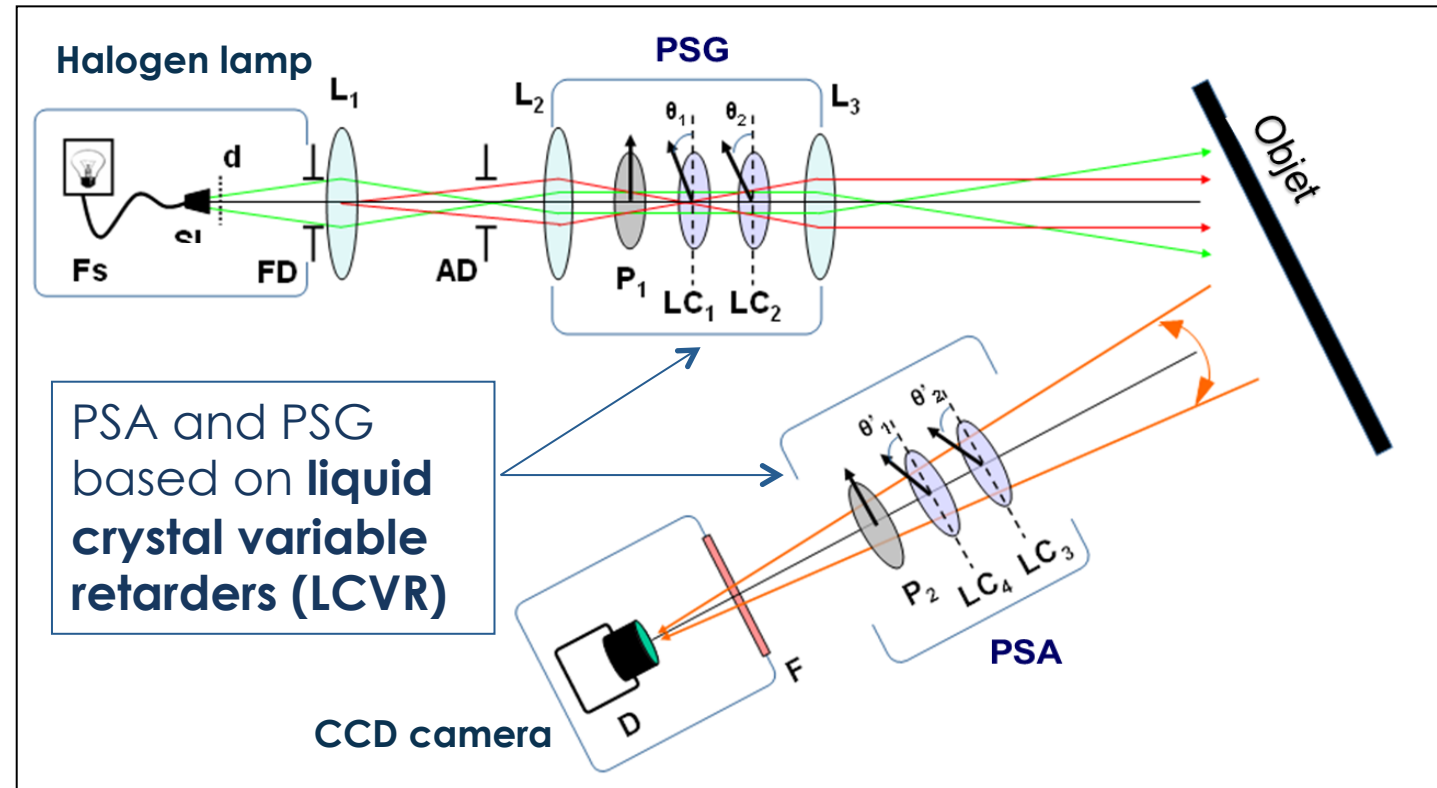
Objective: optimize the contrast by adjusting PSA and PSG

- How to build such an adaptive polarization imager?
- How to find the optimal states of PSG and PSA?
- What if nothing is known about the scene?

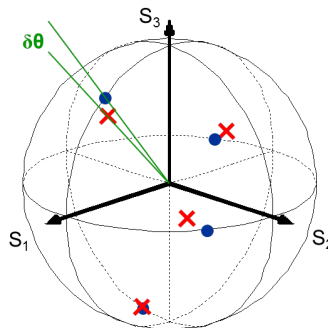
How to build a adaptive polarization imager?



Matlab+Labview

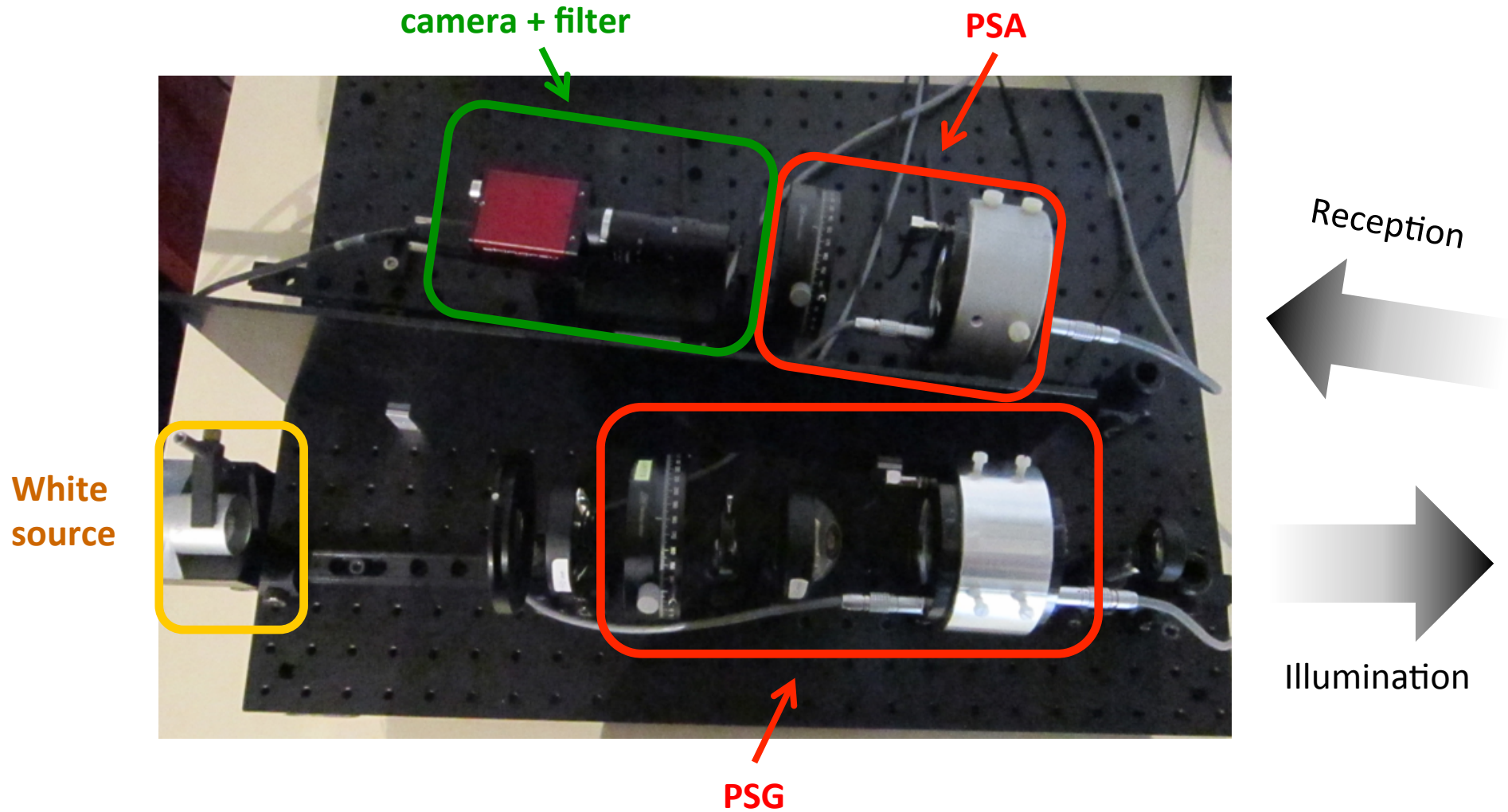


Accuracy: 2° on Poincaré sphere



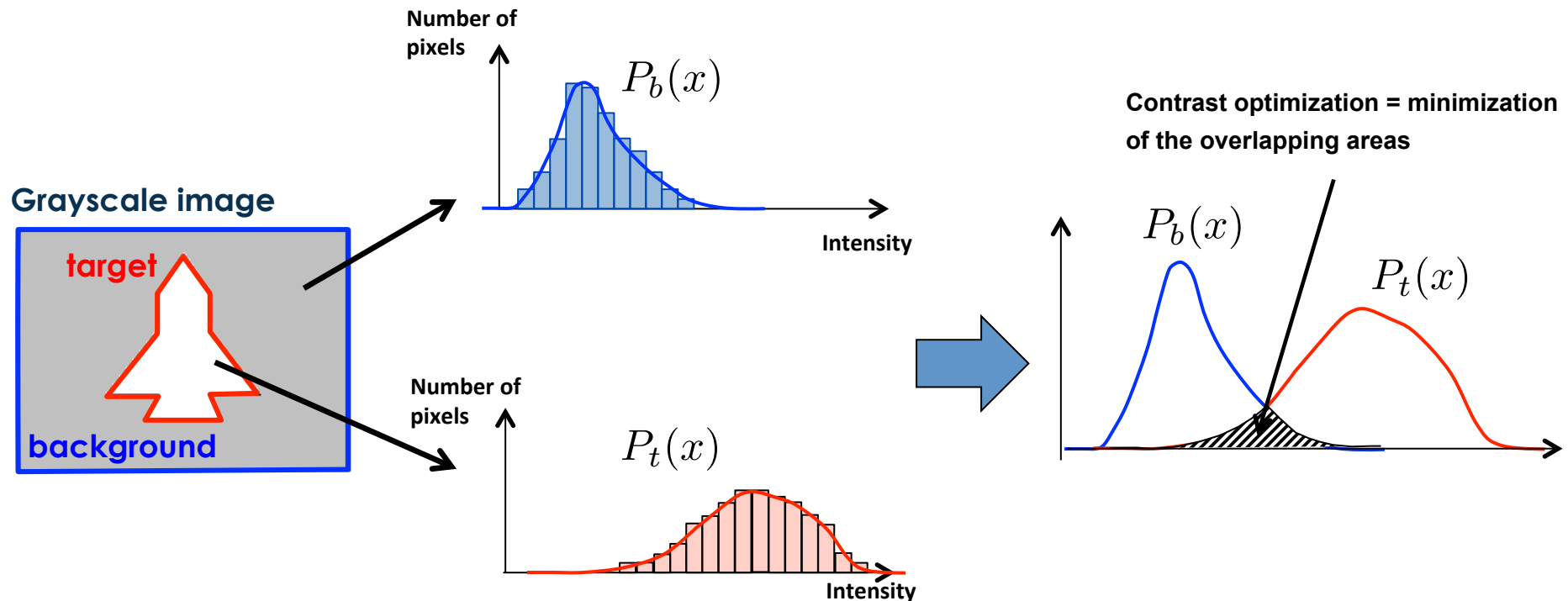
Response time: 100ms

How to build a adaptive polarization imaging?



How to define the contrast of a image?

Statistical definition of the contrast:

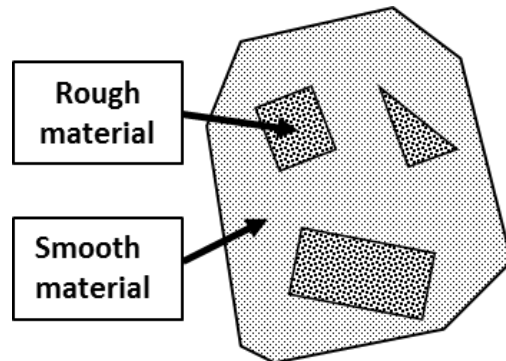


Bhattacharyya distance: optimizing the contrast = maximizing the distance

$$\mathcal{B}_{t,b} = -\log \left[\int_{\mathcal{D}} [P_t(x)P_b(x)]^{1/2} dx \right] = \mathcal{B}_{t,b}(\mathbf{S}, \mathbf{T}, \mathbf{M})$$

How to find the optimal polarization states?

Scheme of the scene



Painted with the same paint

Same painting

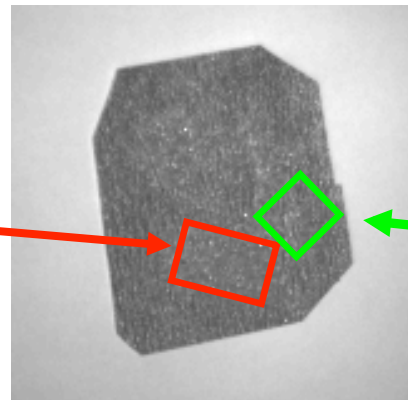
↪ No contrast on intensity images

Different roughness

↪ Different Mueller matrix

Intensity image

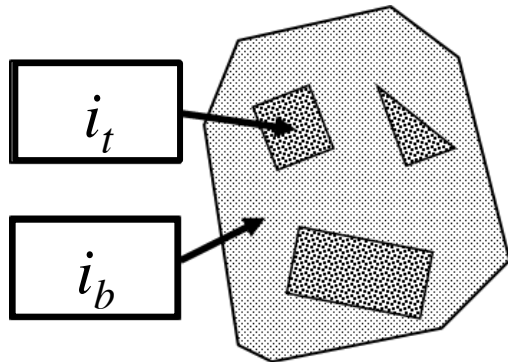
$$M_a = \begin{bmatrix} 1 & -0.01 & 0.01 & -0.00 \\ -0.03 & 0.25 & -0.04 & 0.03 \\ 0.04 & -0.02 & -0.27 & 0.01 \\ 0.02 & 0.01 & 0.03 & -0.19 \end{bmatrix}$$



$$M_b = \begin{bmatrix} 1 & -0.02 & 0.01 & -0.00 \\ -0.02 & 0.56 & -0.10 & 0.05 \\ 0.05 & -0.03 & -0.61 & 0.02 \\ 0.00 & 0.00 & 0.10 & -0.53 \end{bmatrix}$$

How to find the optimal polarization states?

Scheme of the scene



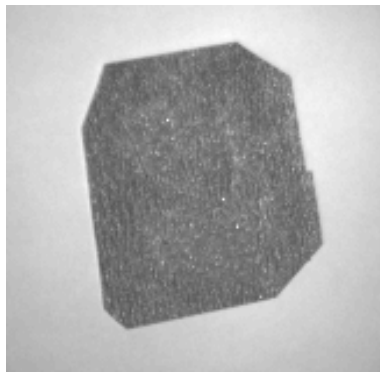
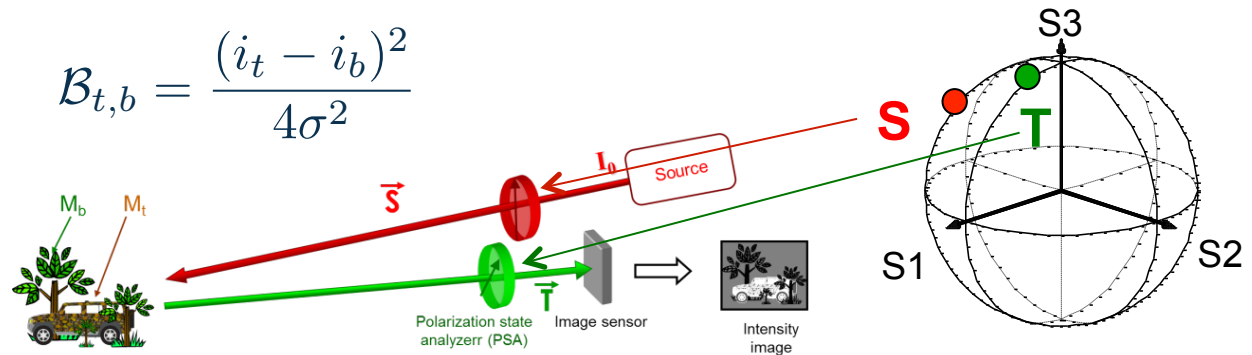
Optimal PSA and PSG states

under hypothesis of **additive Gaussian noise**

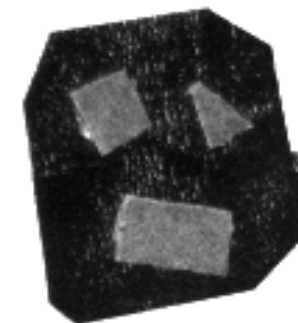
$$B_{t,b} = \frac{(i_t - i_b)^2}{4\sigma^2}$$

$$(\alpha_s, \varepsilon_s) = (-35^\circ, 15^\circ)$$

$$(\alpha_T, \varepsilon_T) = (-60^\circ, 15^\circ)$$



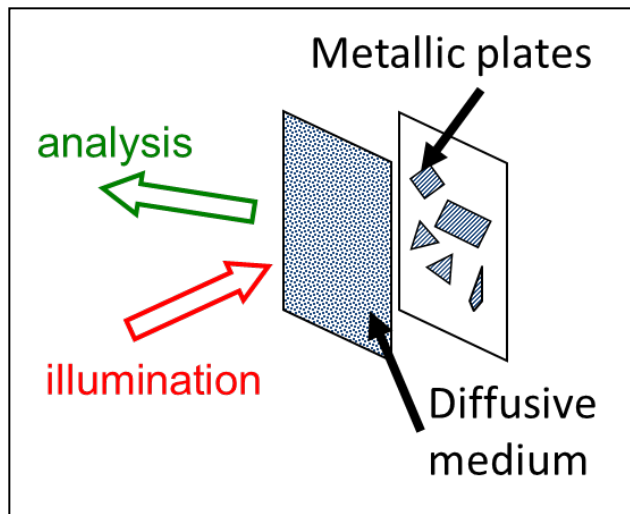
Standard intensity image



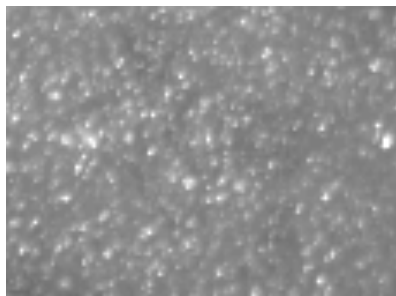
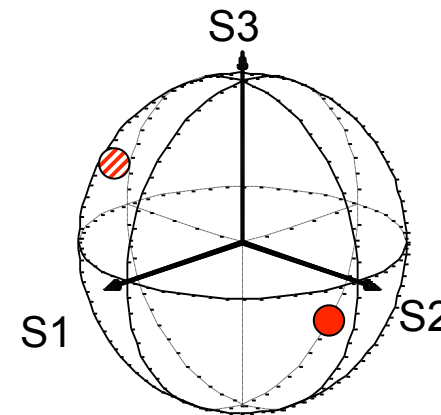
Polarimetric image with optimal states

Impact of the noise model

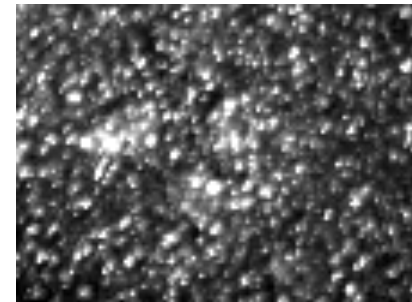
Scheme of the scene



Optimal PSA and PSG state under hypothesis of **additive** **Gaussian noise**



Standard
Intensity image

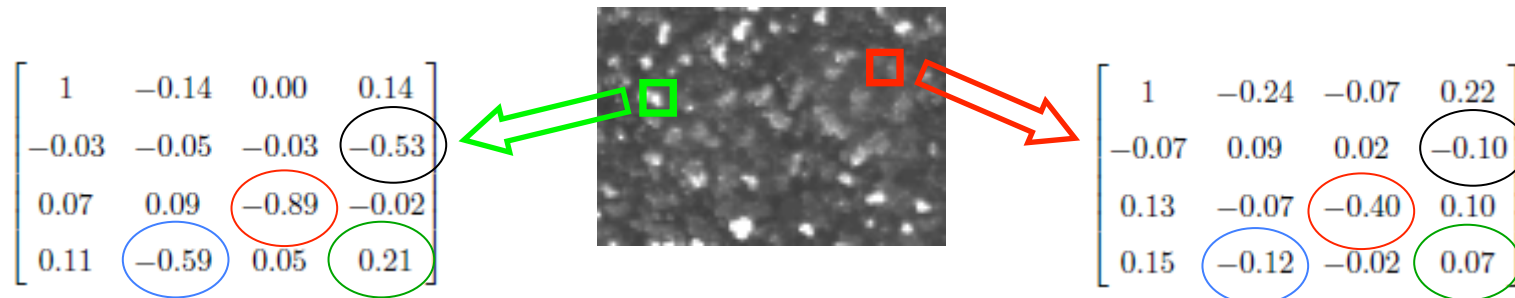


Polarimetric image

The noise
model is **not**
adapted

Spatial fluctuations of the Mueller matrix

Noise is due to **spatial fluctuations of polarimetric properties!**



The Mueller matrix of each pixel is modeled as a Gaussian random vector

$$\mathbf{V}_M = [M_{00} \ M_{01} \ \cdots \ M_{33}]^t$$

defined by its:

Average $\mathbf{V}_{\langle M \rangle} = \langle \mathbf{V}_M \rangle$

Covariance matrix $G = \langle (\mathbf{V}_M - \mathbf{V}_{\langle M \rangle})(\mathbf{V}_M - \mathbf{V}_{\langle M \rangle})^t \rangle$

Expression of the contrast in the presence of spatial fluctuations of the Mueller matrix

Statistics of the intensity image: Gaussian random variables defined by

$$\text{Average } i_u = I_0 [\mathbf{T} \otimes \mathbf{S}]^t \mathbf{V}_{\langle M_u \rangle} \quad \otimes : \text{Kronecker product}$$

$$\text{Variance: } \sigma_u^2 = I_0^2 [\mathbf{T} \otimes \mathbf{S}]^t G_u [\mathbf{T} \otimes \mathbf{S}]$$

Bhattacharyya distance:

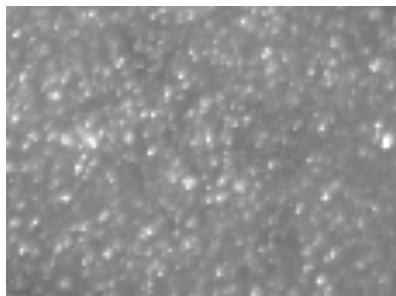
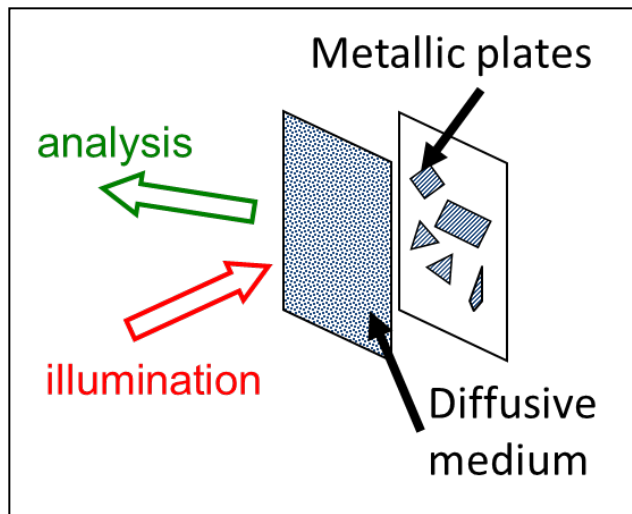
$$B_{t,b} = \frac{(i_t - i_b)^2}{\sigma_t^2 + \sigma_b^2 + \sigma^2} = \frac{(\mathbf{T} \Delta M \mathbf{S})^2}{[\mathbf{T} \otimes \mathbf{S}]^t G_{\text{fluct}} [\mathbf{T} \otimes \mathbf{S}] + \text{SNR}^{-1}}$$

Spatial fluctuations Additive noise

with $\Delta M = M_t - M_b$ $G_{\text{fluct}} = G_t - G_b$ $\text{SNR} = I_0^2 / \sigma^2$

Optimal states

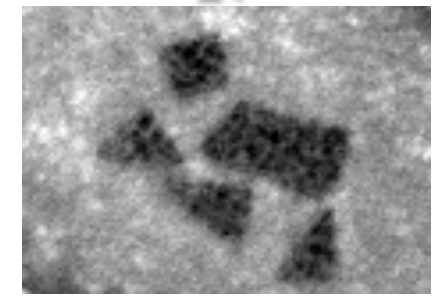
Scheme of the scene



**Standard
Intensity image**

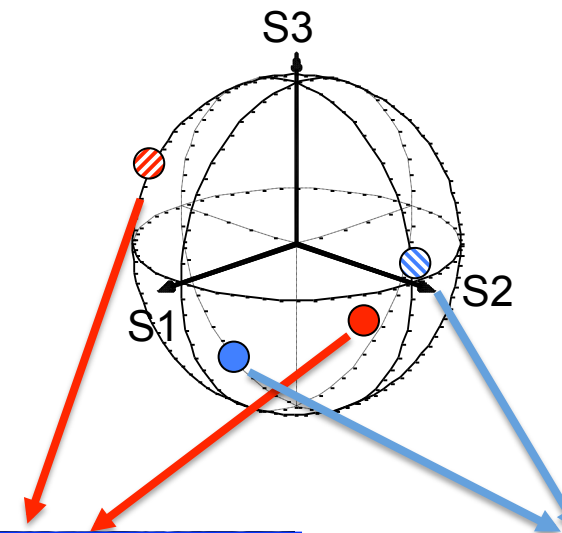


Importance of
accurate
physical
modeling of
the scene



**Polarimetric image
with optimal states**

Optimal PSA and PSG states under hypothesis of **spatial fluctuations of polarimetric properties**



Limitations of this approach

To **optimize the contrast**, one needs to know the *polarimetric properties* of the two regions of interest.

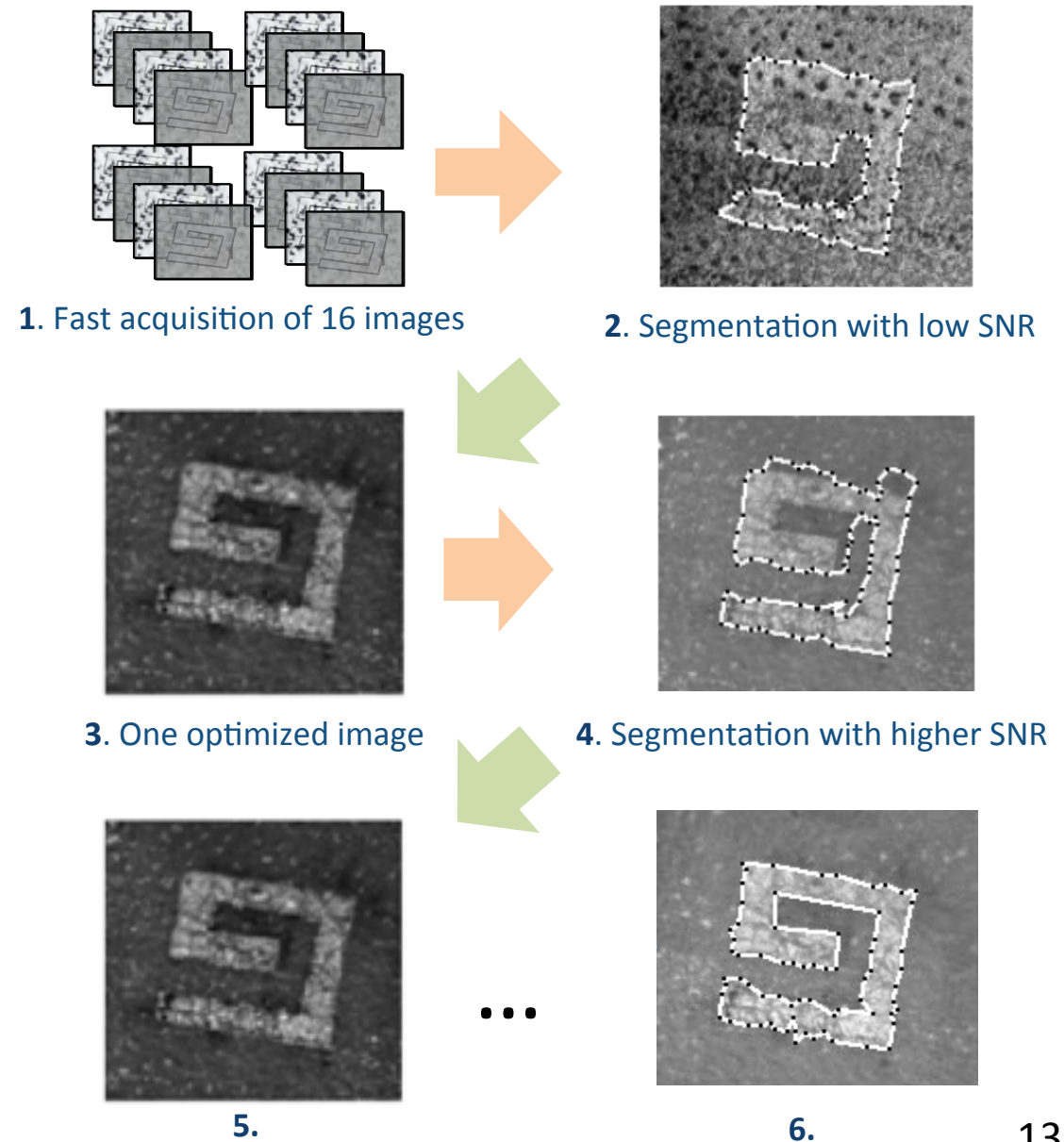
What if we do not know this information?

Proposed solution:

Combining the capabilities of adaptive polarimetric system with **image processing algorithms**

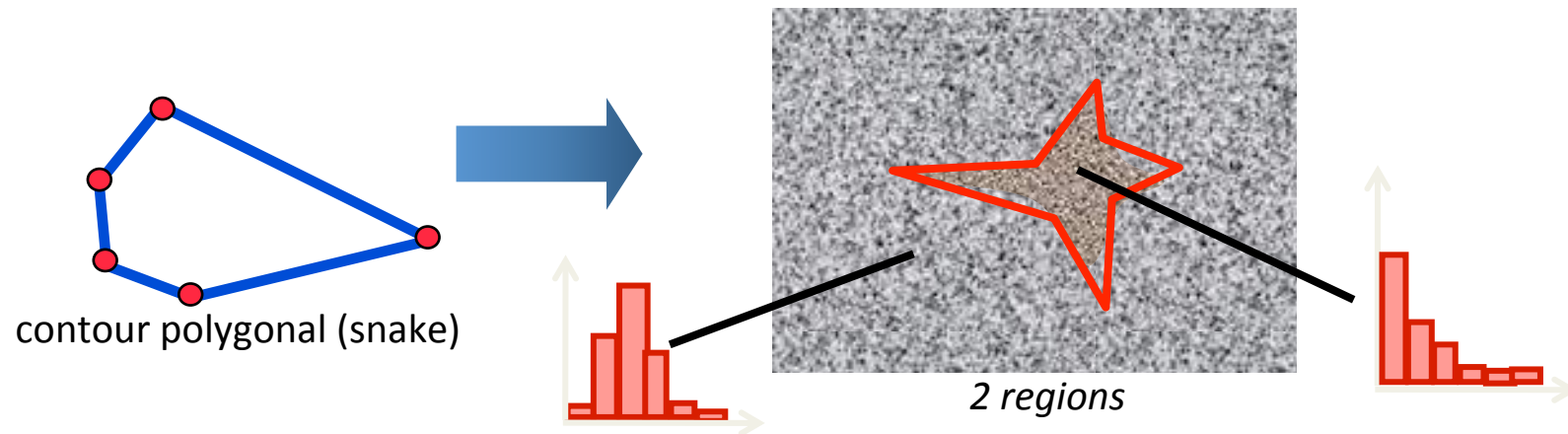
Automation of the method

1. Acquisition of the Mueller image of the scene (16D)
2. Segmentation: separate an object from the background
3. Estimation of the optimal T and S and acquisition of the optimized image
4. Segmentation of the optimized image
5. Re-iteration of steps 3 and 4
6. Final image



Choice of the segmentation algorithm

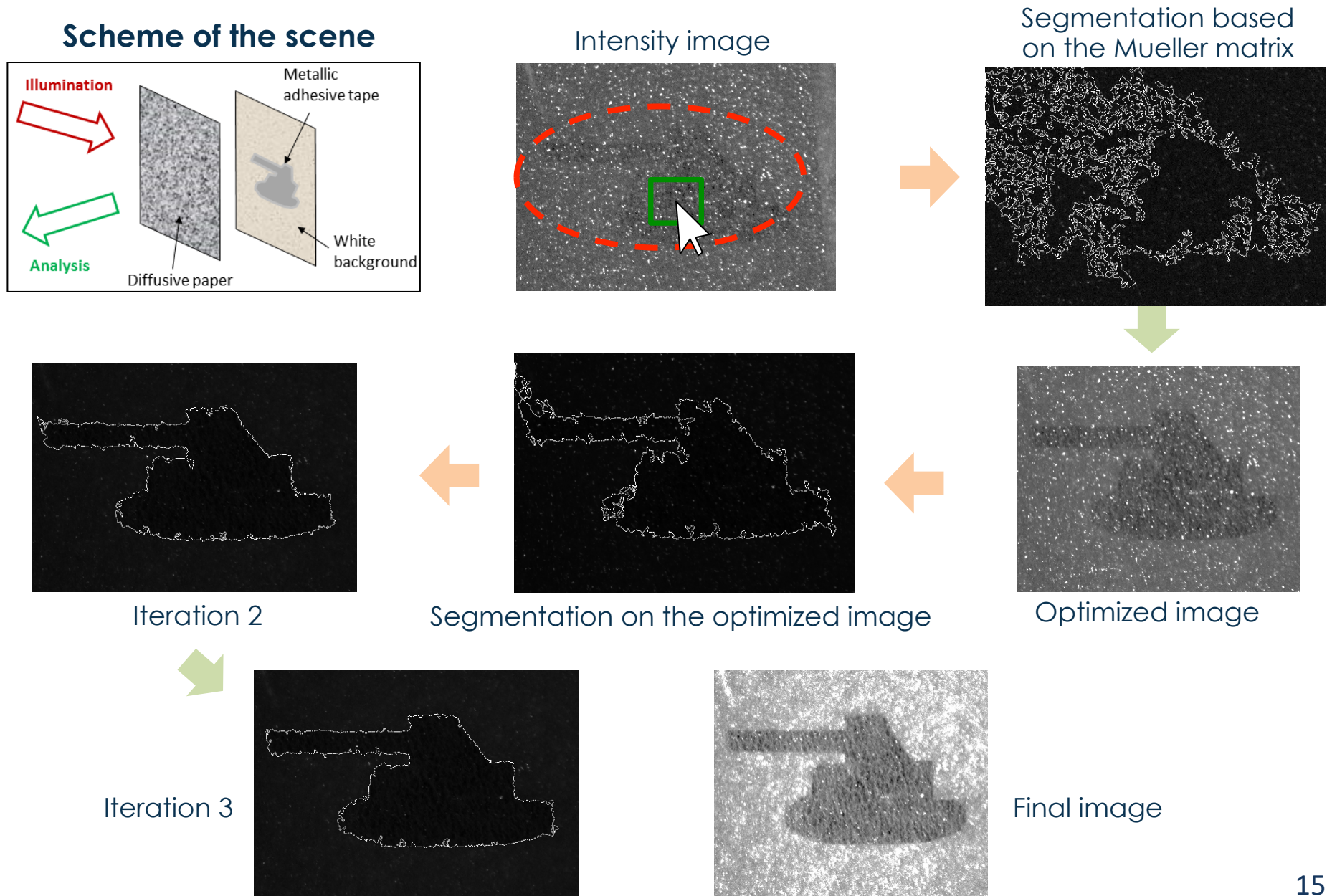
Statistical polygonal active contour



- ↪ No parameter to adjust in the method
- ↪ Adapted to low SNR images
- ↪ Fast : computation time < 10 ms on 256 x 256 images



Illustration of the complete procedure



Conclusions & perspectives

- **Contrast optimization using adaptive imager**
 - ↳ Generation/analysis of any polarization state with **good accuracy** ($\sim 2^\circ$)
 - ↳ **video frame rate**
- **Polarimetric imaging controlled by image processing algorithm**
 - ↳ **(semi-)automation** of contrast optimization without **a priori knowledge** on target polarimetric properties

ONGOING/FUTURE WORKS

- **SWIR imager for decamouflaging applications**
- **Explore other applications**



Thank you for your attention
