

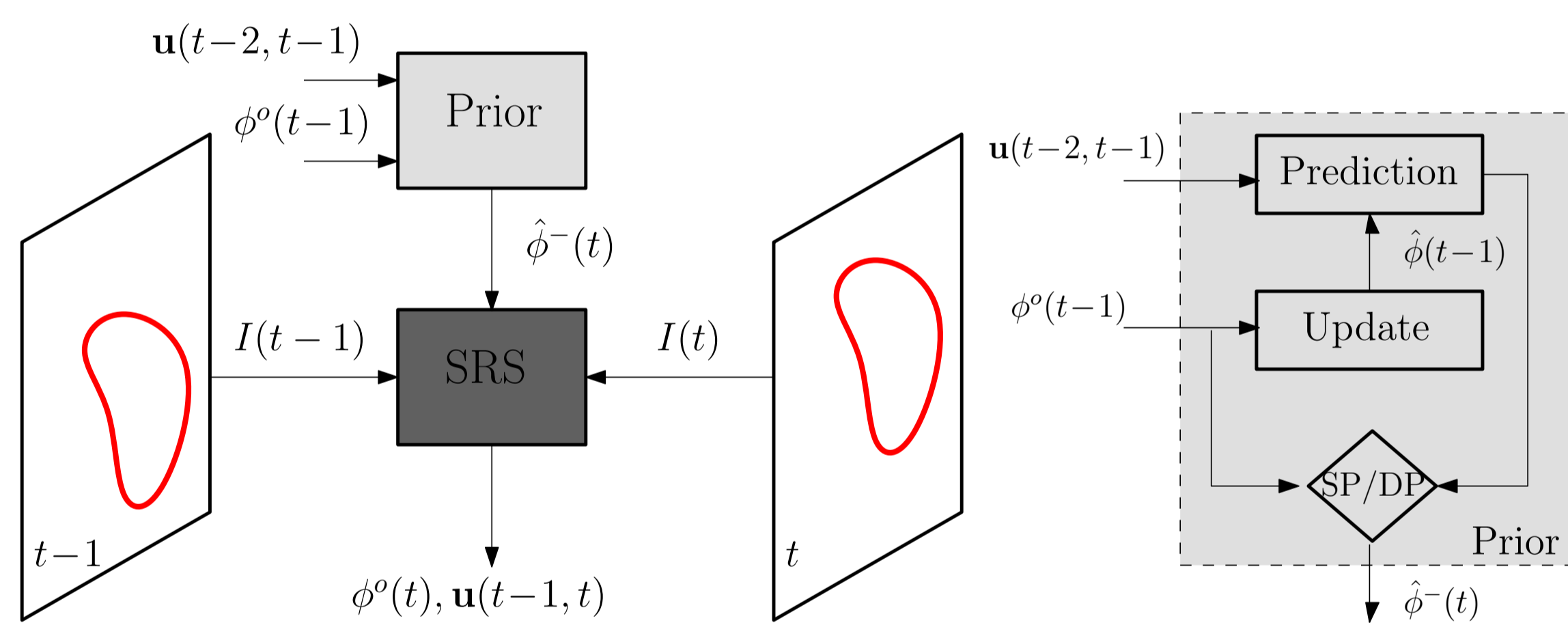
Problem Definition and Contributions

We consider the problem of **segmentation and tracking** in time sequence imagery. The key components are:

- A **generalized model**, robust to non-uniform additive **illumination changes**, is introduced for the level-set based simultaneous registration and segmentation (SRS) framework.
- The new model is derived in a principled way by formulating the SRS problem in a maximum *a posteriori* (MAP) framework.
- Multiple cues** are tightly coupled for establishing correspondence and also compensates for the illumination changes over consecutive views.

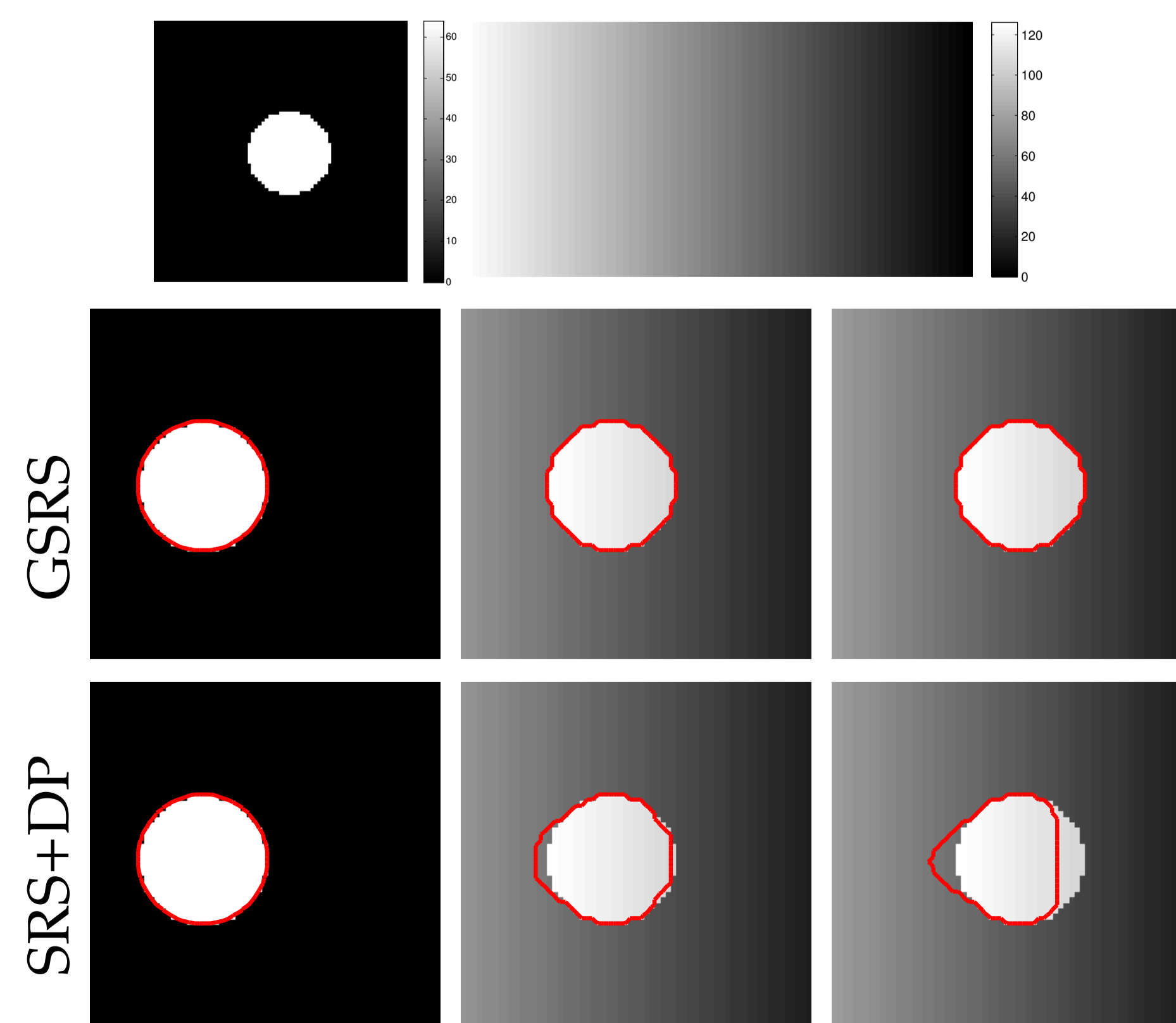
Overview of SRS+DP, Ghosh et al. ICCV'09

- SRS+DP simultaneously finds the segmentation for the current frame while estimating the registration between consecutive frames.
- The **dynamic prior** is designed based on all the past observations, thus providing robustness against noise/clutter.



Limitation of SRS+DP

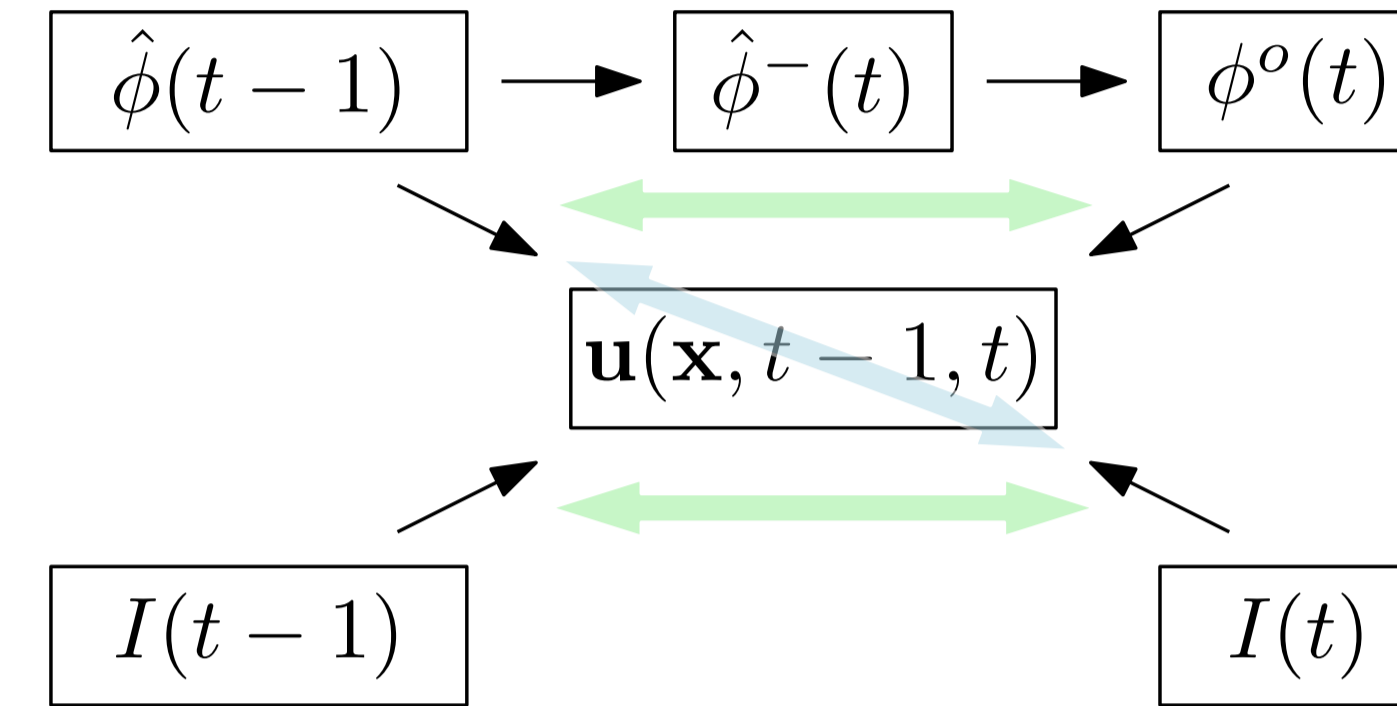
- SRS+DP** model can not handle for the **constant/linear illumination** change or other intensity based disturbances.



Proposed Model (GSRS)

- Consider the following maximum *a posteriori* (MAP) problem:
$$\langle u_{\text{opt}}, \phi_{\text{opt}}^o(t) \rangle = \underset{\mathbf{u}, \phi^o(t)}{\text{argmax}} \mathcal{P}(\mathbf{u}, \phi^o(t) | I(t), I(t-1), \hat{\phi}^-(t-1), \hat{\phi}^-(t))$$

- Dependency** between different variables:



- MAP estimate can be obtained as:

$$\langle u_{\text{opt}}, \phi_{\text{opt}}^o(t) \rangle = \underset{\mathbf{u}, \phi^o(t)}{\text{argmax}} \underbrace{\mathcal{P}(\mathbf{u} | \phi^o(t), I(t), I(t-1), \hat{\phi}^-(t-1))}_{\mathcal{A}_1} \underbrace{\mathcal{P}(\phi^o(t) | I(t), \hat{\phi}^-(t))}_{\mathcal{A}_2}$$

Different Approximations for \mathcal{A}_1

- In SRS+DP (separated intensity and shape based cues):

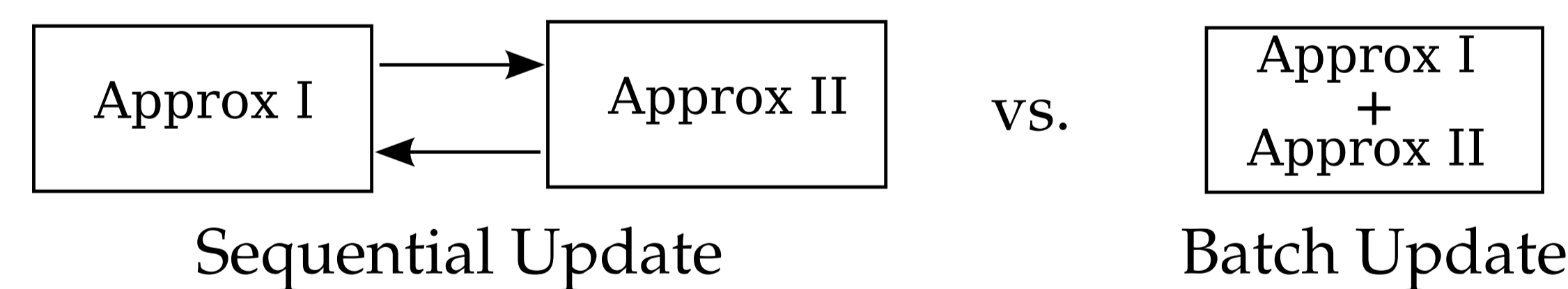
$$\mathcal{A}_1 \approx \mathcal{P}(\mathbf{u} | I(t), I(t-1)) \mathcal{P}(\mathbf{u} | \phi^o(t), \hat{\phi}^-(t-1)) \Rightarrow \text{Approx I}$$

- Other alternative (with cross connection between multiple cues):

$$\mathcal{A}_1 \approx \mathcal{P}(\mathbf{u} | I(t), \hat{\phi}^-(t-1)) \mathcal{P}(\mathbf{u} | \phi^o(t), I(t-1)) \Rightarrow \text{Approx II}$$

- Proposed approximation** (Batch Update):

$$\mathcal{A}_1 \approx \underbrace{\mathcal{P}(\mathbf{u} | I(t), I(t-1))}_{\mathcal{A}_1^1} \underbrace{\mathcal{P}(\mathbf{u} | \phi^o(t), \hat{\phi}^-(t-1))}_{\mathcal{A}_1^2} \underbrace{\mathcal{P}(\mathbf{u} | I(t), \hat{\phi}^-(t-1))}_{\mathcal{A}_1^3}$$



Design of a New Functional and Modification to \mathcal{A}_1^1

- New functional**, \mathcal{A}_1^3 , connecting multiple cues:

$$E(\mathbf{u}; \hat{\phi}^-(t-1), I(t)) = \int_{\Omega} g_{I(t)}(\mathbf{x}) |\nabla H_{\epsilon}(\hat{\phi}(\mathbf{T}(\mathbf{x}), t-1))| dx$$

- Adapted \mathcal{A}_1^1** to cope with the illumination change:

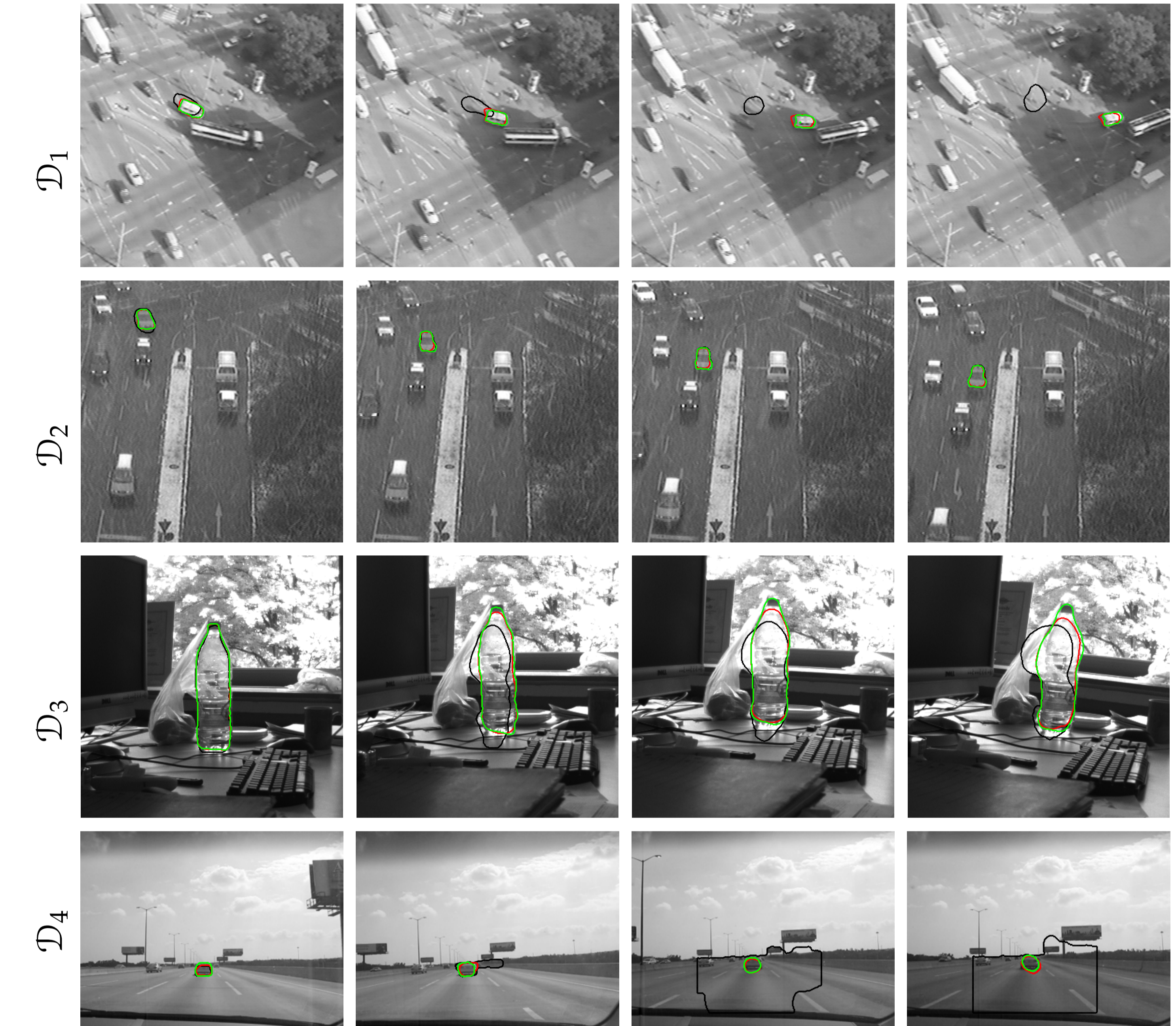
$$\mathcal{A}_1^1 = \frac{1}{2} \int_{\Omega} \left[(I(\mathbf{x}, t) - I(\mathbf{T}(\mathbf{x}), t-1) + \mathcal{G}(\mathbf{x}))^2 + \alpha \text{trace}(\nabla \mathbf{u} \nabla \mathbf{u}^T) + \gamma |\nabla \mathcal{G}|^2 \right] dx$$

- GSRS** Algorithm:

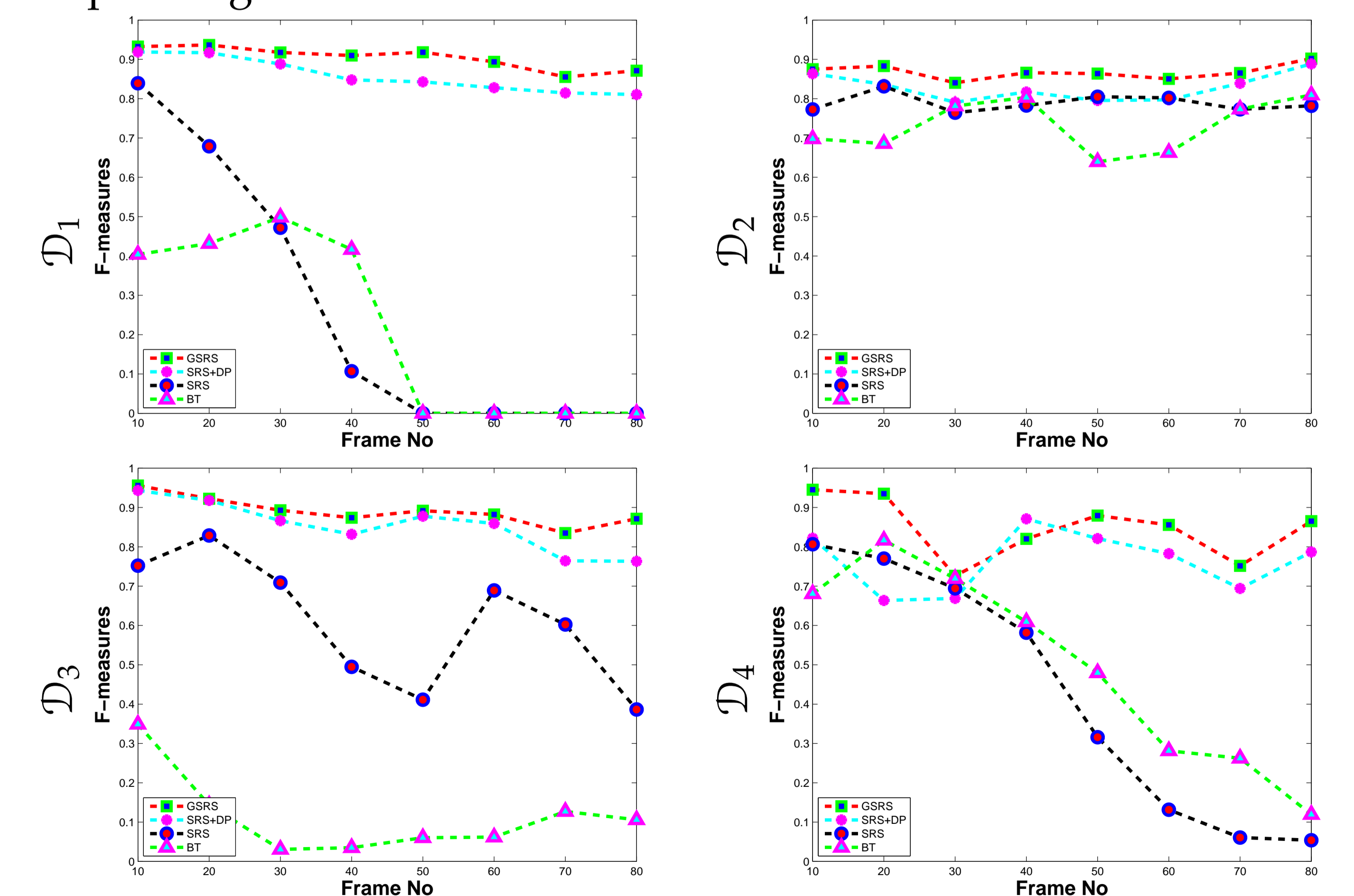
- Optimize over $\mathcal{A} = \mathcal{A}_1^1 \mathcal{A}_1^2 \mathcal{A}_1^3$ (\mathbf{u} related functionals).
- Optimize over $\mathcal{A} = \mathcal{A}_1^2 \mathcal{A}_2$ ($\phi^o(t)$ related functionals).
- Optimize over $\mathcal{A} = \mathcal{A}_1^1$ (\mathcal{G} related functionals).
- Update \mathbf{u} and $\phi^o(t)$ and \mathcal{G} .

Experimental Results and Conclusions

- Each of these test sequences are 80 frames long.



- Corresponding F-measures-



- Conclusions:

- GSRS **tracking** is **invariant** to additive non-uniform illumination changes, inter-reflection and scattering.
- Illumination** changes is explicitly **modeled** as a non-parametric function.
- Additional **complexity** is linear in terms of the size of the image.