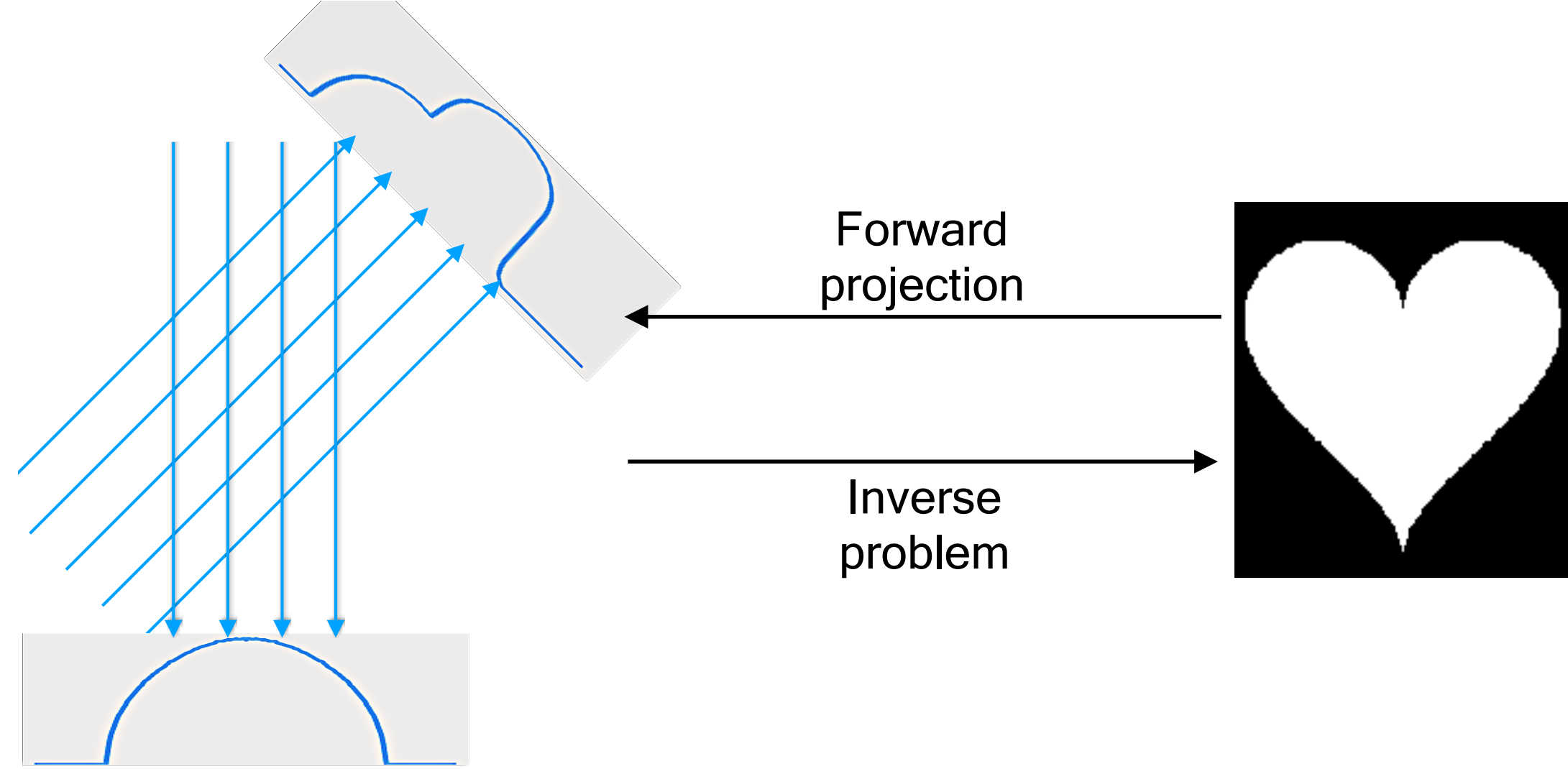


Introduction

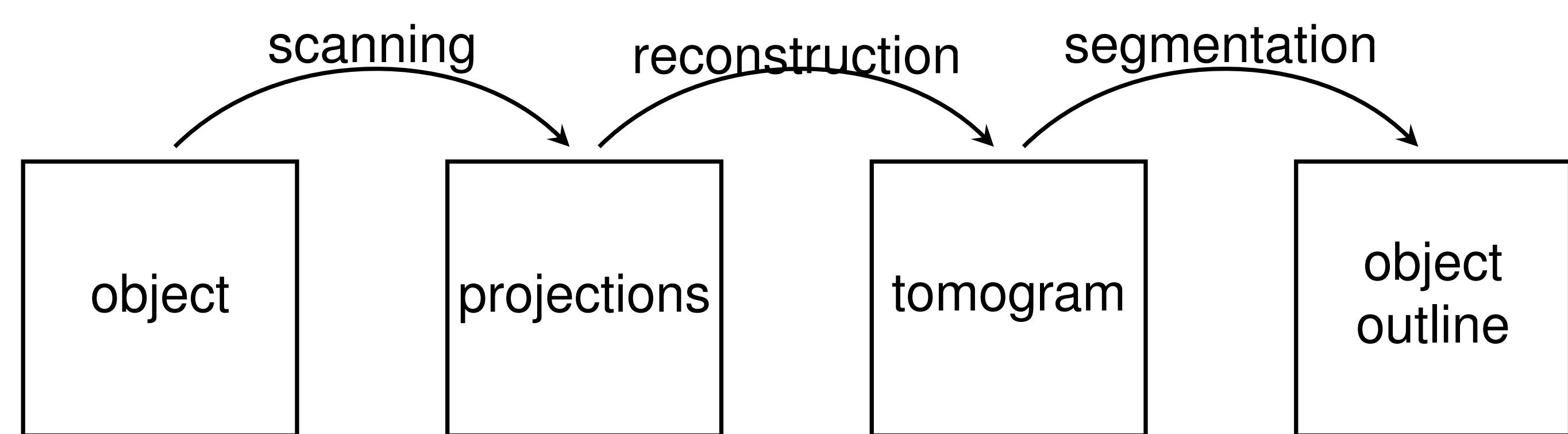
Problem definition: Tomographic reconstruction and segmentation

- ▶ Input: projections (sinogram), e.g., from X-ray.
- ▶ Output: outline of the scanned object



Contribution: We propose a direct segmentation method from projection data. Its main benefit is to directly optimize the error between sinogram data and the segmentation. The segmentation is represented by a parametric curve, which can reduce the pixel artifacts and provide more accurate geometrical measurements than traditional methods.

Existing approaches



Errors are propagated due to multi-step pipeline.

- ▶ As for the reconstruction method, algebraic iterative methods are popular, which are robust to noisy projection data, easy to impose prior knowledge such as non-negativity. But they are still not robust in challenging situations such as limited angle data.
- ▶ As for the segmentation method, level-set method is popular due to topological adaptivity.

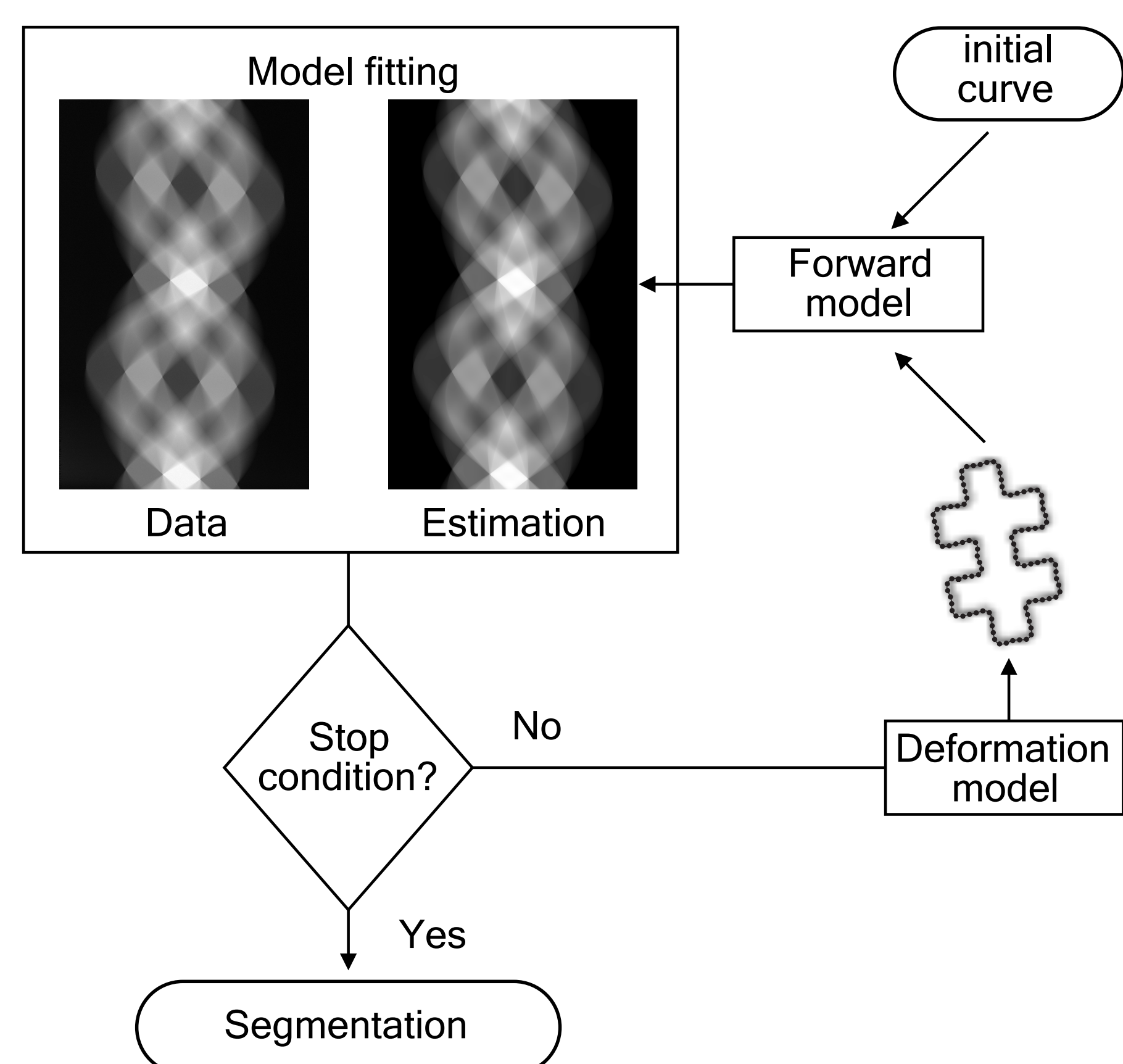
Our approach

Segmentation without reconstruction [1]:

- ▶ Represent the object using a discrete parametric curve in the reconstruction domain.
- ▶ Roughly initialize and then deform the curve to fit projection data.
- ▶ The objects of interest are assumed to have a constant attenuation coefficients.

$$\min E(C, \mu) = \sum_{\theta} \int_S (p(\theta, s) - \mu \hat{p}(\theta, s))^2 ds + R(C)$$

- ▶ p : sinogram data
- ▶ \hat{p} : estimated sinogram by our forward projection
- ▶ μ : Attenuation coefficient (is assumed to be constant)
- ▶ R : regularization (e.g., curve length)



Optimization

Forward model The forward model transforms the curve into sinogram domain:

$$\hat{p}(\theta, s) = \int_{\text{int}(C)} \delta(L_{\theta}(x, y) - s) dx dy$$

where $L_{\theta}(x, y)$ is related to projection ray, given an angle θ and detector location s (e.g., $L_{\theta}(x, y) = x \cos \theta + y \sin \theta$ in parallel beam). We provide an efficient method to calculate forward projection from a curve:

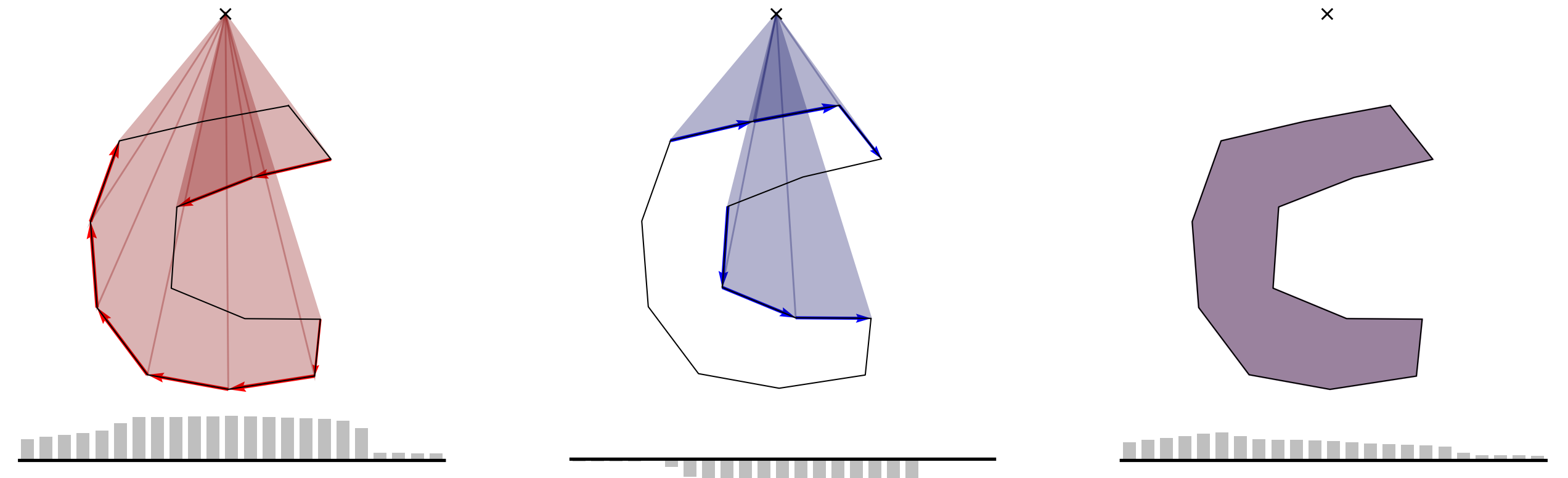


Illustration of forward model with fan beam, given a projection angle θ

Depending on the orientation of the curve, we add [Left] or subtract [Middle] the length of line between the source and the points in the curve. After summing these values, we obtain the final projection values [Right].

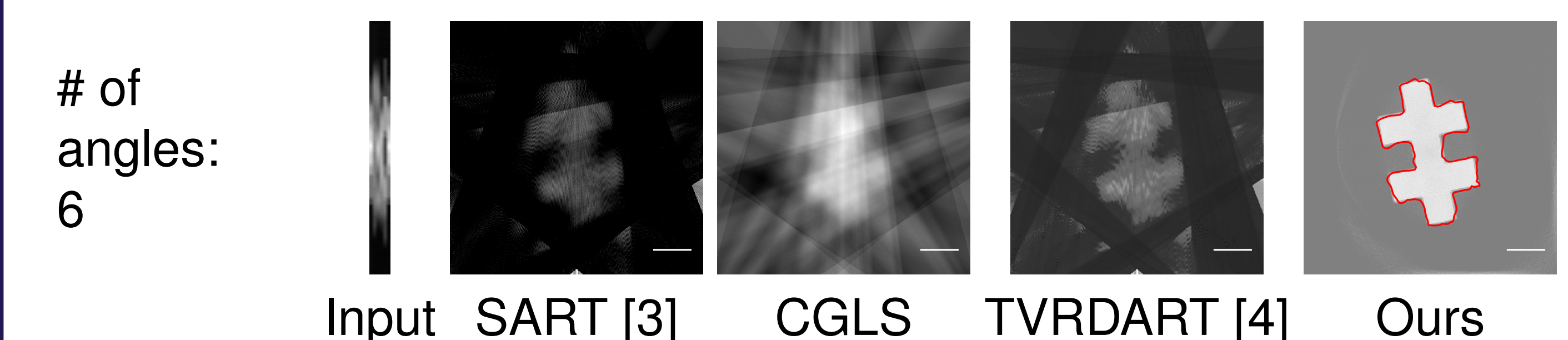
Attenuation coefficient Optimize the energy with respect to μ .

Curve deformation Optimize the energy with respect to the curve. For the data fidelity term, we can derive the equation of curve evolution:

$$C_t(q) = \tau \mu \sum_{\theta} (p - \mu \hat{p})(\theta, L_{\theta}(C(q))) \mathcal{N}(q)$$

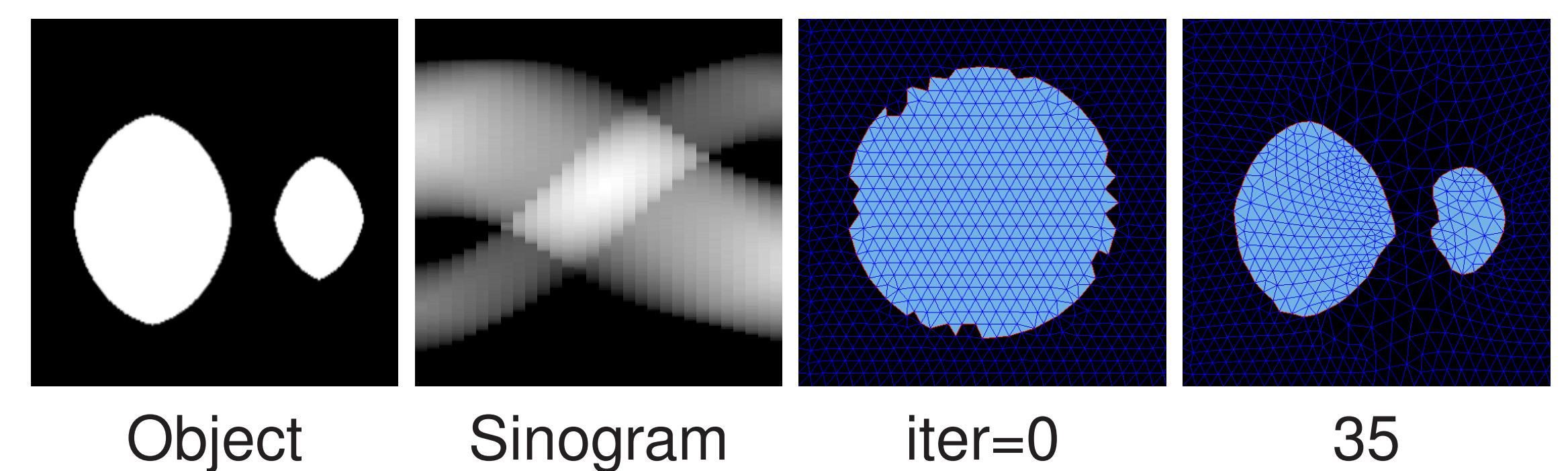
where \mathcal{N} represents the outward unit normal vector and τ the step size.

Experiment on real data with limited angle



Future plan

We are working on applying mesh-based curve representation [2] for supporting multiple materials and topological changes.



We plan to extend to 3D data and dynamic tomography.

References

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- [4] X. Zhuge, W. J. Palenstijn, and K. J. Batenburg, "TVR-DART: A More Robust Algorithm for Discrete Tomography From Limited Projection Data With Automated Gray Value Estimation," TIP 2016.

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