

# Geodesic Tracking of Retinal Vascular Trees with Optical and TV-Flow Enhancement in $SE(2)$

Nicky J. van den Berg\*, Shuhe Zhang†, Bart M.N. Smets\*, Tos T.J.M. Berendschot†, Remco Duits\*

Published in Scale Space and Variational Methods in Computer Vision, Springer 2023

\* Eindhoven University of Technology, Eindhoven, The Netherlands, † Maastricht University, Maastricht, The Netherlands



## Introduction

The retinal vasculature is representative of the vasculature throughout the human body. Consequently, it is beneficial to automate an entire vascular tracking (in 1 run) in retinal images. This poster exhibits our results of the incorporation of optical and TV-flow enhancements in metric tensor fields used for geodesic tracking.

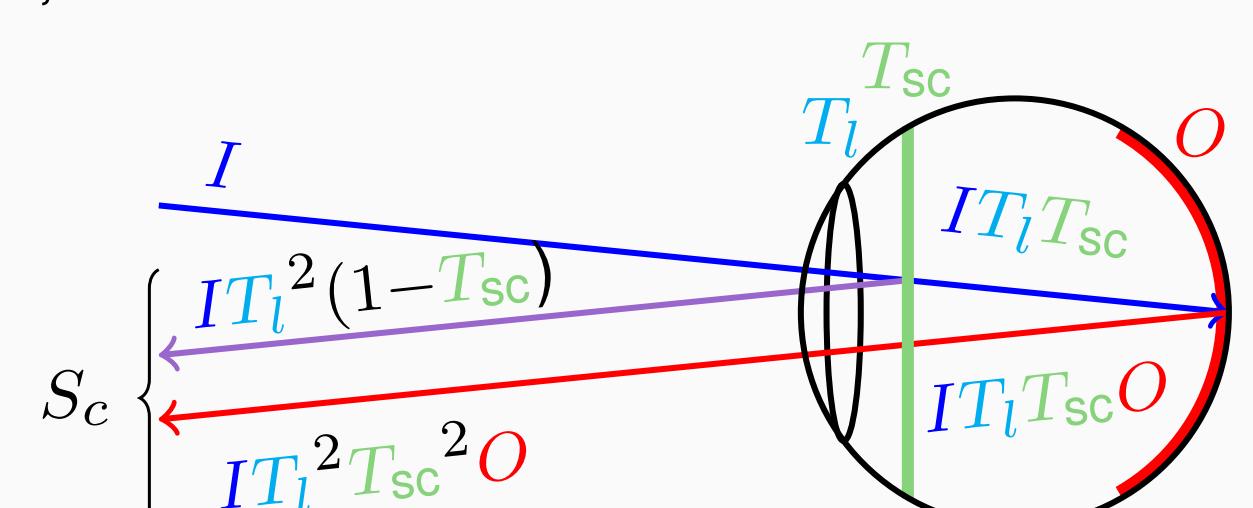
## Orientation Scores

To minimize tracking mistakes at crossings, we lift the image (via cake-wavelet  $\psi$ ) to the space of positions and orientations  $\mathbb{M}_2 := \mathbb{R}^2 \times S^1 \cong \mathbb{R}^2 \times SO(2) =: SE(2)$ .



## Illumination Enhancement of Image

The unprocessed picture  $S$ , taken by the ophthalmologist, may deviate from the actual retinal image  $O$ , which we aim to recover.



## TV-Flow via PDEs on Orientation Scores

TV-flow enhancement is a technique for denoising surfaces but simultaneously preserving sharp edges. TV-flow  $U = W_\phi f \mapsto W_0(\cdot, t) = \lim_{\varepsilon \rightarrow 0} W_\varepsilon(\cdot, t)$  is given by

$$\begin{cases} \frac{\partial W_\varepsilon}{\partial t}(\mathbf{p}, t) = \operatorname{div}\left(\frac{\nabla_G W_\varepsilon(\cdot, t)}{\sqrt{\varepsilon^2 + |\nabla_G W(\cdot, t)|^2}}\right)(\mathbf{p}), & \mathbf{p} = (\mathbf{x}, \mathbf{n}) \in SE(2), t \geq 0, \\ W_\varepsilon(\mathbf{p}, 0) = U(\mathbf{p}), \end{cases}$$

with metric  $\mathcal{G}_{\mathbf{p}}(\dot{\mathbf{p}}, \dot{\mathbf{p}}) = C(\mathbf{p})^2 \left( \xi^2 \underbrace{|\dot{\mathbf{x}} \cdot \mathbf{n}|^2}_{\text{forward}} + \frac{\xi^2}{\zeta^2} \underbrace{\|\dot{\mathbf{x}} \wedge \mathbf{n}\|^2}_{\text{sideways}} + \underbrace{\|\dot{\mathbf{n}}\|^2}_{\text{angular}} \right)$ ,

where  $U(\mathbf{p})$  and  $W_0(\mathbf{p}, t)$  denote the (enhanced) orientation score.

## Metric Tensor Field

The metric tensor field  $\mathcal{F}^M$  mixes the asymmetric Reeds-Shepp car model  $\mathcal{F}$  (near crossings) with extra constraint  $\dot{\mathbf{x}} \cdot \mathbf{n} \geq 0$ , and a data-driven variant  $\mathcal{F}^U$ .

## Theorem for Asymmetric Geodesic Tracking

The shortest curve  $\gamma : [0, 1] \rightarrow SE(2)$  with  $\gamma(0) = \mathbf{p}$  and  $\gamma(1) = \mathbf{p}_0$  can be computed by steepest descent tracking on distance map  $W(\mathbf{p}) = d_{\mathcal{F}^M}(\mathbf{p}, \mathbf{p}_0)$

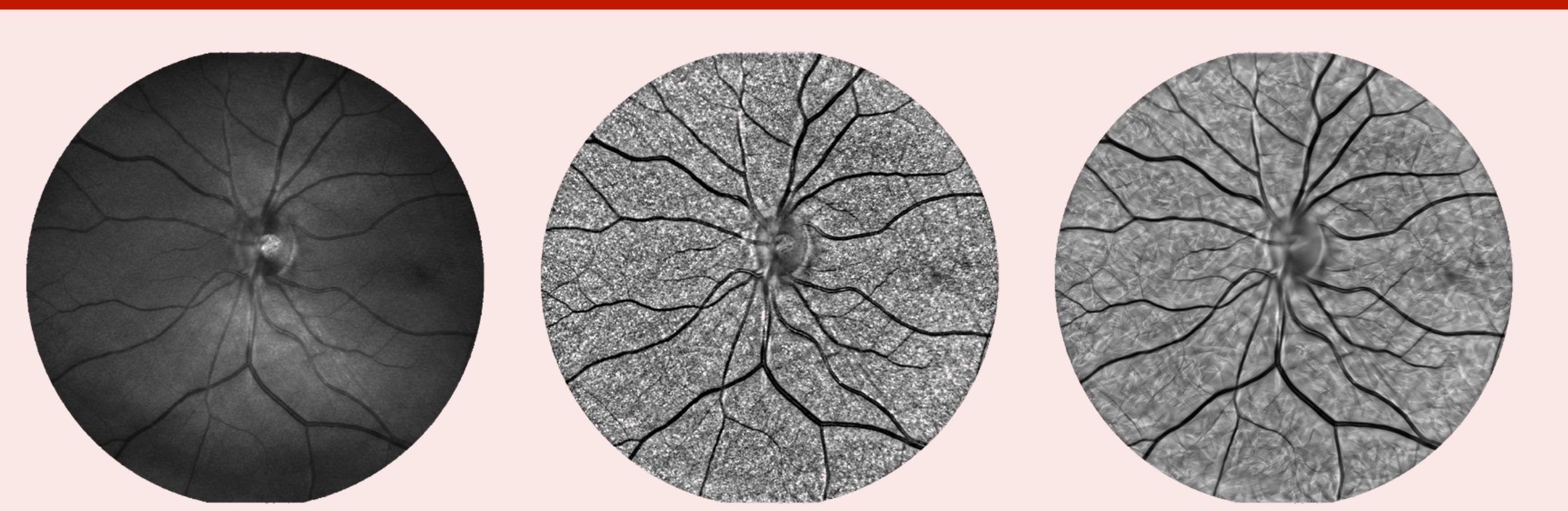
$$\gamma(t) := \gamma_{\mathbf{p}, \mathbf{p}_0}^U(t) = \operatorname{Exp}_{\mathbf{p}}(t v(W)), \quad t \in [0, 1], \quad (1)$$

where  $\operatorname{Exp}$  integrates the following vector field on  $SE(2)$ :  $v(W) := -W(\mathbf{p}) \nabla_{\mathcal{F}^M} W$  and where  $W$  is the viscosity solution of the eikonal PDE system

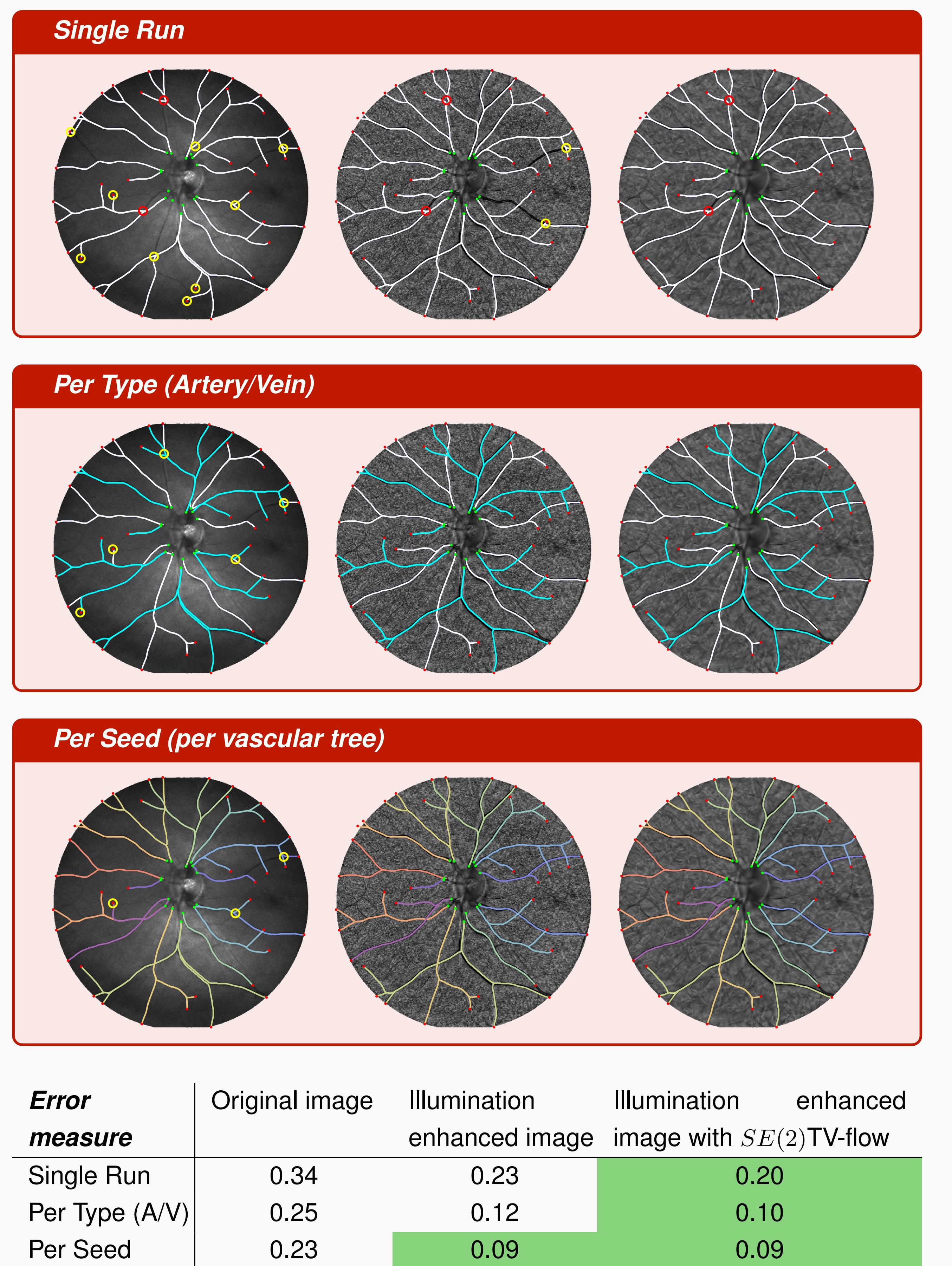
$$\begin{cases} \mathcal{F}_M^*(\mathbf{p}, dW(\mathbf{p})) = 1 & \mathbf{p} \in SE(2), \\ W(\mathbf{p}_0) = 0, \end{cases} \quad (2)$$

with dual Finsler function  $\mathcal{F}_M^*(\mathbf{p}, \dot{\mathbf{p}}) := \max\{\langle \dot{\mathbf{p}}, \dot{\mathbf{p}} \rangle \mid \dot{\mathbf{p}} \in T_{\mathbf{p}}(SE(2)) \text{ with } \mathcal{F}^M(\mathbf{p}, \dot{\mathbf{p}}) \leq 1\}$ .

Retinal image, illumination enhanced image excl. and incl. TV-flow on  $SE(2)$



## Experimental Results



## Conclusion

We developed a new asymmetric, data-driven left-invariant Finsler geometric model that includes contextual contrast enhancement via TV-flows on  $SE(2)$ . The new model reduces many errors and performs very well on both realistic and challenging low-quality retinal images where entire vascular trees are computed from a single asymmetric Finslerian distance map.

