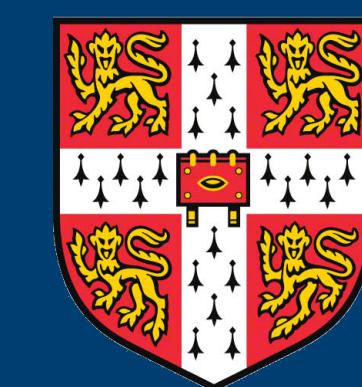


Multi-Contrast MRI Reconstruction with Structure-Guided Total Variation



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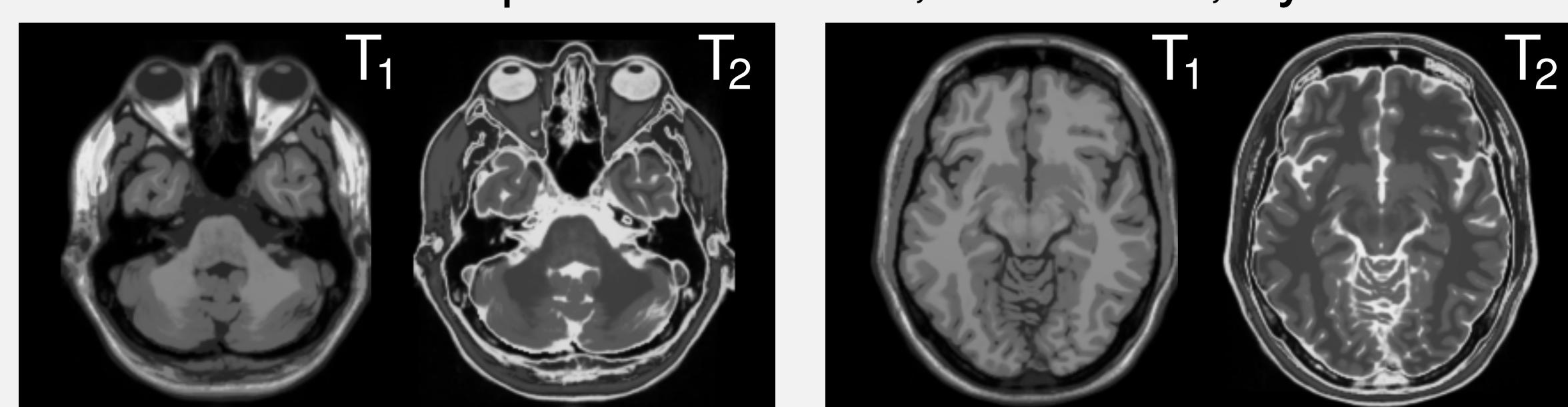
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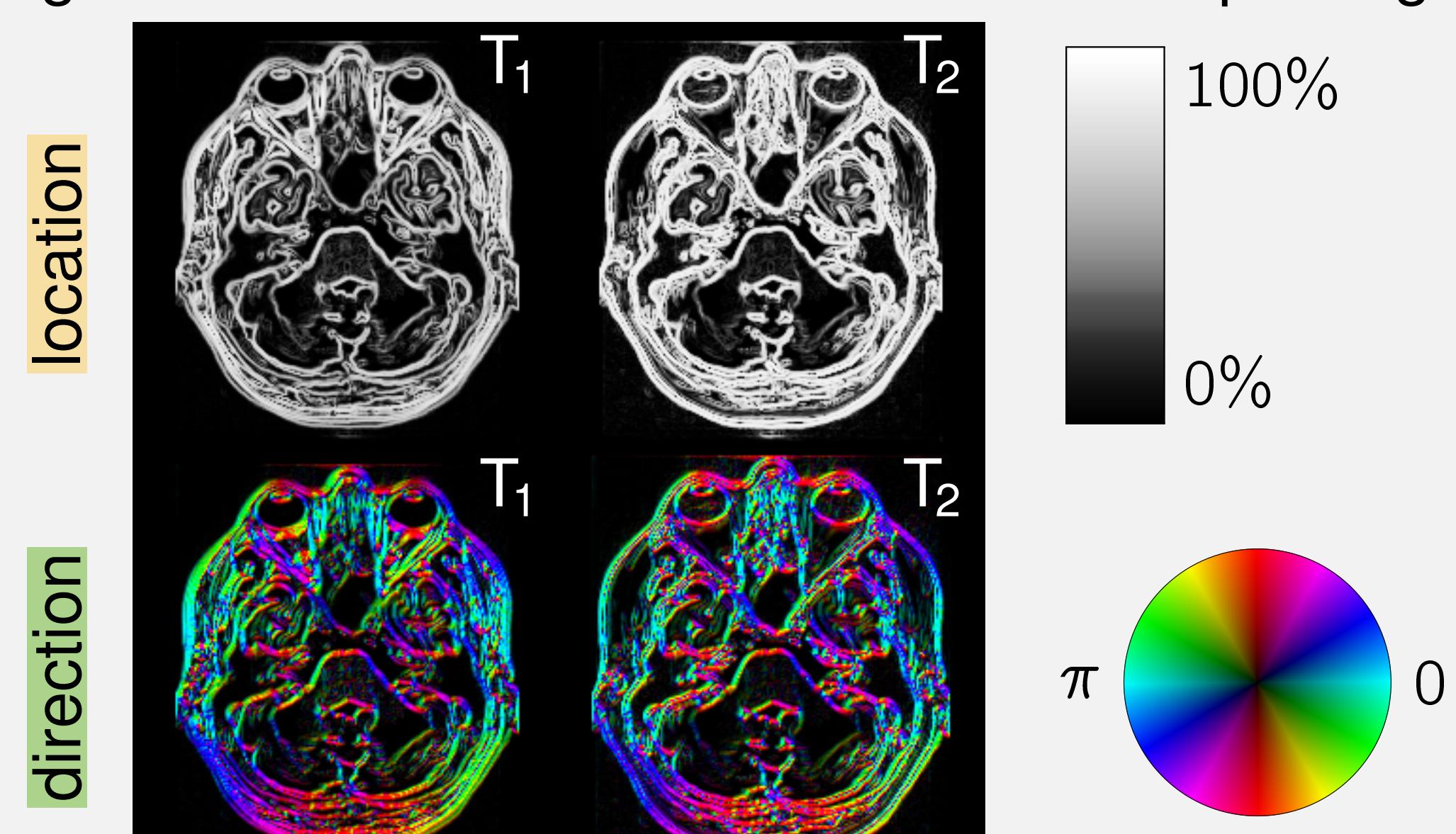
Motivation and Purpose

- magnetic resonance imaging (MRI) is a versatile technology with many **different contrasts** (e.g. see figure below for T_1 and T_2)
- MRI contrasts show **similar structures** due to same anatomy [1]
- we **exploit redundancy**, transfer structure from one contrast to another and reconstruct with less data
- **shorter scan times**: patient comfort, save time, dynamic imaging



What is Structure?

- Difficult to compare images of different contrasts
- Base image structure on **location** or **direction** of spatial gradients



Structure-Guided Total Variation

- Embed side information v in prior (regularization functional) with spatially varying matrix-field $\mathcal{D}: \Omega \rightarrow \mathbb{R}^{N \times N}$

$$\operatorname{argmin}_u \left\{ \frac{1}{2} |Au - b|^2 + \alpha \int_{\Omega} |\mathcal{D}(x) \nabla u(x)| dx \right\}$$

- Reduces to normal total variation (TV) for $\mathcal{D} = \mathcal{I}$

- Isotropic structure (**location**) [2–4]:

$$\mathcal{D}(x) = \eta / |\nabla v(x)|_{\eta} \quad (\text{wTV})$$

with $|\nabla v(x)|_{\eta} = \sqrt{|\nabla v(x)|^2 + \eta^2}$, $\eta > 0$

- Anisotropic structure (**direction**) [4–6]:

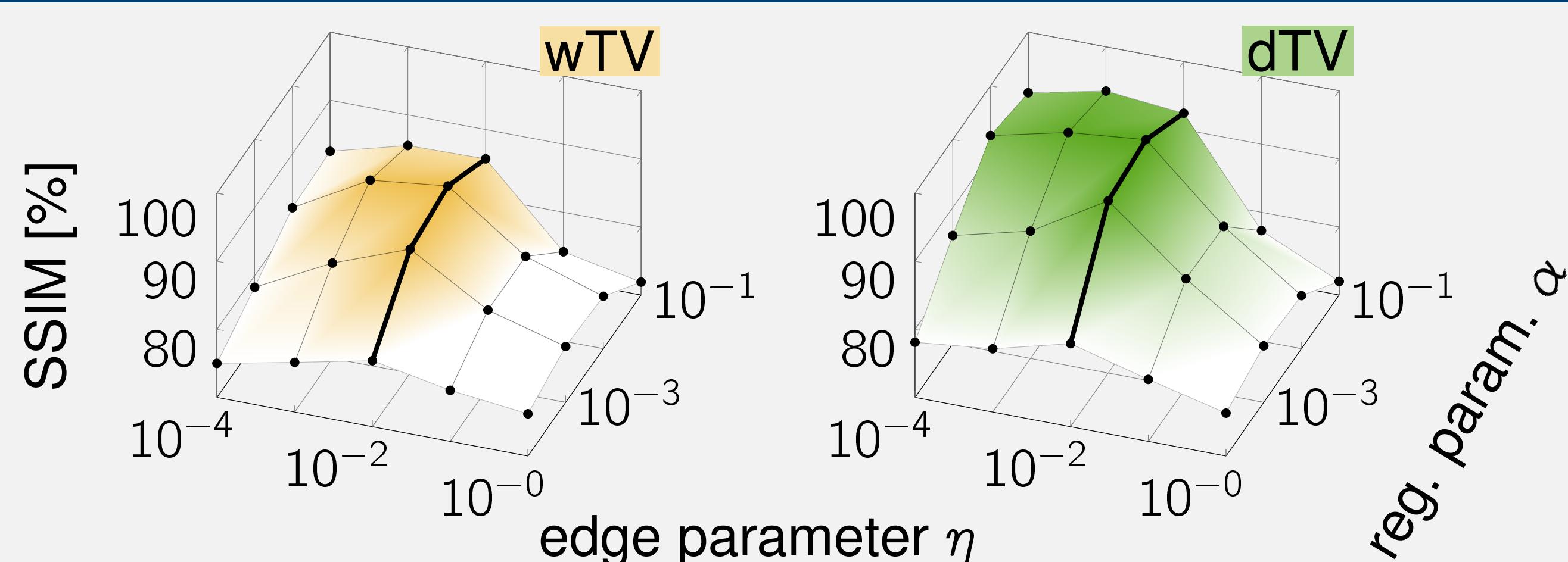
$$\mathcal{D}(x) = \mathcal{I} - \xi(x) \xi^T(x) \quad (\text{dT}V)$$

with $\xi(x) = \nabla v(x) / |\nabla v(x)|_{\eta}$

- Dual formulation allows efficient algorithms [4]

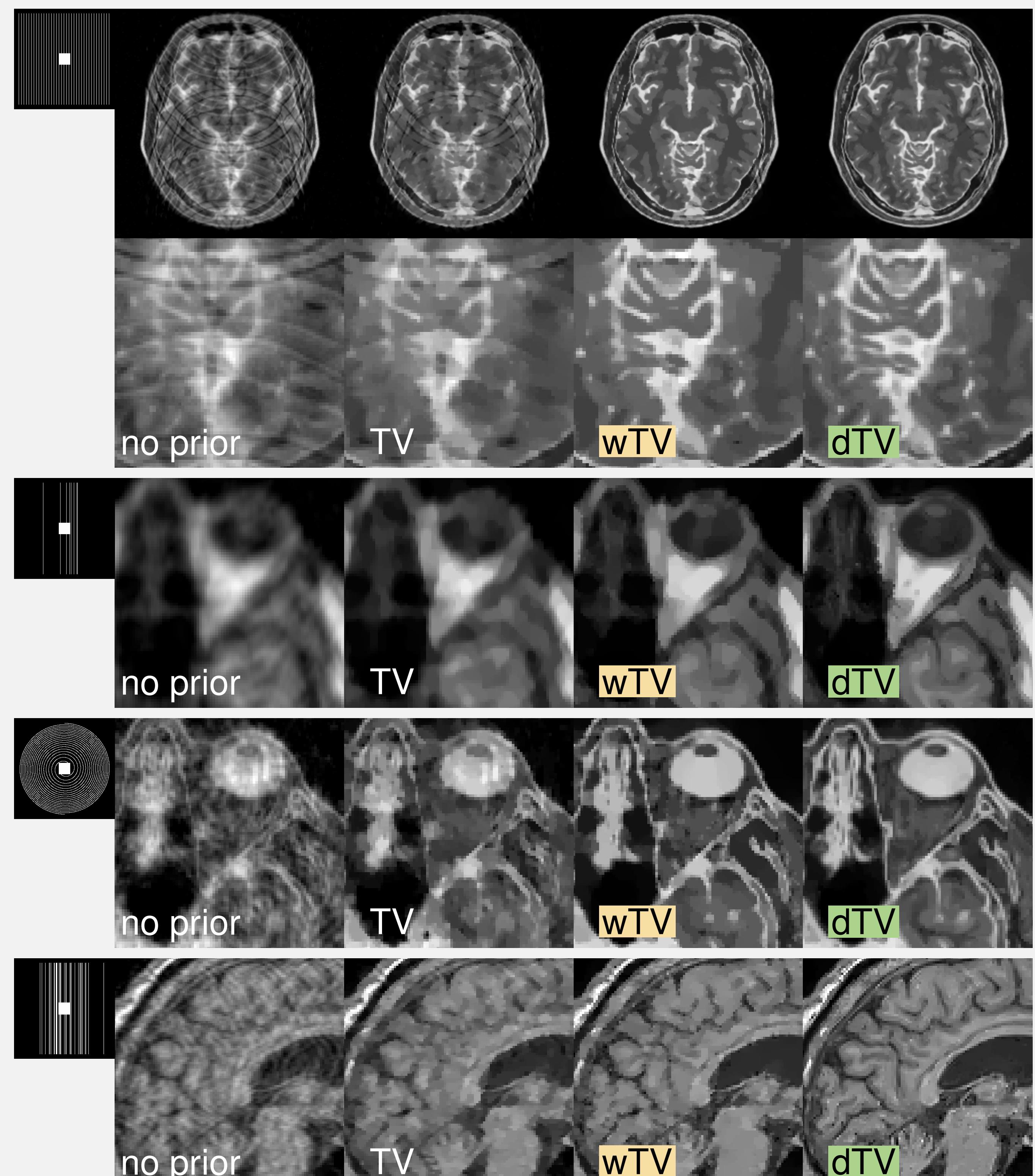
$$\int_{\Omega} |\mathcal{D}(x) \nabla u(x)| dx = \sup_{|\varphi(x)| \leq 1} \left\{ \int_{\Omega} u(x) \operatorname{div} [\mathcal{D}^T(x) \varphi(x)] dx \right\}$$

Edge Parameter η



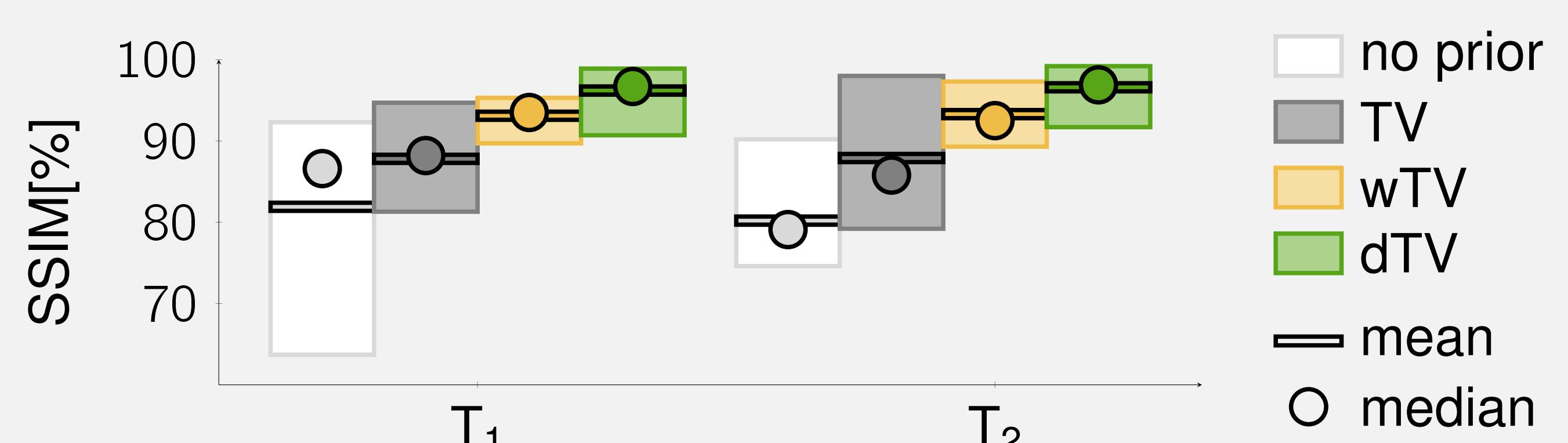
- Best results for $\eta = 10^{-2}$. Trade-off: regularization v structure
- For large η , both methods perform the same (as TV; not shown)

Visual Results



- From left to right: priors enhance visual quality

Quantitative Results



- Range (min to max), mean and median over 12 data sets

Conclusions

- Exploiting redundancy (utilizing either **location** or **direction**) allows **reconstruction from less data**
- The **anisotropic** prior consistently outperforms the **isotropic** one, leading to **less artefacts** and a **higher level of detail**

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