



BACKGROUND & SETTING

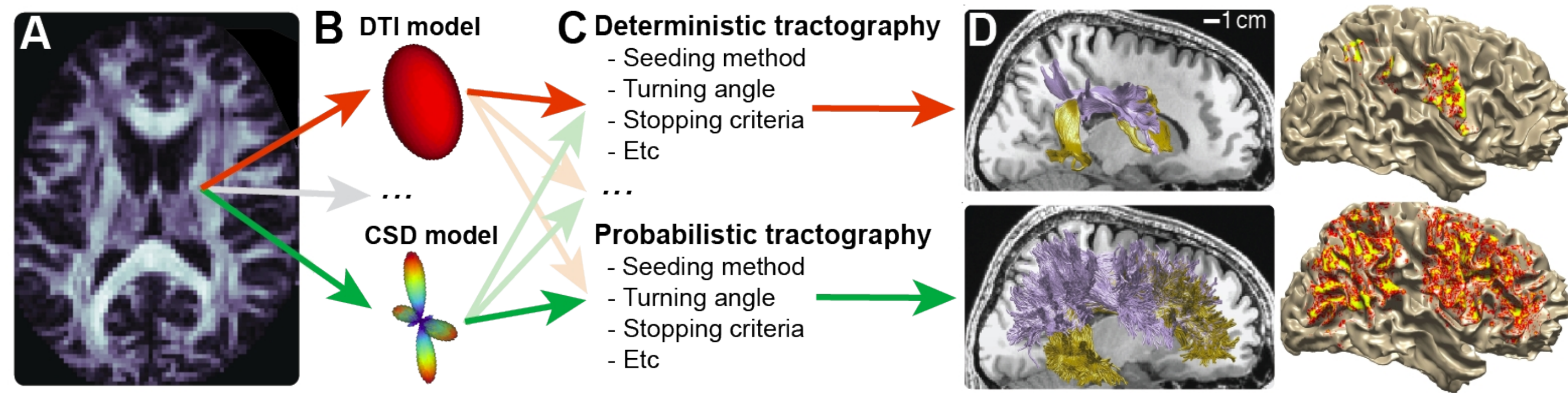
1. Motivations

Goal:

- Given:
 - Diffusion magnetic resonance imaging (dMRI) data
- We want to:
 - Map structural brain connectome

Applications:

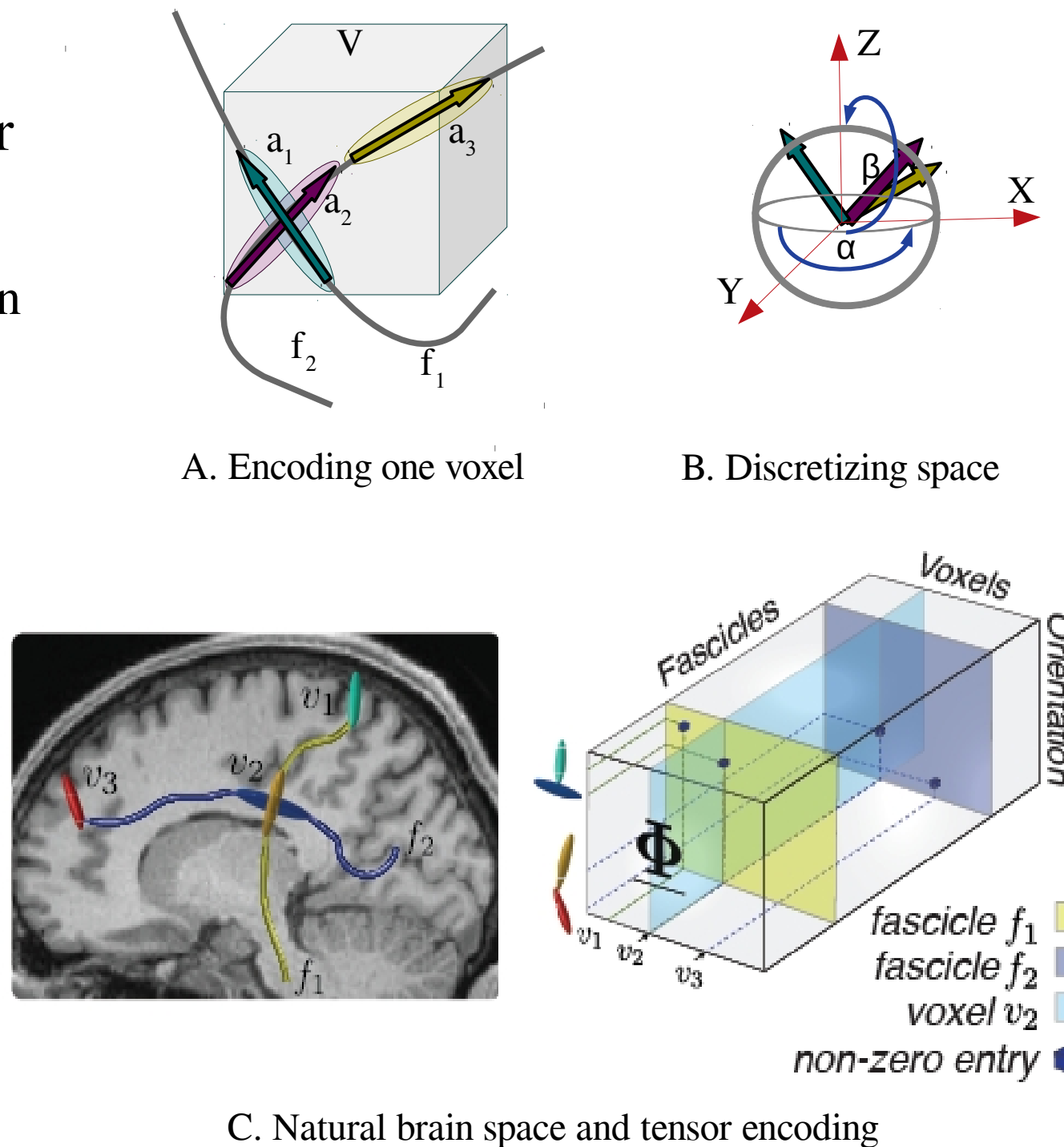
- Investigating white matter health and disease



2. Encoding Brain Connectomes as Tensors

ENCODE:

- Represents brain structure by a 3D sparse tensor
 - $\Phi \in R^{Na \times N_v \times N_f}$
 - N_a : #orientations, fascicles orientation at each position
 - N_v : #voxels, fascicles spatial position
 - N_f : #fascicles, indices of each fascicle
- Unifies dMRI signal with connectome structure
 - Matrix of dMRI signal $Y \in R^{N_0 \times N_v}$,
 - θ is gradient direction
 - Factorizing Y into Φ and dictionary D
 - $D \in R^{N_0 \times N_a}$
 - $Y \approx \Phi \times_l D \times_3 W$, where $W \in R^{N_f}$



THEORY & ALGORITHMS

3. A Tractography Objective for Extracting Brain Connectomes

- Unconstrained objective to extract Φ

$$\Phi = \operatorname{argmin}_{\Phi} \|Y - \Phi \times_l D \times_3 I\|^2, \text{ where } I \in R^{N_f}$$

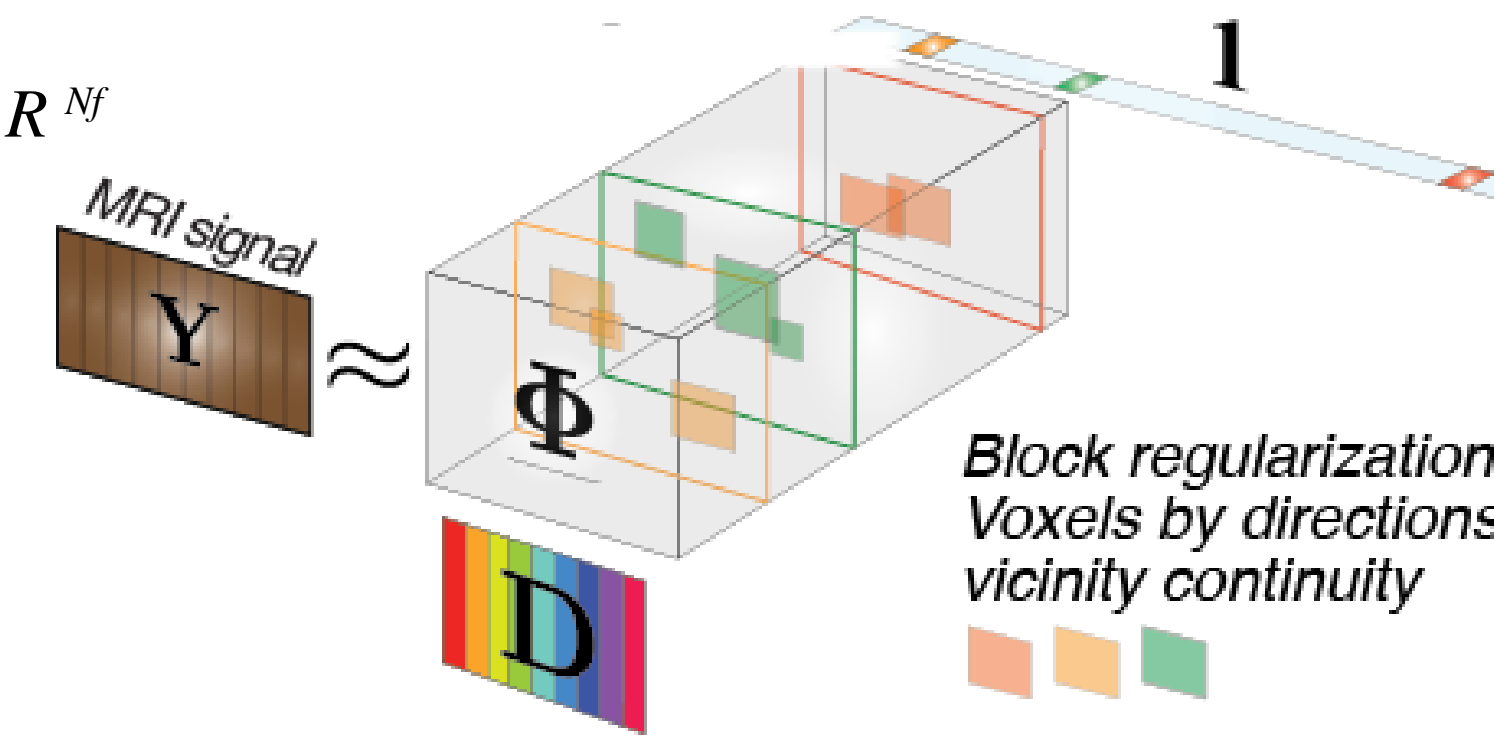
- Regularizer – smooth, continuous, sparse

- Constrained objective to learn Φ

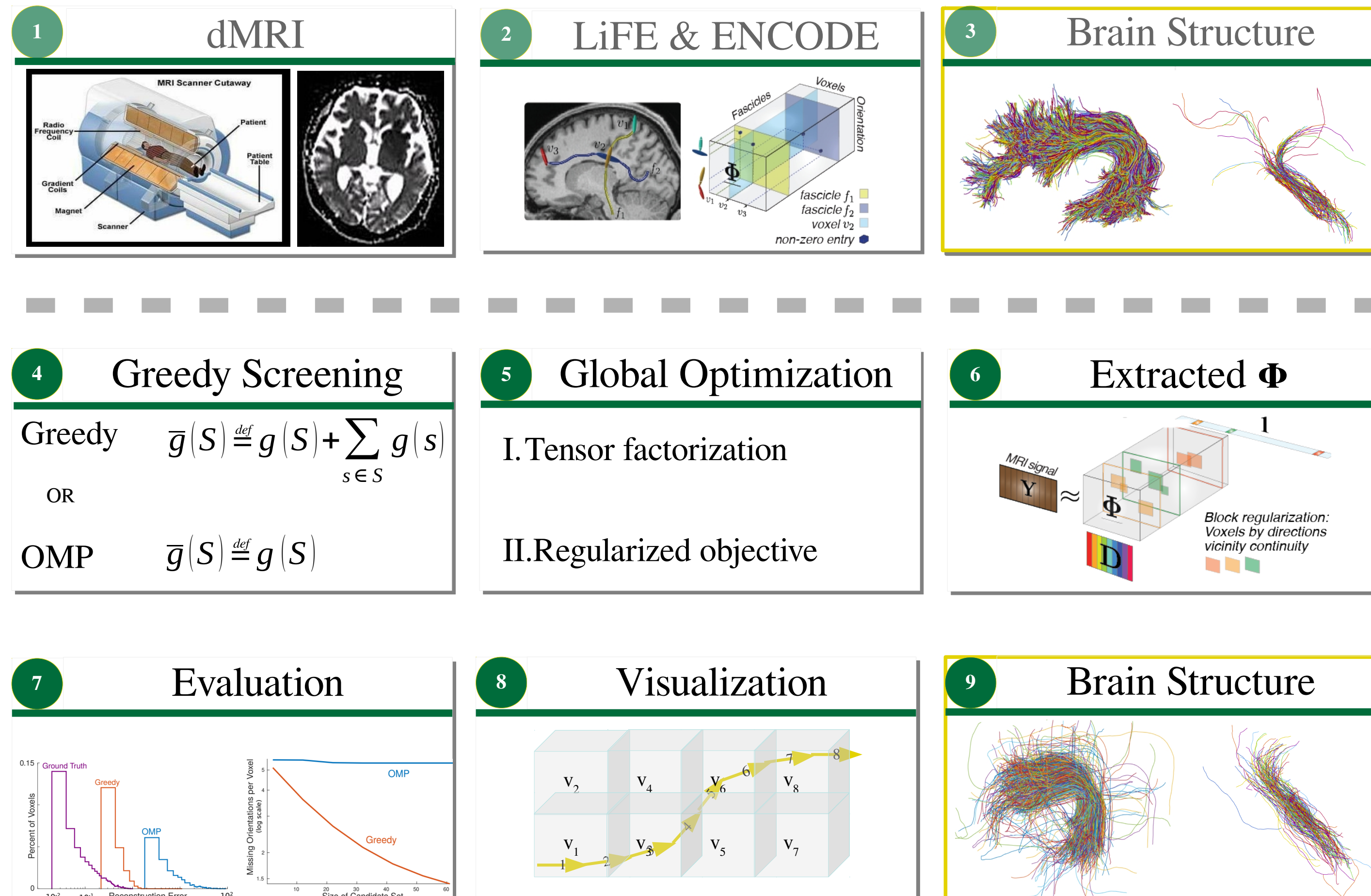
$$\min_{\Phi} \|Y - \Phi \times_l D \times_3 I\|^2 + \lambda R(\Phi),$$

$$R(\Phi) = \sum_{f \in F} \sum_{G_v \subseteq V} \sum_{G_A \subseteq A} \sqrt{\sum_{v \in G_v} \left(\sum_{a \in G_A} |\Phi_{a,v,f}| \right)^2}$$

Model formulation and block regularizer



4. A Pipeline for Extracting Brain Connectomes



EMPIRICAL RESULTS

5. Reconstructing the Anatomical Structure of Tracts

