

# Registration of Brain Resection MRI with Intensity and Location Priors



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# Image Registration Goal

- Align post-resection and preoperative brain MRI of epilepsy patients



- Challenge: **Missing Correspondences**
  - Cause misalignment of other actual corresponding features

# Approaches to Handle Missing Correspondences

## Previous Methods

- Biomechanical models + deformable registration for tumor/normal brain alignment<sup>1</sup>
- Point-based algorithm for pre-/intra-operative resection images<sup>3</sup>
- Estimate registration and missing data based on image similarity<sup>4</sup>
- Our recent simultaneous registration/correspondence estimation with intensity prior<sup>5</sup>

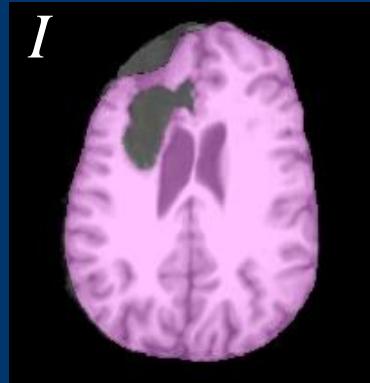
## Our Approach

- Jointly register and classify correspondences in statistical estimation framework<sup>2</sup>
- Voxel-based, no feature pre-processing needed for images to be registered
- Consider other terms beside image similarity to determine missing correspondences
- Include indicator map prior for location and smoothing

<sup>1</sup>Mohamed et al., MedIA 2006; <sup>2</sup>Pohl et al., Neuroimage 2006; <sup>3</sup>Liu et al., ISBI 2010;  
<sup>4</sup>Periaswamy and Farid, MedIA 2006; <sup>5</sup>Chitphakdithai and Duncan, MICCAI 2010

# Registration and Indicator Map Estimation (RIME): Overview

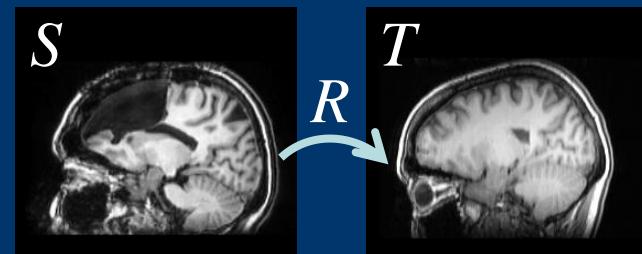
- Introduce “hidden” indicator map to segment valid tissue, resection, and background



- Given indicator map  
→ easier to register
- Given correct alignment  
→ easier to classify regions

- Maximum a posteriori framework:

$$\hat{R} = \arg \max_R \log \sum_I p(R, I | S, T)$$



# Registration and Indicator Map Estimation (RIME): EM Algorithm

- E-Step: Indicator Map Weights  $p(I(\mathbf{x})=l | S, T, R^k) =$

$$\frac{p(T(R^k(\mathbf{x})) | S, I(\mathbf{x})=l, R^k) p(S(\mathbf{x}) | I(\mathbf{x})=l) p(I(\mathbf{x})=l)}{\sum_{l'} p(T(R^k(\mathbf{x})) | S, I(\mathbf{x})=l', R^k) p(S(\mathbf{x}) | I(\mathbf{x})=l') p(I(\mathbf{x})=l')}$$

*Similarity Term*      *Intensity Prior*      *Indicator Map Prior*

- M-Step: Update registration  $R^{k+1} =$

$$\arg \max_R \left[ \sum_{\mathbf{x} \in S} \sum_{l \in L} p(I(\mathbf{x})=l | S, T, R^k) \log p(T(R(\mathbf{x})) | S, I(\mathbf{x})=l, R) + \log p(R) \right]$$

*E-Step Weights*      *Similarity Term*  
*Transformation Prior*

# Similarity Metric, Intensity Prior, and Transformation Prior

- Similarity Metric

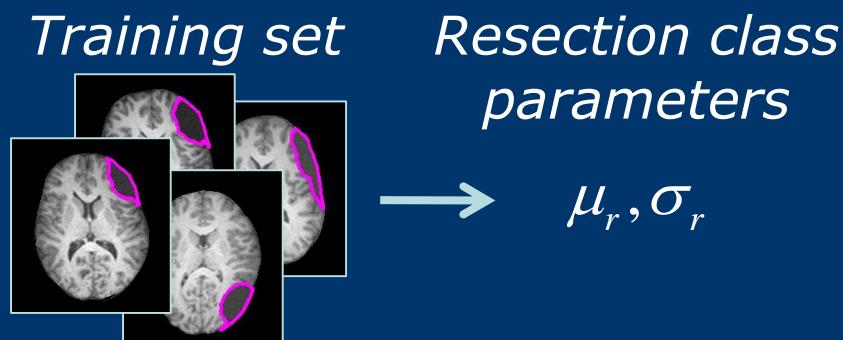
$$P(T(R(\mathbf{x}))) | S, R, I(\mathbf{x}) = l \sim \begin{cases} Unif\left(\frac{1}{c}\right) & , l = \text{resection} \\ N(S(\mathbf{x}), \sigma_1) & , l = \text{valid tissue} \\ N(S(\mathbf{x}), \sigma_2) & , l = \text{background} \end{cases}$$

- Intensity Prior

- Different distribution for intensities given indicator label

- Transformation Prior

- Prior constrains B-spline control points for uniform cubic B-spline FFD model



# Indicator Map Prior: MRF and Mean Field Approximation

- Impose Markov random field onto indicator map

$$p(I) = \frac{1}{Z} \exp\left[-\beta \sum_c V_c(I)\right] \quad \begin{matrix} \text{Problem:} \\ \text{Intractable} \\ \text{calculation} \end{matrix}$$

- Approximate using mean field theory

$$p(I) \approx \prod_{\mathbf{x}} \frac{1}{Z'_{\mathbf{x}}(\beta)} \exp\left[-\beta \sum_c V'_c(I(\mathbf{x}), \bar{I})\right]$$

where  $\bar{I} \approx I^k = \arg \max_l p(I(\mathbf{x})=l | S, T, R^k)$

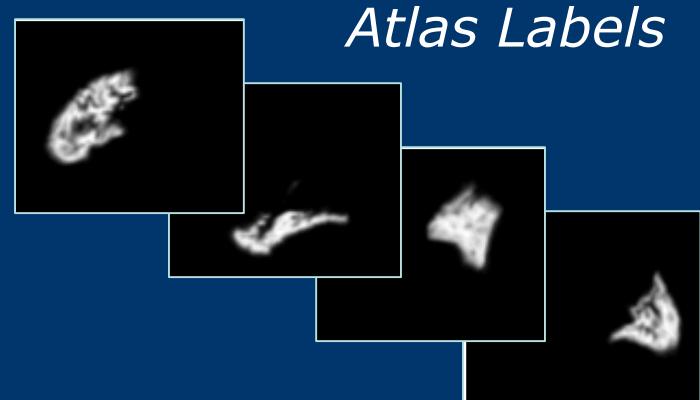
- Prior probability at an indicator map voxel:

$$p(I(\mathbf{x})=l | \beta_L, \beta_S) \approx \frac{1}{Z'_{\mathbf{x}}(\beta_L, \beta_S)} \exp\left[\beta_L V'_L(l, I^A(\mathbf{x})) + \beta_S \sum_{\mathbf{n} \in N(\mathbf{x})} V'_S(l, I^k(\mathbf{n}))\right]$$

# Indicator Map Prior: Location Prior Term

$$p(I(\mathbf{x})=l | \beta_L, \beta_S) \approx \frac{1}{Z'_x(\beta_L, \beta_S)} \exp \left[ \beta_L V'_L(l, I^A(\mathbf{x})) + \beta_S \sum_{\mathbf{n} \in N(\mathbf{x})} V'_S(l, I^k(\mathbf{n})) \right]$$

- $V'_L(l, I^A(\mathbf{x}))$  = probability voxel  $\mathbf{x}$  takes on label  $l$  as given by atlas  $I^A$
- User chooses atlas region (e.g. left temporal lobe) which best corresponds to resection location  
→ Helpful for CSF voxels which have similar intensity to resection voxels

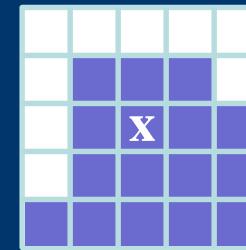


# Indicator Map Prior: Smoothing Prior Term

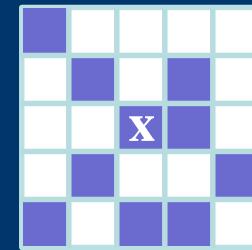
$$p(I(\mathbf{x})=l | \beta_L, \beta_S) \approx \frac{1}{Z'_{\mathbf{x}}(\beta_L, \beta_S)} \exp \left[ \beta_L V'_L(l, I^A(\mathbf{x})) + \beta_S \sum_{\mathbf{n} \in N(\mathbf{x})} V'_S(l, I^k(\mathbf{n})) \right]$$

- Depends on values of neighbors  $N(\mathbf{x})$  of  $\mathbf{x}$  in current indicator map estimate  $I^k$
- Use Potts model:

$$V'_S(l, I^k(\mathbf{n})) = \delta(l, I^k(\mathbf{n}))$$



*Higher*  $\sum_{N(\mathbf{x})} V'_S$



*Lower*  $\sum_{N(\mathbf{x})} V'_S$

→ Helpful for eliminating small mislabeled regions

# Indicator Map Prior: Weighting Parameters

$$p(I(\mathbf{x})=l | \beta_L, \beta_S) \approx \frac{1}{Z'_x(\beta_L, \beta_S)} \exp \left[ \beta_L V'_L(l, I^A(\mathbf{x})) + \beta_S \sum_{\mathbf{n} \in N(\mathbf{x})} V'_S(l, I^k(\mathbf{n})) \right]$$

$\beta_L$  *Location weight*  
  $\beta_S$  *Smoothing weight*

- In M-step, estimate  $\beta_L^{k+1}, \beta_S^{k+1} =$

$$\arg \max_{\beta_L, \beta_S} \left[ \sum_{\mathbf{x}} \sum_l p(I(\mathbf{x})=l | S, T, R^k) \log p(I(\mathbf{x})=l | \beta_L, \beta_S) + \log p(\beta_L, \beta_S) \right]$$

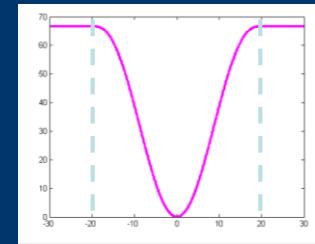
- Assign normal dist with small mean on  $\beta_L, \beta_S$   
 → Constraint so indicator map estimation not overly dependent on prior

# Experiments: Methods Compared

- “Standard” non-rigid registration<sup>1</sup> (*SNRR*)
- Robust similarity<sup>2</sup> (*RSR*)
  - Reduce influence of outliers based on data term
- Registration and indicator map estimation with only intensity prior<sup>2</sup> (*RIME-int*)
- Proposed approach (*RIME-imp*)

\*All methods use:

- Voxel intensity-based similarity
- FFD transformation model

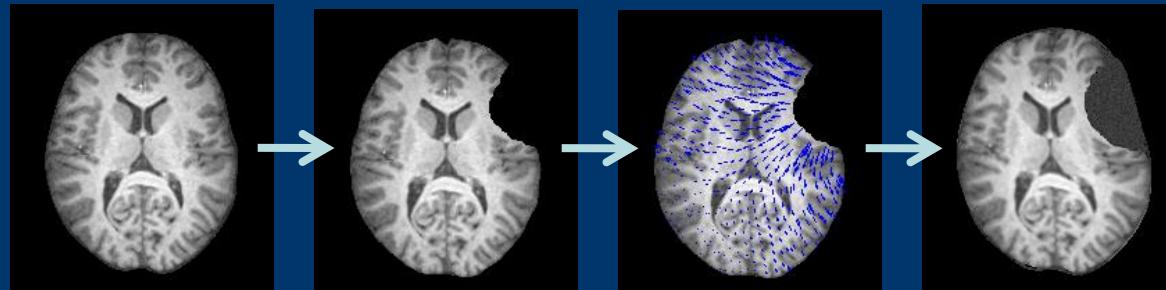


*Tukey Function*

<sup>1</sup>Rueckert et al., TMI 1999; <sup>2</sup>Chitphakdithai and Duncan, MICCAI 2010

# Synthetic Data: Setup

- Synthetic post-resection image creation



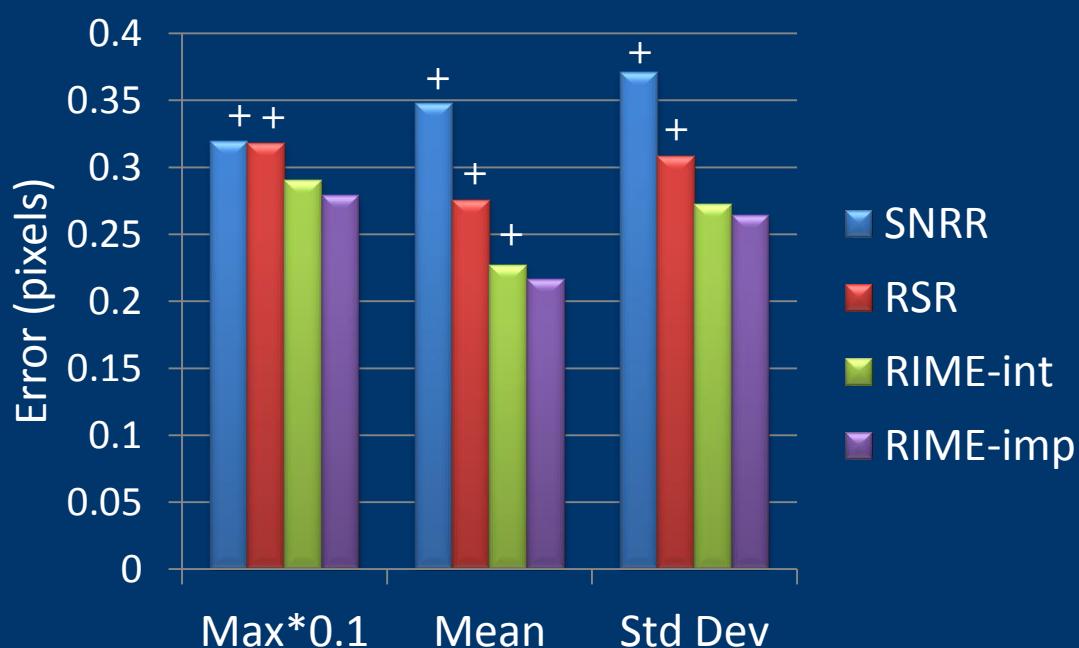
- Probabilistic atlas labels



- For intensity prior: leave-one-out method,  
trained on 10 images

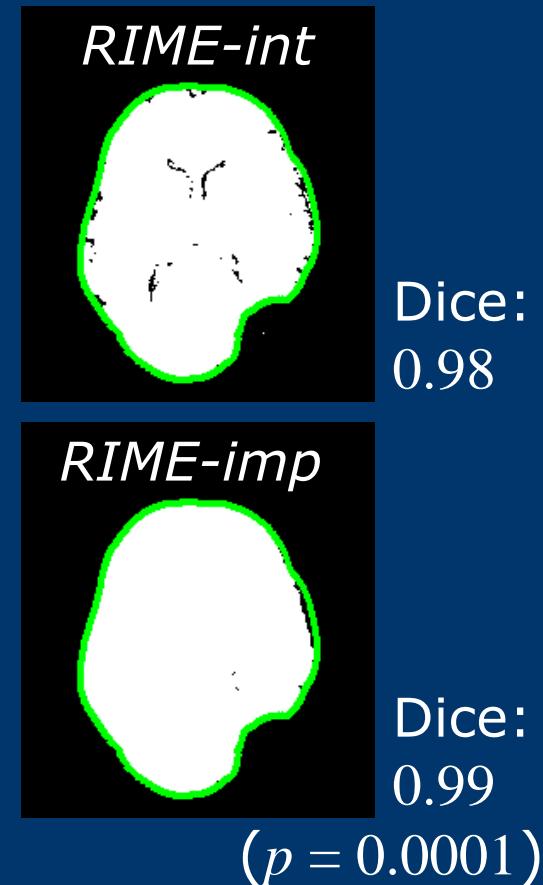
# Synthetic Data: Results

Average Displacement Field Errors



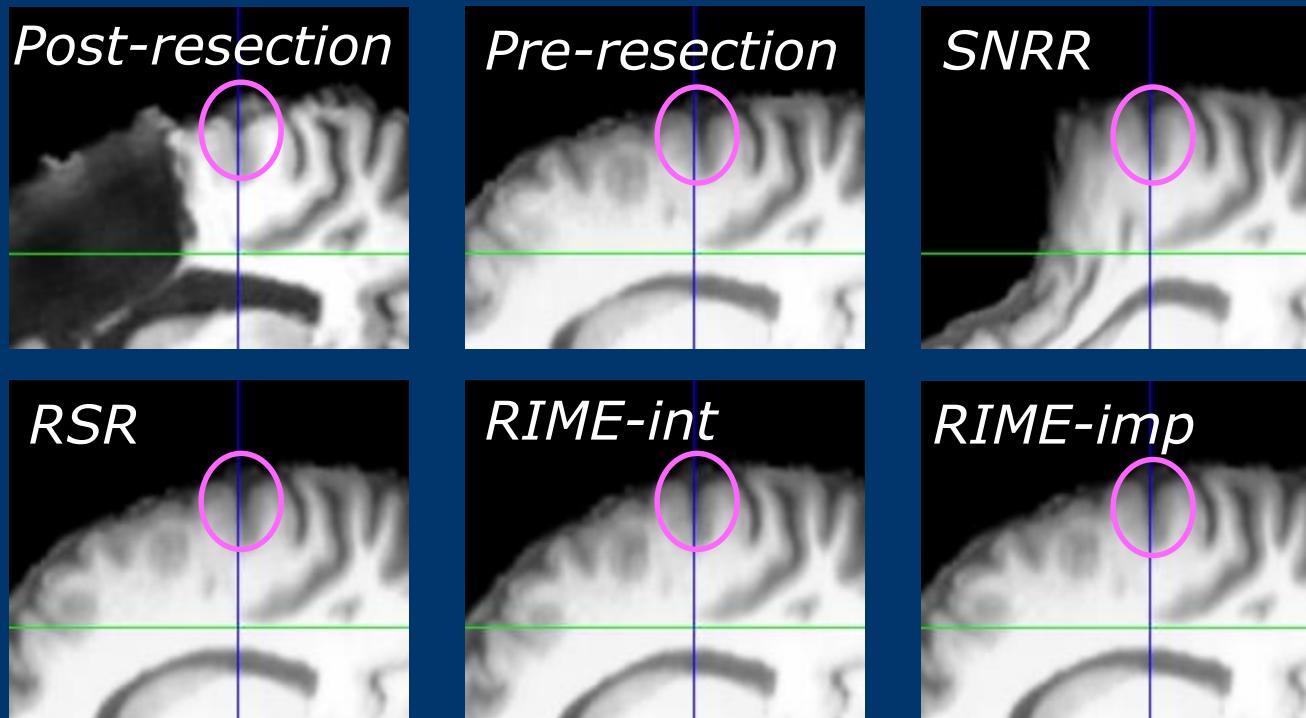
+ Significantly higher compared  
to *RIME-imp* ( $p < 0.05$ )

Valid Tissue Map



# Patient Data: Registration Results

- 6 pairs, affine-aligned to standard MNI brain
- Custom atlas based on MNI atlas: 19 labels

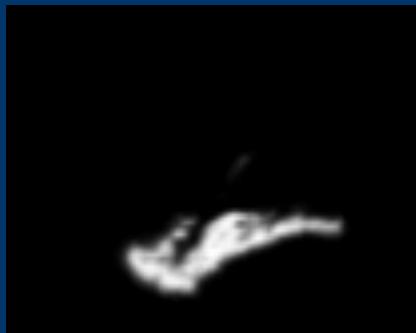


*RIME-imp* produced lowest RMSE ( $p < 0.05$ )

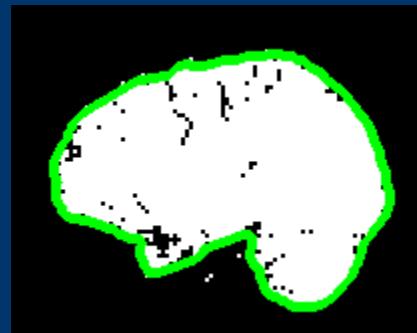
# Real Data: Indicator Map Estimation



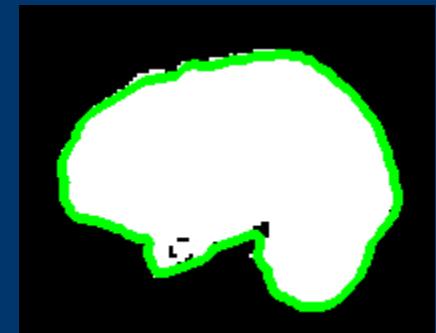
Left temporal  
lobe  
resection



Atlas for  
missing  
region  
location prior



Valid  
correspondence  
region using  
*RIME-int*



Valid  
correspondence  
region using  
*RIME-imp*

Average dice overlap between  
true and valid tissue regions:

0.92                    0.96

\* *RIME-imp* significantly increased overlap ( $p < 0.05$ )

# Conclusions and Future Work

- Contributions
  - Registration handles missing correspondences by simultaneously estimating indicator map
  - Included MRF-based indicator map prior to incorporate location information and map smoothness
    - Improved label segmentation and registration accuracy
- Future Work
  - Increase number of labels, e.g. for CSF
  - Adapt general framework to other registration problems with missing correspondences

# Thank You!

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