

# False Sense of EPI-to-Structural Alignment with Common Cross-Modality Registration Methods



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## The Problem

- Aligning EPI volumes to T<sub>1</sub>-weighted volumes using Mutual Information (MI) or Correlation Ratio (CR) as the cost functional can produce registrations that **look good but are actually bad**
- Brain outlines from the two volumes might match well, but this **can be very misleading**:

- Interior structures (ventricles, fissures, sulci) that are visible in both types of images often are displaced 5 mm — or more
- This is **not a software issue**: AFNI (3dAllineate), SPM (COREG), and FSL (FLIRT) all often fail to give good anatomical matchings, upon close visual inspection

### Sample Images:

- T<sub>1</sub>-weighted volumes as the grayscale background, each one registered to the EPI volume with a distinct method
- EPI volume is edge-detected and only its edges are shown in the color overlay
- Two implementations of MI (AFNI/3dAllineate & SPM/COREG)
- Two implementations of CR (AFNI/3dAllineate & FSL/FLIRT)
- Our new LPC cost functional (AFNI/align\_epi\_anat.py)
- EPI interior edges track structural edges only with LPC**

## Local Pearson Correlation Cost Functional (LPC)

- Weighted correlation  $r(x)$  calculated over neighborhood  $N(x)$  of any point  $x$ ; then  $r(x)$ 's are nonlinearly combined to give final cost:

$$W(x) = \sum_{y \in N(x)} w(y) \quad \text{[local sum of weights]}$$

$$M(x; F) = \frac{1}{W(x)} \sum_{y \in N(x)} w(y) \cdot F(y) \quad \text{[local weighted mean of volume F]}$$

$$Q(x; F, G) = \sum_{y \in N(x)} w(y) \cdot [F(y) - M(x; F)] \cdot [G(y) - M(x; G)] \quad \text{[local scalar product of F & G]}$$

$$r(x) = \frac{Q(x; E, S)}{[Q(x; E, E) \cdot Q(x; S, S)]^{1/2}} \quad \text{[local weighted correlation coefficient]}$$

$$C_{LPC}[E, S] = \sum_{x \in P} W(x) \cdot s(r(x)) \cdot |s(r(x))| / \sum_{x \in P} W(x) \quad \text{[combined correlation coefficients]}$$

- where:  $E(x)$ =EPI;  $S(x)$ =T<sub>1</sub>;  $s(r)$ = $\tanh^{-1}(r)$ ;  $N(x)$ =Kepler's rhombic dodecahedron centered at  $x$ ;  $P$ =FCC space-filling lattice of rhombic dodecahedra covering the brain volume; and  $W(x)$ =weight proportional to  $E(x)$  to accentuate matching of CSF (bright in EPI, dark in T<sub>1</sub>); the algorithm looks for the **most negative correlation** by minimizing  $C_{LPC}[E(x), S(T(x, \theta))]$  over affine transformations  $T(\bullet)$
- CSF (usually) tracks ventricles, fissures, sulci fairly well in EPI; LPC produces a robust match between those central *and* cortical anatomical structures visible in both EPI and structural volumes
- Computing correlations locally and then combining protects against shading artifacts and signal dropouts

## Conclusions

- Accurate and truly "robust" alignment of structural and EPI volumes requires a modality-specific cost functional
- And requires visual inspection of results, especially if you are relying on the function-to-structure correspondence:
- projection to cortical surface models; surgical planning

## Assessment Methodology

- Three raters (blinded to method and presentation order) each scored each of 27 {EPI, T<sub>1</sub>} volume pairs for alignment on a 4 point scale (from awful to excellent), for 8 different registration methods/tools:
  - 1=awful 2=errors > 5 mm 3=errors 2.5 mm 4=errors 0..2 mm
  - while viewing edge-enhanced images, in all three planes, overlaid in color and/or flickering between viewing  $E$  and  $S$
  - Sample datasets at 1.5 and 3.0 Tesla, from diverse sites
  - Raters agreed remarkably well (Spearman correlations=0.8)
  - Contingency table statistics confirms the obvious: **LPC wins**

