

The TL;DR

Background: Data have biases by collection site/scanner. If we combine data from multiple sites, we need to correct for site bias.

Idea: Compress our data so that it is uninformative of site.

Desiderata:

"Find a representation z of x such that $I(z : s) - \lambda I(x : z)$ is minimal for tradeoff λ between compression and relevance."

Surprisingly, we can solve this optimization problem, giving us:

- Site/scanner-invariant representations of images.
- Reconstructed images with **bounded** info. about their original site/scanner.

Takehome Intuition: Compression + Cond. Reconstruction \rightarrow Harmonization

Details

We want a function from scan x to new scan x' that removes the original site information s and replaces it with s' .

Data Proc. Ineq.: Let $q : x \mapsto z$ and $p : z, s' \mapsto x'$. If $z \perp s$, then $x' \perp s$. Moreover,

$$I(x', s) \leq I(z, s)$$

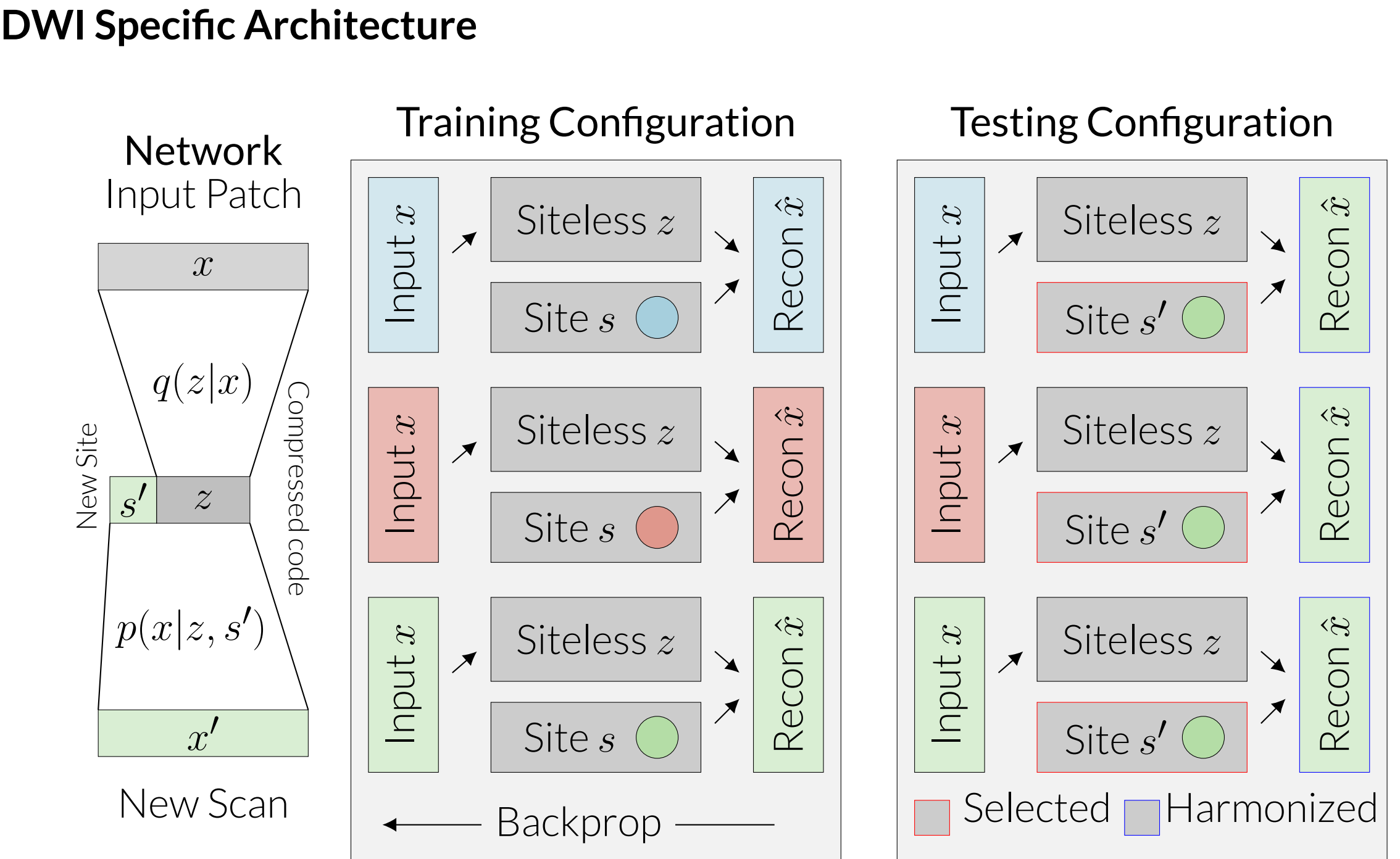
So by reducing $I(z, s)$, we can remove site information.

Further, if we let p depend on a new s' , we can attempt to map x to x' which has site information from s' .

Bound for $I(z, s)$

$$0 \leq I(z, s) \leq \underbrace{-\mathbb{E}_{x,s,z \sim q} [\log p(x|z, s)]}_{\text{Reconstruction}} + \underbrace{\mathbb{E}_x [KL[q(z|x) \parallel q(z)]]}_{\text{Compression}} - \underbrace{H(x|s)}_{\text{Const}}.$$

By minimizing this bound on $I(z, s)$ over q, p , we can learn a remapping function that tries to "forget" the original site variable, with a guarantee of the maximum amount of information retained.



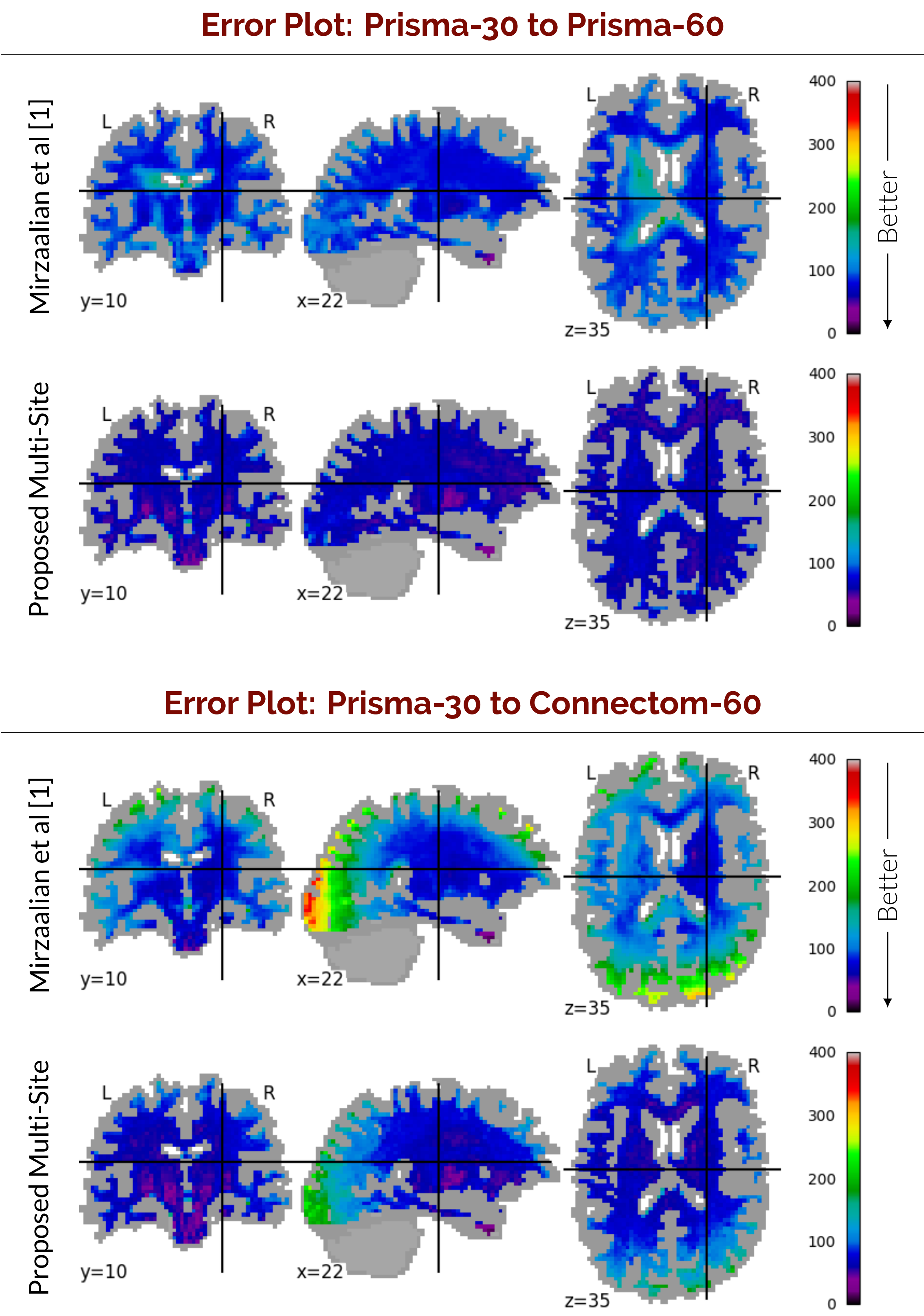
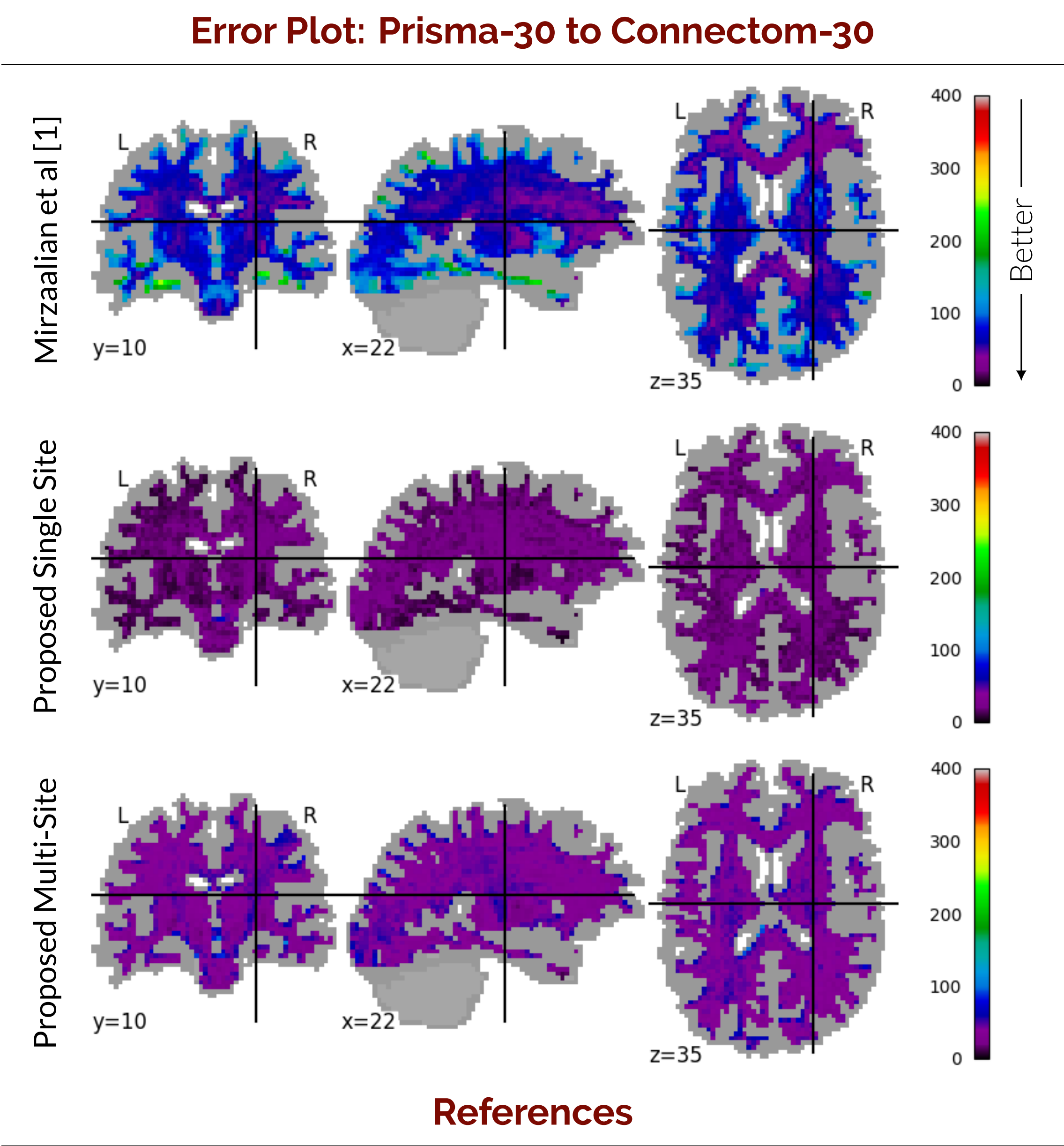
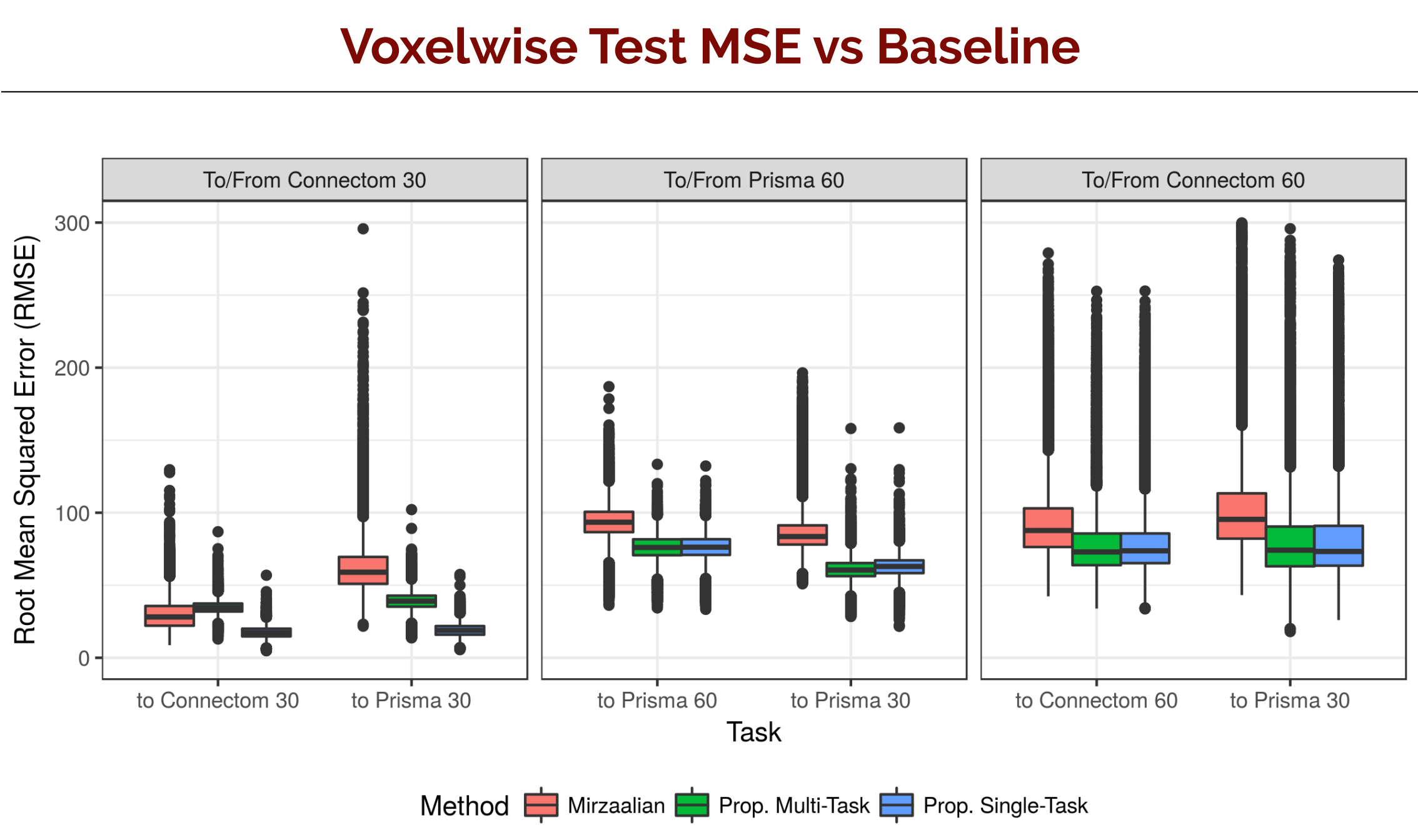
Data and Demonstration Experiements

We use the MICCAI CDMRI Diffusion Harmonization Challenge Dataset [2]:

- N=10 subjects scanned at all ``sites'', 8/2 Train-test split.
- All subjects Eddy and EPI corrected, then rigidly aligned. High spatial-resolution scans resampled to 2.4 mm isotropic.
- Projected to ℓ_2 -minimal 8th order Spherical Harmonics basis using SVD.

Site Name	Scanner-type	Voxel Size (iso)	TE/TR	# Grad.
Prisma-30	Prisma	2.4 mm	89/7200 ms	30
Prisma-60	Prisma	1.5 mm	80/4500 ms	30
Connectom-30	Connectom	2.4 mm	89/7200 ms	30
Connectom-60	Connectom	1.2 mm	68/5400 ms	60


All scans collected two diffusion shells, $1200 \frac{s}{mm^2}$ and $3000 \frac{s}{mm^2}$, and at least 4 inter-spersed b_0 volumes, one with a reversed phase encoding.



Please contact Daniel Moyer at moyerd@usc.edu for any questions/comments.

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The Paper



arxiv.org/abs/1904.05375

[1] Hengameh Mirzaalian et al.
Multi-site harmonization of diffusion MRI data in a registration framework.
Brain Imaging and Behavior, 12(1):284--295, 2018.

[2] Lipeng Ning et al.
Multi-shell diffusion MRI harmonisation and enhancement challenge (MUSHAC): progress and results.