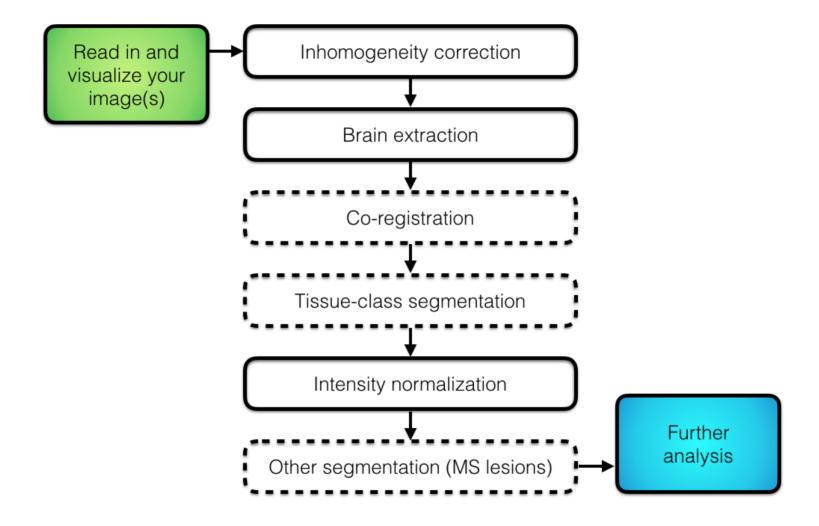
# **Intensity Normalization**

Processing math: 100%

### **Overall Pipeline**



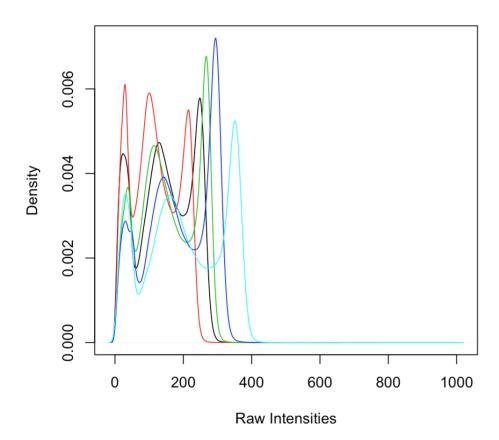
### Intensity normalization

- · Conventional MRI intensites (T1-w, T2-w, PD, FLAIR) are acquired in arbitrary units
- · Images are not comparable across scanners, subjects, and visits, even when the same protocol is used.
  - This affects algorithm performance, prediction, inference.
  - Even simple things like thresholding an image
- Intensity normalization brings the intensities to a common scale across people.
- · In this tutorial we will normalize intensities within subject using two methods:
  - Whole-brain normalization
  - White Stripe normalization (Shinohara et al. 2014).

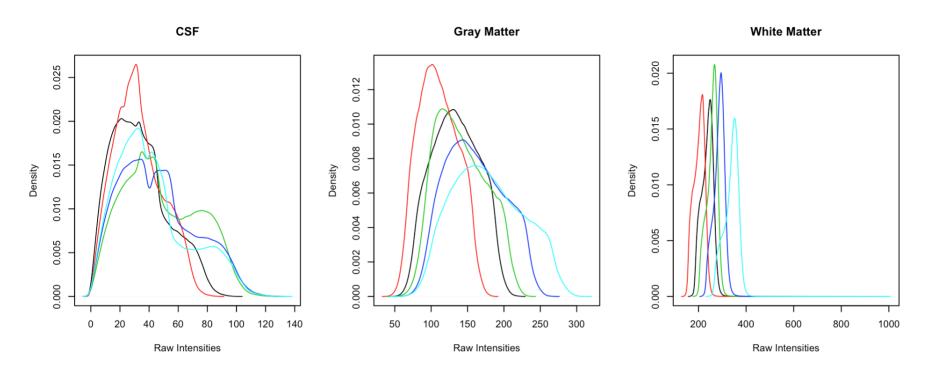
### Visualizing whole-brain intensities (each line is a person)

- · We will work with the T1-w images from the training data.
- · Full brain densities are mixtures of the three tissue class distributions.

#### Distribution of all Voxels in Brain Mask



### Visualizing the intensities by tissue class



And these are all the same scanner/protocol!

### Whole-brain normalization

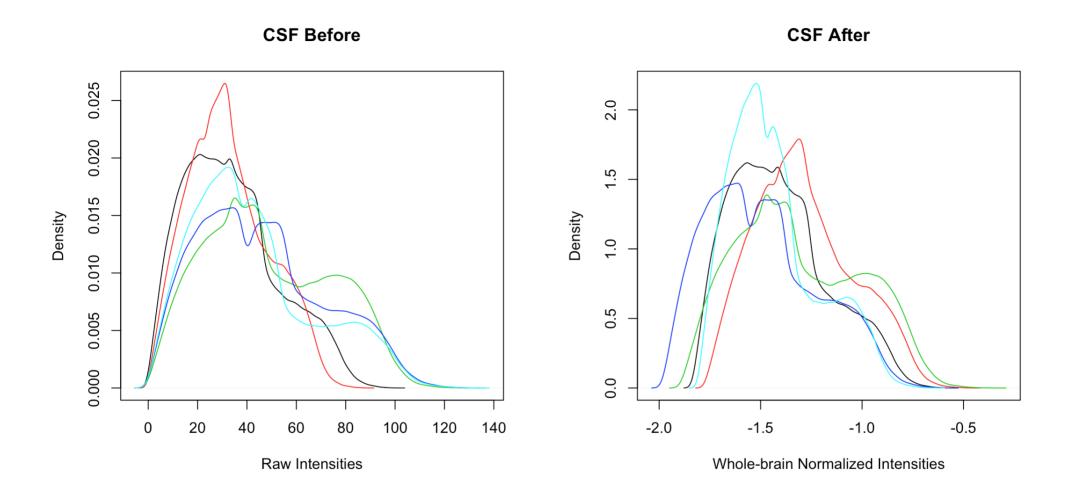
· Let's Z-score each voxel using mean  $\mu_{WB}$  and standard deviation  $\sigma_{WB}$  computed from all voxels in the brain mask.

$$T1_{WB} = T1 - \mu_{WB}\sigma_{WB}$$

- zscore\_img is a function in neurobase that does this.
- · It takes an image and a binary mask. The default is to use all voxels in the brain mask.

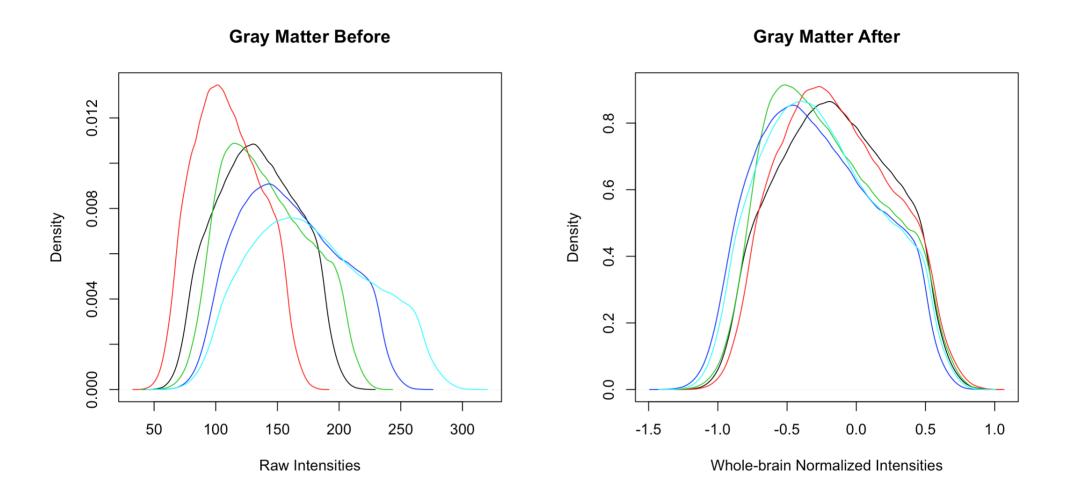
```
zscore_img(img = img, mask = mask)
```

### Whole-brain normalized intensities



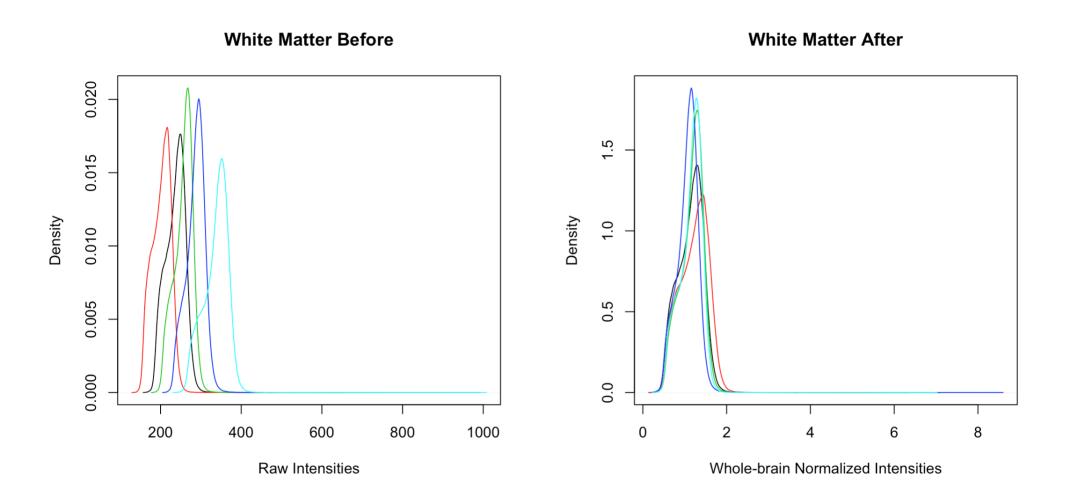
### Whole-brain normalized intensities

· Gray matter distributions are more comparable.



### Whole-brain normalized intensities

· White matter distributions are more comparable.

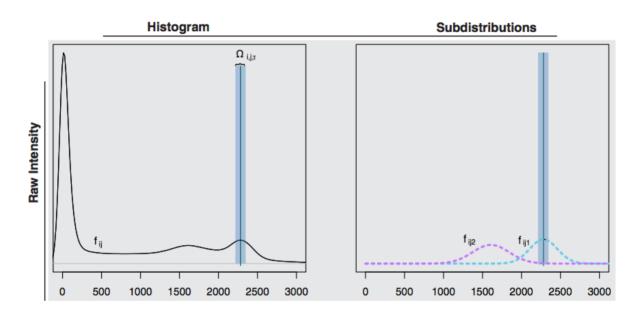


### Other Normalizations: White Stripe

- · Whole-brain normalization may be sensitive to outliers.
- Lesions in MS can have very high intensities, which lead to bad estimates of mean/variance
  - Other more robust transformations may be used, such as using the median to center, IQR to scale, etc.
- White Stripe (Shinohara et al. 2014) is based on parameters obtained from a sample of normal appearing white matter (NAWM), which is robust to outliers.
  - The idea is to make normal appearing white matter comparable across subjects and visits.

### Procedure:

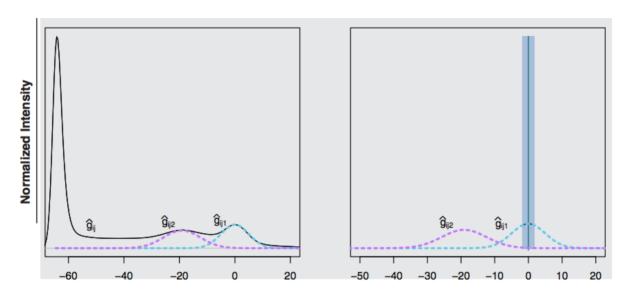
1. Find white matter area on histogram



#### Procedure:

- 1. Find white matter area on histogram
- 2. Estimate mean  $\mu_{\mbox{WS}}$  and variance  $\sigma_{\mbox{WS}}$  of voxel intensities in that area
- 3. Normalize with those means/variances:  $T1_{WS} = T1 \mu_{WS}\sigma_{WS}$

- After normalization, NAWM will have a standard normal distribution and units will be in standard deviations of NAWM.
- · Gray matter and CSF distributions may not be comparable after White Stripe.



#### Procedure:

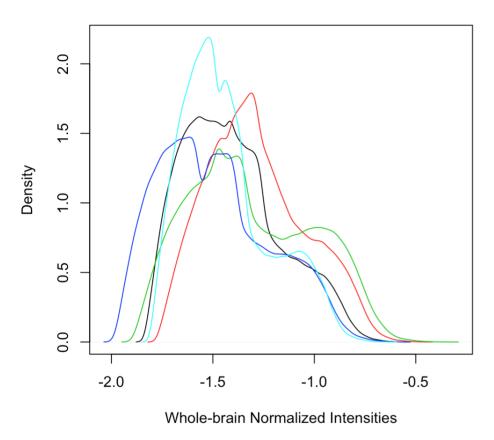
- 1. Find white matter area on histogram
- 2. Estimate mean  $\mu_{WS}$  and variance  $\sigma_{WS}$  of voxel intensities in that area
- 3. Normalize with those means/variances:  $T1_{WS} = T1 \mu_{WS}\sigma_{WS}$
- After normalization, NAWM will have a standard normal distribution and units will be in standard deviations of NAWM.
- · Gray matter and CSF distributions may not be comparable after White Stripe.

```
library(WhiteStripe)
ind = whitestripe(img = t1, type = "T1", stripped = TRUE)$whitestripe.ind
ws_t1 = whitestripe_norm(t1, indices = ind)
```

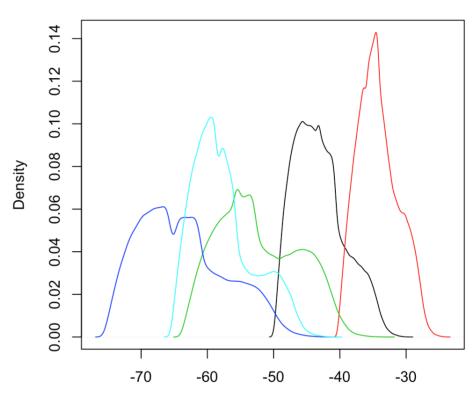
- The whitestripe function takes an image, image type (in our case T1), and a logical indicating whether the image has been skull stripped.
- The indicies of voxels in the NAWM used for estimating the normalization parameters are located in the list element \$whitestripe.ind.
- The function whitestripe\_norm takes an image and the indicies from a call to whitestripe and returns the White Stripe normalized image as a nifti.

# WhiteStripe normalized intensities





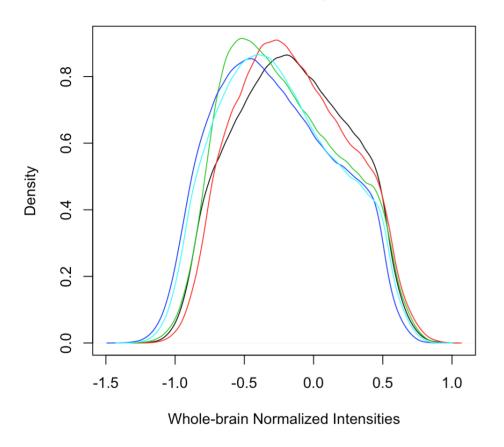
White Stripe: CSF



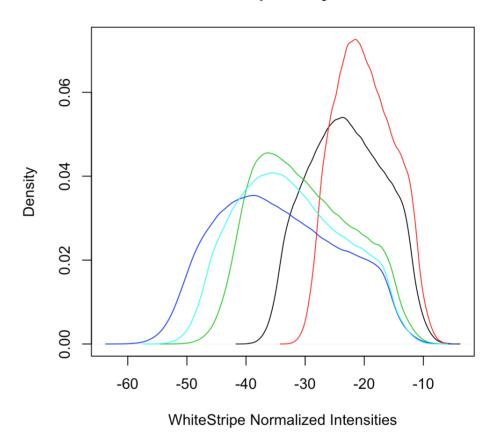
WhiteStripe Normalized Intensities

# WhiteStripe normalized intensities

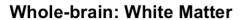
Whole-brain: Gray Matter

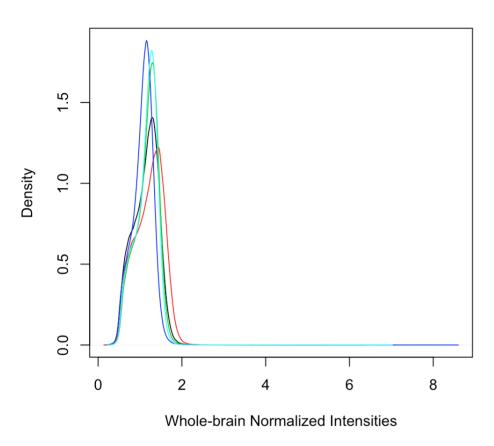


#### White Stripe: Gray Matter

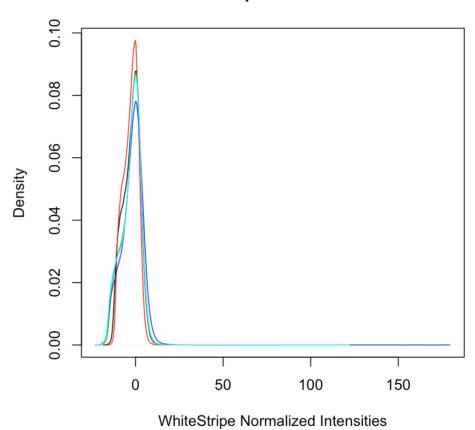


# WhiteStripe normalized intensities





#### White Stripe: White Matter



#### Conclusions

- Intensity normalization is an important step in any image analysis with more than one subject or time point to ensure comparability across images.
- White Stripe normalization may work better and have better interpretation than whole-brain normalization for subsequent lesion segmentation algorithms and analysis.
- Other intensity normalization methods that make intensites comparable across subjects for all tissues exist.
  - RAVEL, which is an extension of WhiteStripe is one example (Fortin et al. 2016).
  - Located at https://github.com/Jfortin1/RAVEL
  - This was shown to have better comparability than histogram matching

### Website

http://johnmuschelli.com/imaging\_in\_r

### References

Fortin, Jean-Philippe, Elizabeth M Sweeney, John Muschelli, Ciprian M Crainiceanu, Russell T Shinohara, Alzheimer's Disease Neuroimaging Initiative, and others. 2016. "Removing Inter-Subject Technical Variability in Magnetic Resonance Imaging Studies." 132. Elsevier: 198–212.

Shinohara, Russell T, Elizabeth M Sweeney, Jeff Goldsmith, Navid Shiee, Farrah J Mateen, Peter A Calabresi, Samson Jarso, et al. 2014. "Statistical Normalization Techniques for Magnetic Resonance Imaging." 6. Elsevier: 9–19.