

# Intensity Normalization

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# Intensity normalization

- ▶ Conventional MRI intensities (T1-w, T2-w, PD, FLAIR) are acquired in arbitrary units, making the images not comparable across scanners and visits.
- ▶ Intensity normalization brings the intensities to a common scale.

# Goals of this tutorial

- ▶ Visualize the intensities using boxplots and densities
- ▶ Apply the WhiteStripe intensity normalization (Shinohara et al. 2014)

## Reading in the images

- ▶ For the moment, we will work with the T1-w images from the training data.
- ▶ `vals` below is restricting to T1-w voxels that are within the brain mask.

```
library(ms.lesion)
library(neurobase)
library(WhiteStripe)
fnames = get_image_filenames_list_by_subject(group = "train",
  type = "coregistered")
t1s = lapply(fnames, function(x) readnii(x["MPRAGE"]))
tissues = lapply(fnames, function(x) readnii(x["Tissue_Classification"]))
masks = lapply(fnames, function(x) readnii(x["Brain_Mask"]))

vals = mapply(function(t1, mask){
  mask_vals(t1, mask)
}, t1s, masks, SIMPLIFY = FALSE)
```

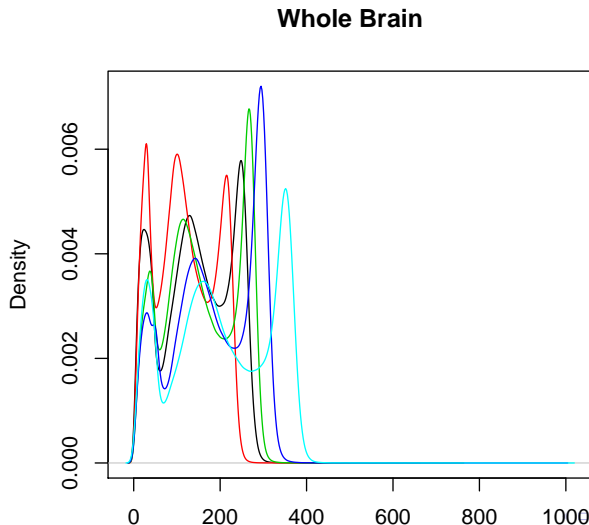
## Code for plotting the data

```
plot_densities = function(dens, xlab = "Raw Intensities",  
                           main = "Whole Brain") {  
  range_x = sapply(dens, function(d) range(d$x))  
  range_x = range(range_x)  
  range_y = sapply(dens, function(d) range(d$y))  
  range_y = range(range_y)  
  plot(dens[[1]], xlim = range_x, ylim = range_y,  
        xlab = xlab, main = main)  
  for (idens in 2:length(dens)) {  
    lines(dens[[idens]], col = idens)  
  }  
}  
  
plot_boxplots = function(vals,  
                           main = "Whole Brain") {  
  boxplots <- lapply(vals, boxplot, outline = FALSE, plot = FALSE)  
  boxplots = lapply(boxplots, function(x) x$stats)  
  boxplots <- do.call(cbind, boxplots)  
  boxplot(boxplots, main = main)
```

# Visualizing the intensities

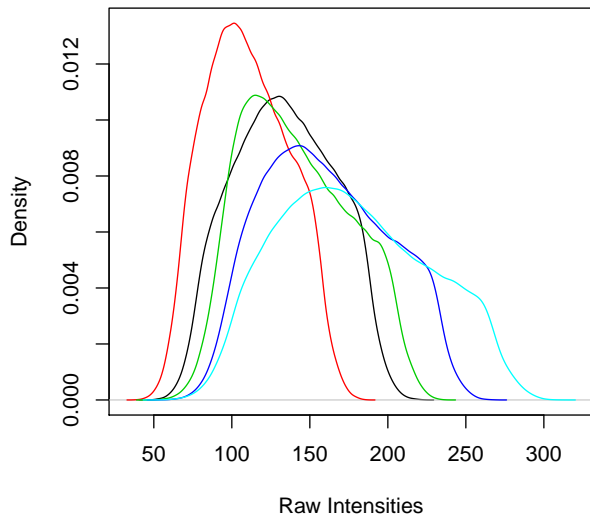
- Full brain densities are mixtures of the tissue class distributions.

```
plot_densities(dens)
```



# Visualizing the intensities by tissue class

**Gray Matter**

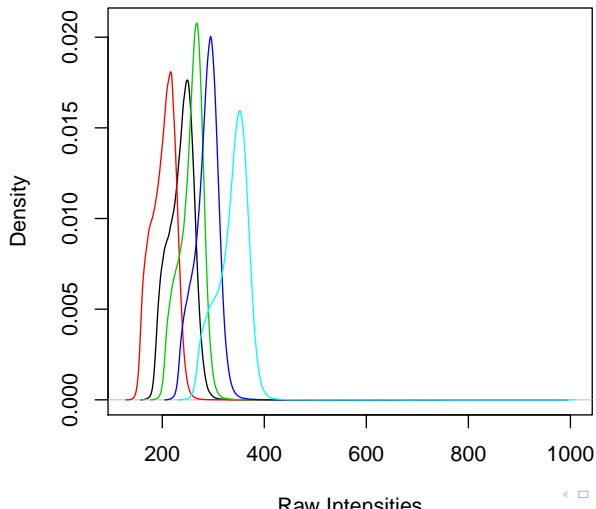


**Gray Matter**

# Visualizing the intensities by tissue class

- Notice the complete non-overlap between some subjects' white matter distributions.

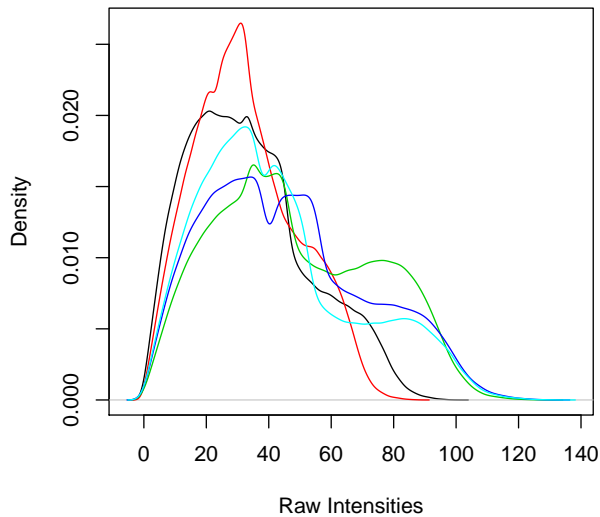
## White Matter





# Visualizing the intensities by tissue class

**CSF**



**CSF**

# Whole-brain normalization

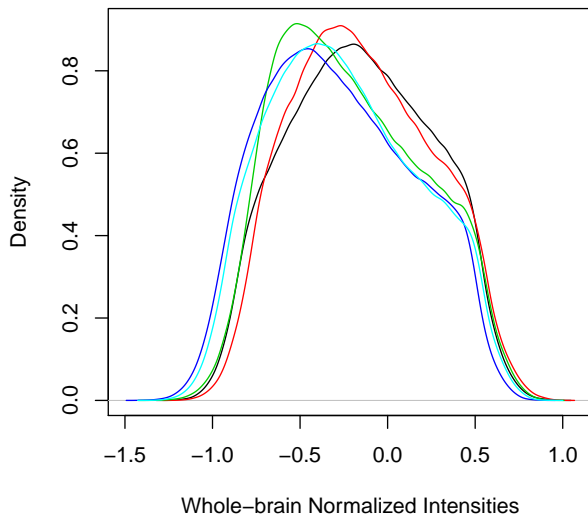
- ▶ Let's Z-score each voxel using mean and standard deviation computed from all voxels in the brain mask.
- ▶ `zscore_img` is a function in `fslr` that does this.

```
t1_norm = mapapply(function(img, mask){  
  zscore_img(img = img, mask = mask, margin = NULL)  
}, t1s, masks, SIMPLIFY = FALSE)
```

# Whole-brain normalized intensities

- Gray matter distributions are more comparable.

## Gray Matter

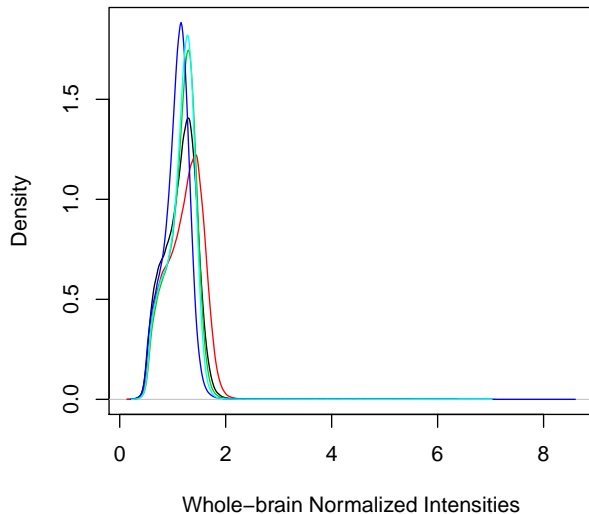


## Gray Matter

# Whole-brain normalized intensities

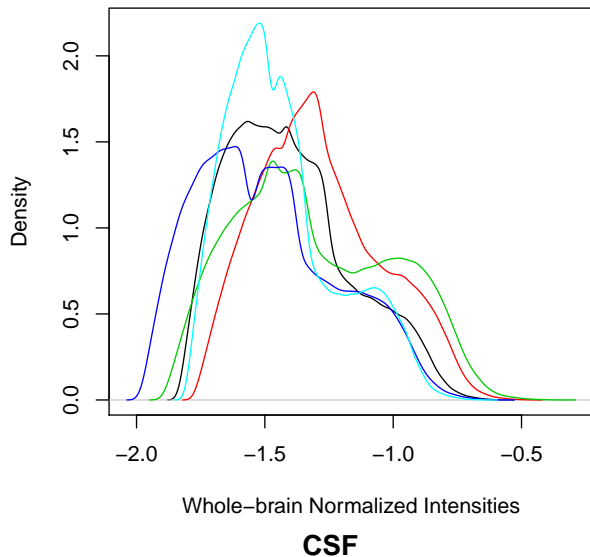
- ▶ White matter distributions are more comparable.

## White Matter



# Whole-brain normalized intensities

## CSF



# Why White Stripe?

- ▶ Whole-brain normalization may reduce signal in the higher intensities, i.e., less contrast between lesion intensities and normal appearing white matter.
- ▶ White Stripe normalization is based on parameters obtained from a sample of normal appearing white matter.
- ▶ The idea is to make normal appearing white matter comparable across subjects while preserving high lesion intensity contrast to aid automatic segmentation algorithms.

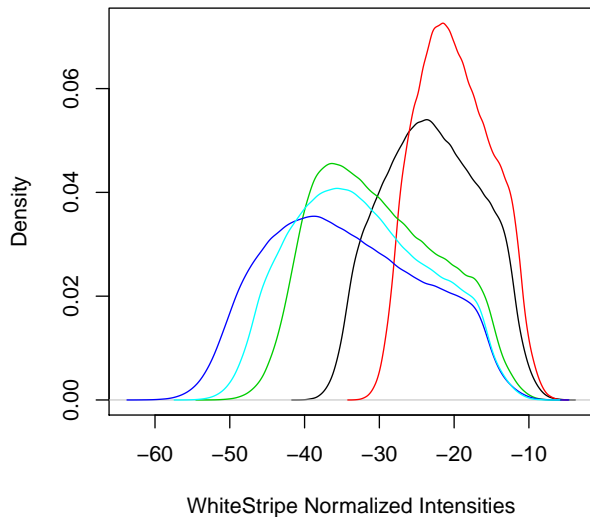
# White Stripe normalization

- ▶ We normalize each voxel using the mean and standard deviation computed from normal appearing white matter voxels.
- ▶ Thus, normal appearing white matter will have a standard normal distribution.
- ▶ Units will correspond to variability (standard deviation) of normal appearing white matter.
- ▶ Gray matter and CSF distributions, however, may not be comparable across subjects.

```
ws_norm = function(t1) {  
  ind = whitestripe(img = t1,  
                    type = "T1",  
                    stripped = TRUE)$whitestripe.ind  
  whitestripe_norm(t1, indices = ind)  
}  
t1_ws_norm = lapply(t1s, ws_norm)
```

# WhiteStripe normalized intensities

**Gray Matter**

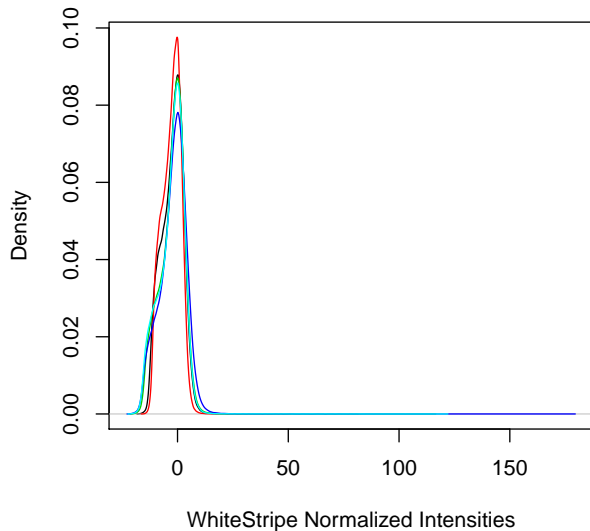


**Gray Matter**



# WhiteStripe normalized intensities

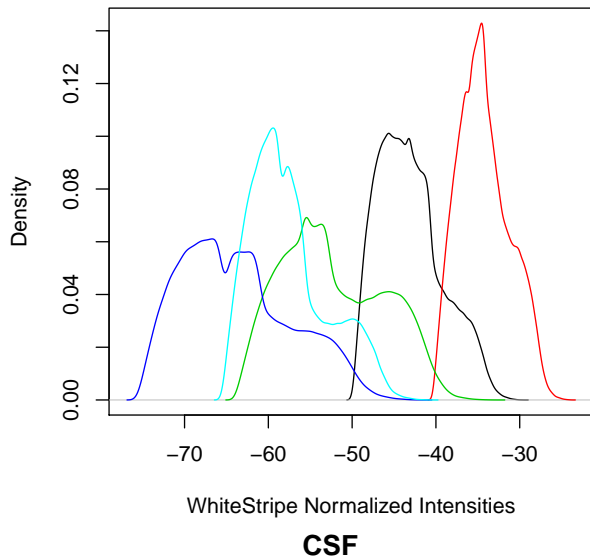
**White Matter**



**White Matter**

# WhiteStripe normalized intensities

**CSF**



# 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 for subsequent lesion segmentation algorithms or analysis.
- ▶ Newer methods exist, such as RAVEL, which extends WhiteStripe to make intensities comparable across subjects for all tissues (Fortin et al. 2016).

# 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." *NeuroImage* 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." *NeuroImage: Clinical* 6. Elsevier: 9–19.