Tissue-Class Segmentation

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Image segmentation

- We are often interested in subdividing or segmenting the brain into meaningful biological regions of interest (ROIs) for an analysis.
- Examples: tissue segmentation, segmentation of gray matter structures, segmentation of pathology (MS lesions, tumors, ...)

Goals of this tutorial

- Perform tissue segmentation in R using FSL and ANTs.
- ▶ Discuss multi-atlas label fusion techniques for segmentation.
- Perform automatic MS lesion segmentation using OASIS.

Loading Data

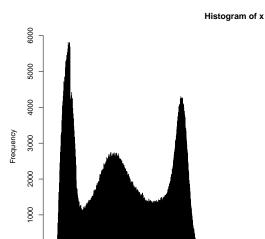
▶ Let's read in the training T1s and brain mask for subject 05.

```
library(ms.lesion)
library(neurobase)
library(fslr)
library(scales)
library(extrantsr)
all_files = get_image_filenames_list_by_subject(
  group = "training",
 type = "coregistered")
files = all files$training05
t1 = readnii(files["MPRAGE"])
mask = readnii(files["Brain Mask"])
# t1s = lapply(files, function(x) readnii(x["MPRAGE"]))
# t1 = t1s[[5]]
# masks = lapply(files, function(x) readnii(x["Brain_Mask"]
\# mask = masks[[5]]
```

Tissue Segmentation: Large Outliers

- Many tissue class segmentations are based on k-means clustering.
- ▶ These methods can be skewed by large outliers.

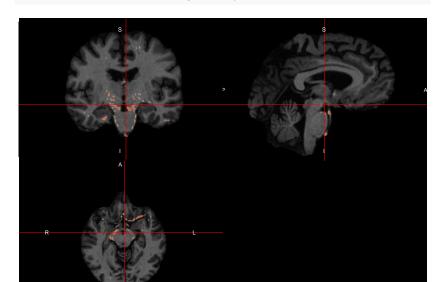
```
hist(t1, mask = mask, breaks = 2000)
```



Where are the outliers

We see some values that may have been improperly segmentated:

```
ortho2(t1, t1 > 400, xyz = xyz(t1 > 400))
```



Tissue Segmentation using FSL FAST

The fslr function fast calls fast from FSL (Zhang, Brady, and Smith 2001). The --nobias option tells FSL to not perform inhomogeneity correction.

Results

FAST assumes three tissue classes and produces an image with the three labels, ordered by increasing within-class mean intensities. In a T1 image, this results in:

Level 1: CSF

Level 2: Gray Matter

Level 3: White Matter

White Matter

ortho2(t1, t1fast == 3, col.y = alpha("red", 0.5), text = "



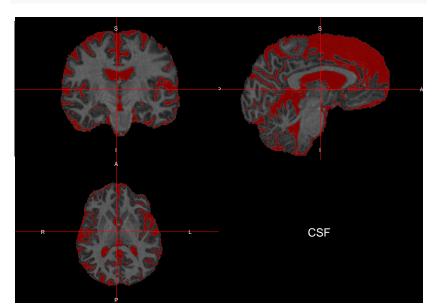
Gray Matter

```
ortho2(t1, t1fast == 2, col.y = alpha("red", 0.5), text = "
```



CSF

ortho2(t1, t1fast == 1, col.y = alpha("red", 0.5), text = "

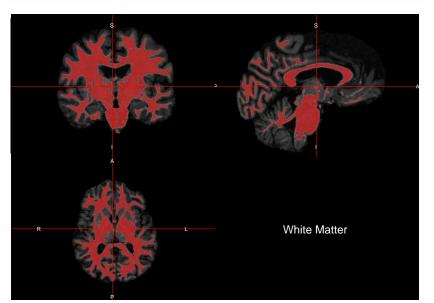


Windowing

```
rb = robust window(t1)
robust_fast = fast(t1,
              outfile = paste0(nii.stub(t1file), "_FAST"),
              opts = "--nobias")
separate = separate img(robust fast, levels = 1:3)
names(separate) = c("CSF", "GM", "WM")
L = c(MPRAGE = list(t1), separate)
L$MPRAGE = robust window(L$MPRAGE)
r = range(L$MPRAGE)
L$MPRAGE = (L$MPRAGE - r[1])/(r[2] - r[1])
dd = dropEmptyImageDimensions(t1fast > 0, other.imgs = L)
L = dd$other.imgs
multi_overlay(L, z = 58, text = names(L), text.x = 0.5, text.x = 0.5)
              text.cex = 2.5)
```

White Matter

ortho2(t1, robust_fast == 3, col.y = alpha("red", 0.5), tex



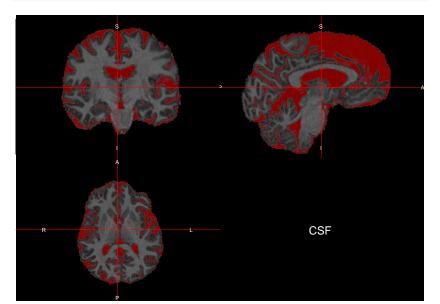
Gray Matter

ortho2(t1, robust_fast == 2, col.y = alpha("red", 0.5), tex



CSF

ortho2(t1, robust_fast == 1, col.y = alpha("red", 0.5), text



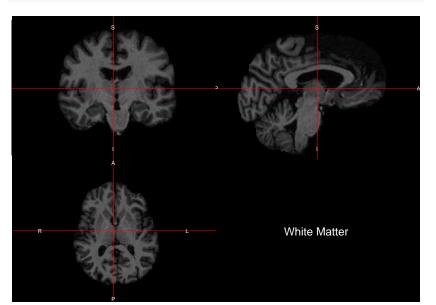
Tissue Segmentation using ANTsR, extrantsr

▶ Uses Atropos (Avants et al. 2011)

```
t1\_otropos = otropos(a = t1, x = mask) # using original da t1seg = t1\_otropos\$segmentation
```

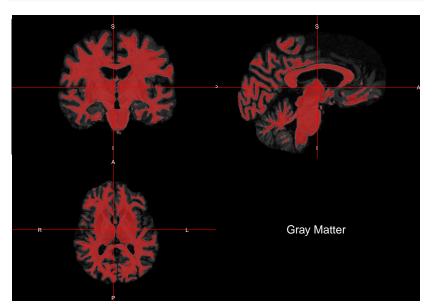
White Matter

```
ortho2(t1, t1seg == 3, col.y = alpha("red", 0.5), text = "
```



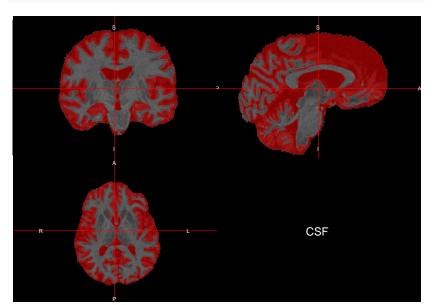
Gray Matter

```
ortho2(t1, t1seg == 2, col.y = alpha("red", 0.5), text="Grader")
```



CSF

ortho2(t1, t1seg == 1, col.y = alpha("red", 0.5), text="CSI



Tissue Segmentation using ANTsR, extrantsr

```
robust_t1_otropos = otropos(a = rb, x = mask) # using robust
robust_t1seg = robust_t1_otropos$segmentation
```

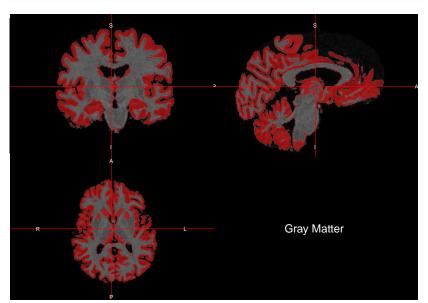
White Matter

```
ortho2(t1, robust_t1seg == 3, col.y = alpha("red", 0.5), te
```



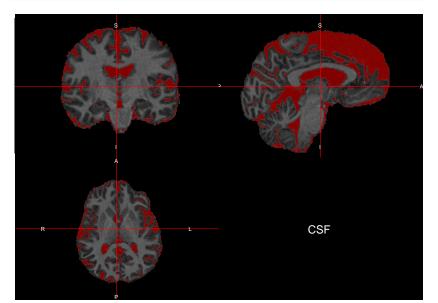
Gray Matter

```
ortho2(t1, robust_t1seg == 2, col.y = alpha("red", 0.5), te
```



CSF

ortho2(t1, robust_t1seg == 1, col.y = alpha("red", 0.5), te



References

Avants, Brian B, Nicholas J Tustison, Jue Wu, Philip A Cook, and James C Gee. 2011. "An Open Source Multivariate Framework for N-Tissue Segmentation with Evaluation on Public Data." *Neuroinformatics* 9 (4). Springer: 381–400.

Zhang, Yongyue, Michael Brady, and Stephen Smith. 2001. "Segmentation of Brain MR Images Through a Hidden Markov Random Field Model and the Expectation-Maximization Algorithm." Medical Imaging, IEEE Transactions on 20 (1): 45–57. http://ieeexplore.ieee.org/xpls/abs_all.jsp?arnumber=906424.