# Inhomogeneity Correction

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#### Inhomogeneity correction

- Scans can have nonuniform intensities throughout the brain
- Usually low frequency smooth over the brain (assumed)
- ▶ Referred to as bias, bias field, or inhomogeneity

#### MS Lesion

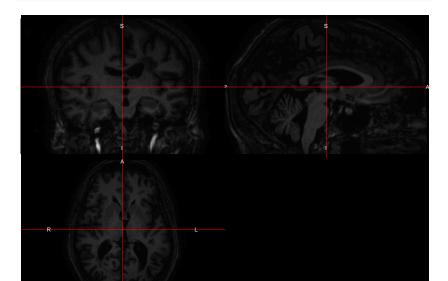
Let's read in the T1 image from a MS lesion data set:

```
library(ms.lesion)
library(neurobase)
files = get_image_filenames_list_by_subject()$training01
t1_fname = files["MPRAGE"]
t1 = readnii(t1_fname)
```

## **Bright Data**

We see areas of brightness, but we also see that an artifact

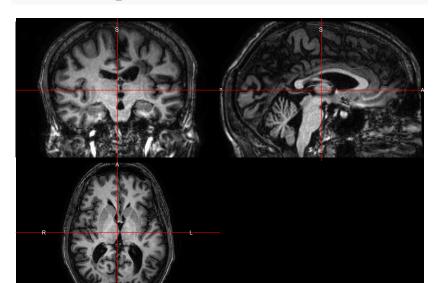
ortho2(t1)



# Image Data

Let's dampen this artifact and replot the data:

ortho2(robust\_window(t1))



## N4 Inhomogeneity Correction

We will use N4: Improved N3 Bias Correction (Tustison et al. 2010).

The model assumed in the N4 is:

$$v(x) = u(x)f(x) + n(x)$$

where v is the given image, u is the uncorrupted image, f is the bias field, and n is the noise (assumed to be independent and Gaussian) and x is a location in the image.

### N4 Inhomogeneity Correction

The data is log-transformed and assuming a noise-free scenario, we have:

$$log(v(x)) = log(u(x)) + log(f(x))$$

- N4 uses a B-spline approximation of the bias field
- It iterates until a convergence criteria is met
  - when the updated bias field is the same as the last iteration
- It outputs the data back in the original units (not log-transformed)

#### Bias Field Correction

Here we will use the bias\_correct function in extrantsr, which calls n4BiasFieldCorrection from ANTsR.

You can pass in the image:

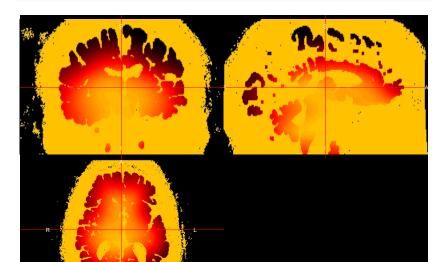
```
library(extrantsr)
bc_t1 = bias_correct(file = t1, correction = "N4")
```

or the filename:

```
bc_t1 = bias_correct(file = t1_fname, correction = "N4")
```

Here we take the ratio of the images and overlay it on the original image:

ratio = t1 / bc\_t1; ortho2(t1, ratio)



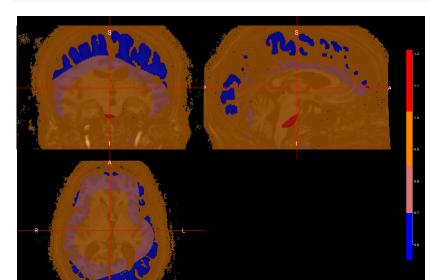
library(scales)

Here we would like to change the colors to something more descriptive. Here we will use a diverging palette and map colors to the quantiles of the ratio image:

```
q = quantile(ratio[ ratio != 0], probs = seq(0, 1, by = 0.)
q = unique(q)
# get a diverging gradient palette
fcol = scales::div_gradient_pal(low = "blue", mid = "orange")
colors = scales::alpha(fcol(seq(0,1, length = length(q) - length))
```

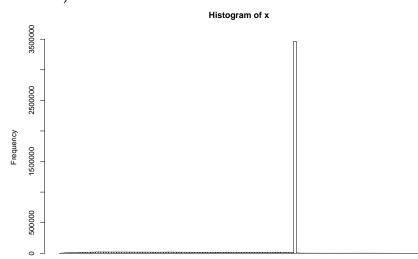
Now we put those breaks into ortho2 to plot it:

```
ortho2(t1, ratio, col.y = colors, ybreaks = q, ycolorbar =
```



### Histogram of Ratio Values

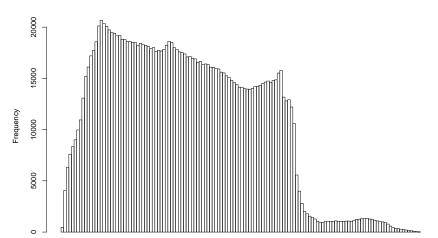
The majority of voxels have a ratio of 1 because n4BiasFieldCorrection does some implicit masking using ANTsR::getMask, and those values are unchanged (backround excluded).



Removing these, we can see what the distribution of ratios look like (most are below 1):

```
hist(ratio[ratio < 0.999 | ratio > 1.0001], breaks = 200)
```

Histogram of ratio[ratio < 0.999 | ratio > 1.0001]



##

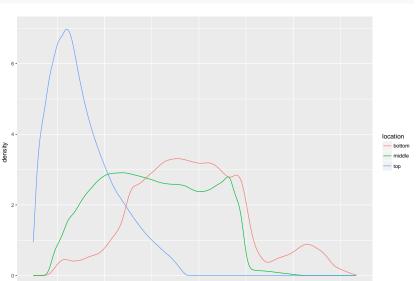
We would like to see how the ratio changes in different areas of the brain. Here we make a data.frame of voxel location and intensity. We cut the location into the bottom, middle, and top of the brain:

```
df = which(ratio < 0.999 | ratio > 1.0001, arr.ind = TRUE)
df = cbind(df, value = ratio[df])
df = data.frame(df, stringsAsFactors = FALSE)
df$location = cut(df$dim3, breaks = c(0, 38, 76, 115),
                  labels = c("bottom", "middle", "top"))
```

```
dim1 dim2 dim3
                   value location
## 1
     131
           97
                1 0.8248509
                             bottom
## 2 132
         97
             1 0.8217719
                             bot.t.om
## 3
    133
         97
              1 0.8195554
                             bottom
## 4
    134
         97
              1 0.8176278
                             bot.t.om
## 5
    135
         97
              1 0.8159310
                             bottom
##
     136
           97
                1 0.8149298
                             bot.t.om
  6
```

Let's plot these with a density plot for each different location:

ggplot(df, aes(x = value, colour = location)) + geom\_line(s)



#### Conclusions

- Inhomogeneity correction is one of the first steps of most structural MRI pipelines
- Inhomogeneity can cause problems for other methods/segmentation
- Corrections try to make tissues of the same class to have similar intensities
- You may also want to run corrections after skull stripping on the brain only
  - we will do this in the brain extraction lecture
  - correction before skull-stripping may be necessary and can improve after correction

#### References

Tustison, Nicholas J., Brian B. Avants, Philip A. Cook, Yuanjie Zheng, Alexander Egan, Paul A. Yushkevich, and James C. Gee. 2010. "N4ITK: Improved N3 Bias Correction." *IEEE Transactions on Medical Imaging* 29 (6): 1310–20. doi:10.1109/TMI.2010.2046908.