



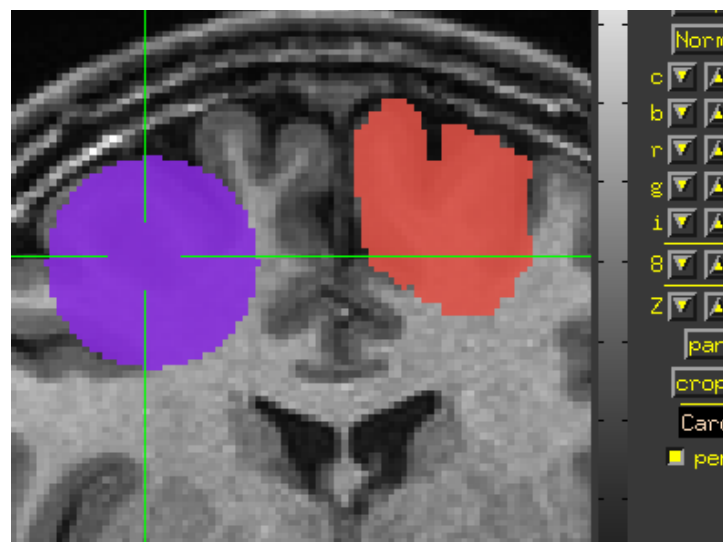
AFNI



Didactics and Demonstrations

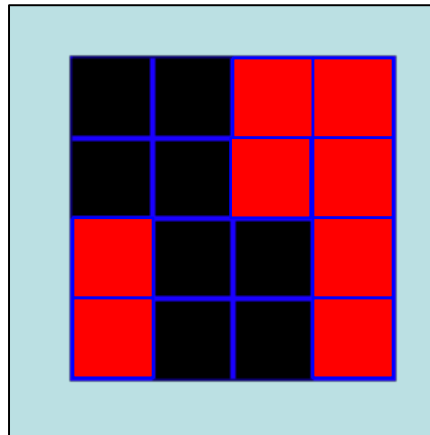
Regions of Interest (ROIs)



Introduction



What is an ROI?

- An ROI is a "region of interest" usually used as a mask of voxels
- In AFNI, ROIs are stored as any other dataset (.HEAD/.BRICK, .nii, ...) , typically with positive integer values for voxels to consider.
 - Zero values are outside the mask.
 - Positive values are inside the mask

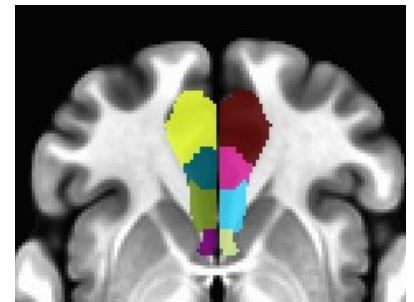
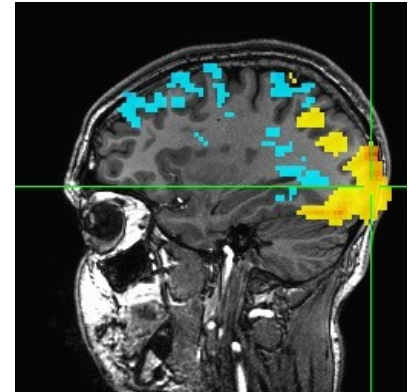


-  Voxel with value of 0
-  Voxel with value of 1

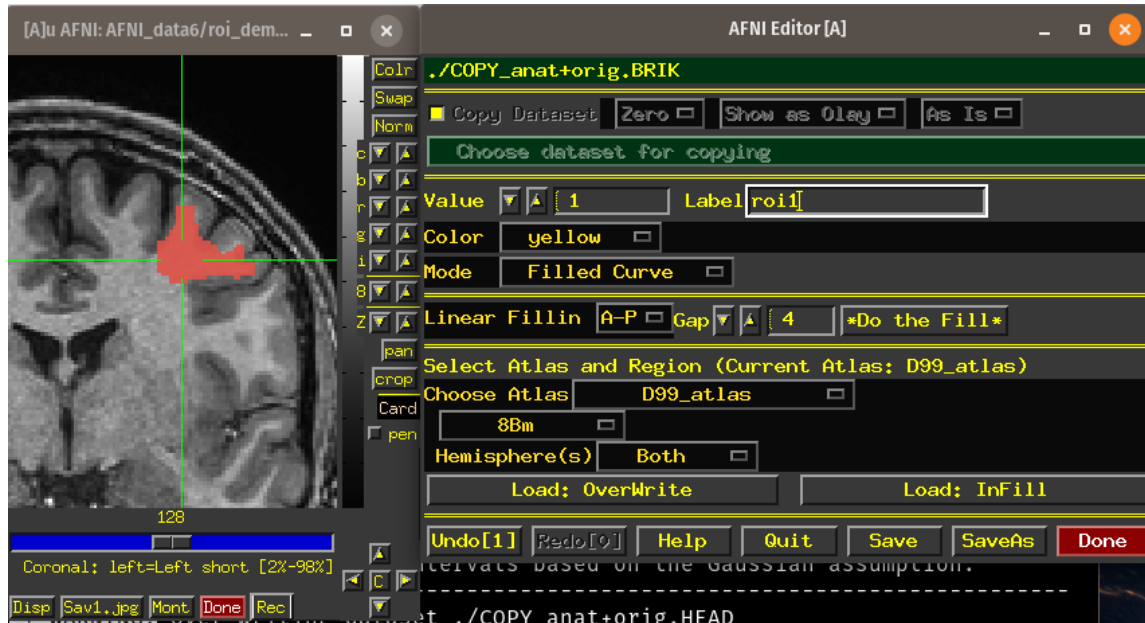
- Note any dataset can be considered as an ROI if it is non-zero in areas in which you are interested.
- Usually stored as byte integers to save disk space and memory

Creating Regions of Interest

- **Method 1: Draw.** Draw ROIs based on anatomical structures, then analyze functional datasets within these regions
 - Previous studies may identify particular regions
 - This method relies on ‘a priori’ assumptions about localization of brain function and neuroanatomy knowledge
- **Method 2: Cluster.** Analyze functional dataset for *entire* brain first, then use clusters of ‘activity’ (neighboring voxels with values above a threshold in some statistical map)
 - Analyze the entire brain first and *then* find interesting areas of activity and do further analyses on those areas
 - Use **3dClusterize** or the AFNI graphical interface button **[Clusterize]** to find larger “blobs” or clusters of activity
 - Apply the clusters from a localizer task to apply to a separate experiment
- **Method 3: Atlas.** Use atlases to select anatomical regions
 - Use **whereami** program or symbolic notation to create masks on the command line (3dcalc, 3dresample, ...)
 - Parcellations and classifications from FreeSurfer, 3dSeg, 3dkmeans, etc.
 - Spheres around published coordinates



Drawing ROIs



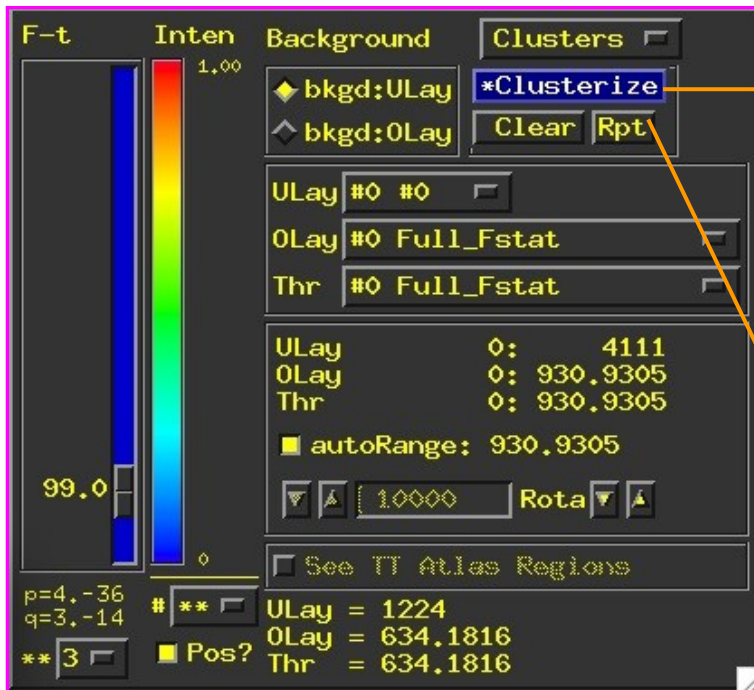
Use AFNI's ROI Draw Dataset plugin (see video)
For large projects, consider drawing tools like a stylus with a touchscreen or Wacom Cintiq.

Also other drawing tools that support NIFTI output can be used - MIPAV, Amira,

Clustering

Clusterize

- The Clusterize button on the main AFNI graphical interface gives users a quick and easy way to locate clusters of activity in a functional dataset. Once the user sets the clusterize parameters, a complete cluster “report” is given, which details the number of clusters found, based on these parameters.



Clusterize Features

```

Voxels survived clustering = 2089
Voxels edited out         = 386
Min cluster size (voxels) = 327
Max cluster size          = 1762
Number of clusters kept   = 2
  
```

#1 xyz Peak 3dclust SaveTable Clust SaveMask Done

Aux Dataset From 0 To 99999 Mean

[No Auxiliary Dataset selected yet]

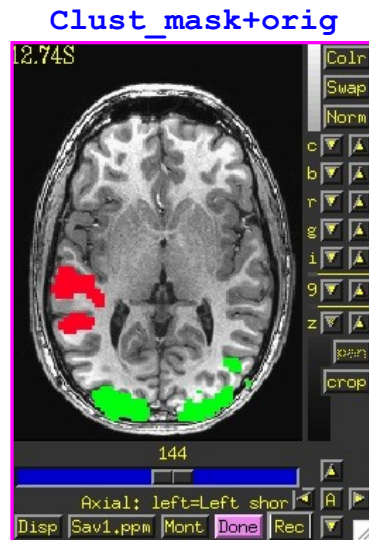
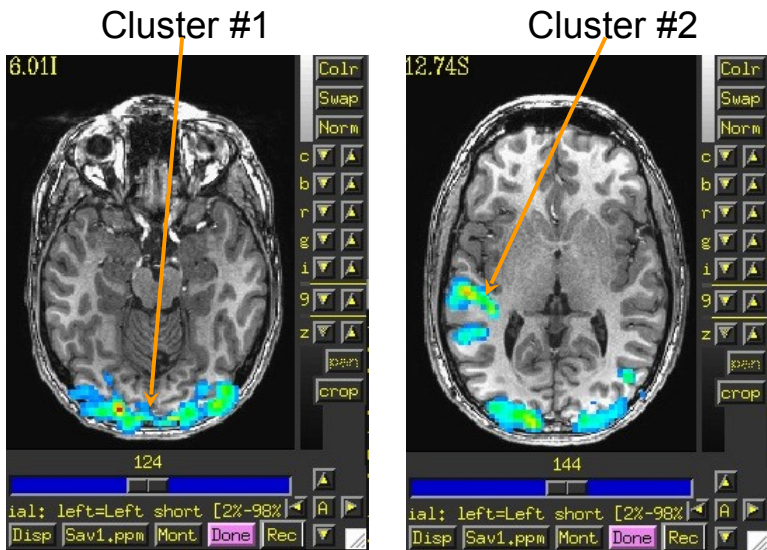
1: 1762 vox	+19.3	+78.0	-5.7	Jump	Flash	Plot	Save
2: 327 vox	+52.3	+6.5	+12.3	Jump	Flash	Plot	Save

Jump: sets the crosshairs to the designated xyz coordinates (default is the peak of the ROI cluster)

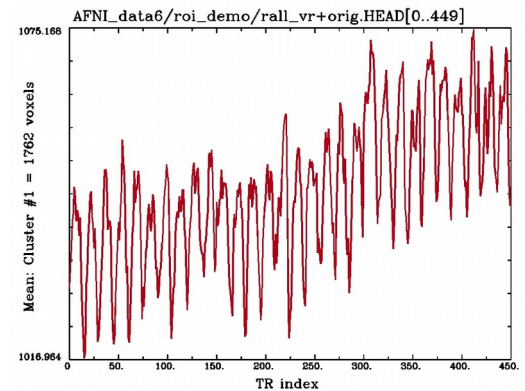
Flash: flashes the cluster voxels in the image viewer

SaveMask: Click on this button to write clusters to a mask dataset called Clust_mask+orig

Plot/Save: Allows user to load a 3D+time dataset (Aux Dset button) and plot the avg time series over a cluster. Plot can be saved in .jpg or .png format.



E.g., 3D+time dataset **rall_vr+orig** loaded and avg time series plotted for voxels within Cluster #1



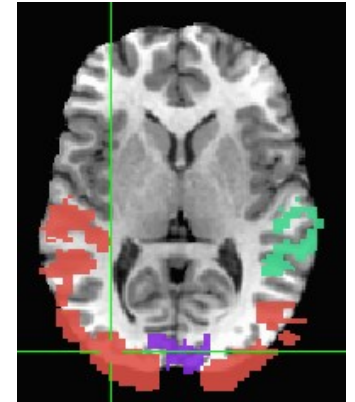
3dClusterize

The program **3dClusterize** looks for clusters – groups of voxels together that meet some threshold

Example:

```
3dClusterize -clust_nvox 200 -bisided -2.0 2.0 -ithr 2 -idat 1 \
  -NN 1 -inset func_slim+orig. -pref_map myclusters
```

The above command tells 3dClusterize to find potential cluster volumes for dataset func_slim+orig, sub-brick #2, where the threshold has been set to 2.0 (i.e., ignore voxels with an activation threshold absolute value <2.0). Voxels must be facing each other in the cluster, and cluster volume must be at least 200 voxels (these are not guidelines, just an example!).



```
# Cluster report
#[ Option summary      = bisided,-2,2,clust_nvox,200,NN1 ]
#[ Threshold value(s) = left-tail stat=-2.000000;right-tail stat=2.000000 ]
#[ Nvoxel threshold   = 200; Volume threshold = 4537.500 ]
#[ Single voxel volume = 22.688 (microliters) ]
#[ Neighbor type, NN   = 1 ]
#[ Voxel datum type   = float ]
#[ Voxel dimensions    = 2.750 mm X 2.750 mm X 3.000 mm ]
#[ Coordinates Order   = RAI ]
#[ Mean and SEM based on absolute value of voxel intensities ]
#
```

#Volume	CM RL	CM AP	CM IS	minRL	maxRL	minAP	maxAP	minIS	maxIS	Mean	SEM	Max Int	MI RL	MI AP	MI IS
16791	-11.0	13.5	9.6	-96.2	82.6	-120.0	94.5	-17.7	78.3	1.0198	0.0087	11.135	41.3	-70.5	-14.7
15563	-14.6	20.2	34.4	-93.4	66.1	-103.5	94.5	-17.7	78.3	0.4392	0.0037	-8.114	-90.7	-7.3	30.3
991	50.9	-5.4	43.7	16.6	68.8	-26.5	23.0	12.3	78.3	0.4297	0.0139	-4.655	55.1	3.7	78.3
421	48.2	-1.7	-9.9	24.8	63.3	-26.5	36.7	-17.7	6.3	0.4582	0.0126	-1.6187	57.8	-4.5	-2.7
418	-52.4	-4.0	-9.5	-74.2	-21.9	-29.3	25.7	-17.7	6.3	0.4287	0.0132	-2.2152	-24.7	-12.8	-14.7
326	-2.6	-55.6	61.6	-24.7	11.1	-89.8	-23.8	45.3	78.3	0.2991	0.011	1.4813	2.8	-87.0	60.3
206	-23.5	-30.9	-5.5	-49.4	-13.7	-45.8	-1.8	-17.7	12.3	0.6163	0.0445	-4.0194	-16.4	-40.3	-14.7
# 34716	-10.8	14.3	16.8							0.7196	0.0048				

(similar alternatives in AFNI 3dclust and 3dmerge)

Creating Regions of Interest from Atlases - whereami

`whereami` can extract ROIs from atlases using symbolic notation

```
whereami -mask_atlas_region TT_Daemon:left:amy
```

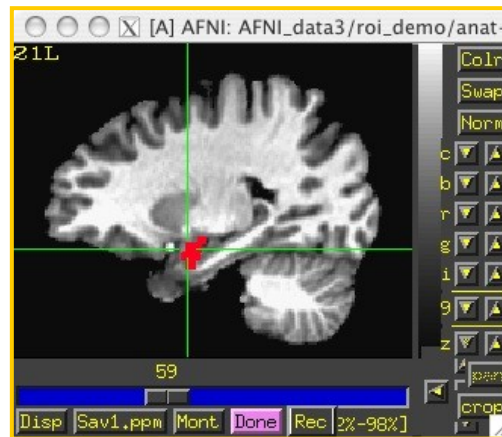
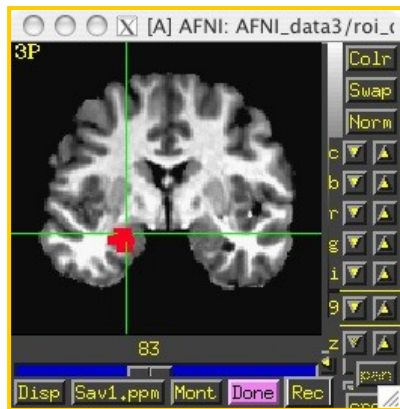
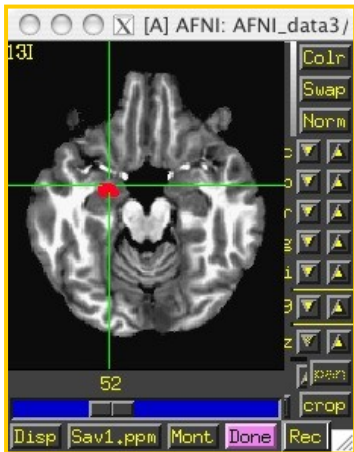
Use the Talairach-Tourneaux atlas (`TT_Daemon`) to create an ROI of the left amygdala.

Find available atlases with `whereami -show_atlases`

Find available regions with `whereami -show_atlas_code`

Another example with approximate name and prefix:

```
whereami -mask_atlas_region MNI_Glasser_HCP_v1.0::L_front_opercular \  
-prefix mniglass_lfront_oper
```



`TT_Daemon.amy.1.tlrc`

You can also specify atlas-based ROI masks directly like this:

```
3dcalc -a ~/abin/MNI_Glasser_HCP_v1.0.nii.gz '<169>' -expr a \  
-prefix left_front_operc
```

or this way is preferable:

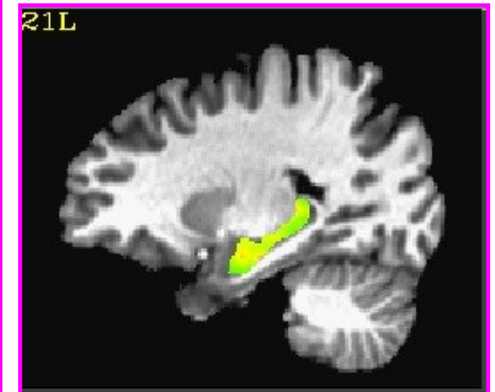
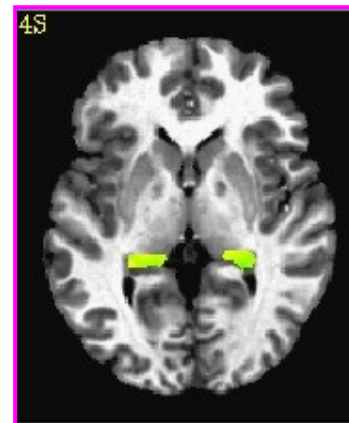
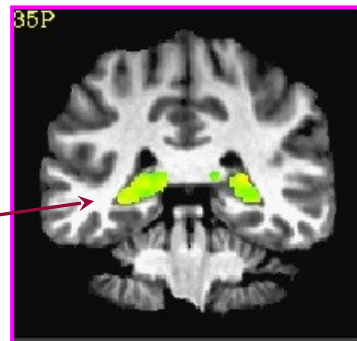
```
3dcalc -a \  
~/abin/MNI_Glasser_HCP_v1.0.nii.gz '<L_Area_Frontal_Opercular>' \  
-expr a -prefix left_front_operc2
```

Another example:

```
3dcalc -prefix nice_roi \  
-a 'CA_N27_ML::hippo' -b 'func_FullF_1mm+tlrc' \  
-expr '(step(a)*b)'
```

ULay: anat+tlrc
OLay: nice_roi+tlrc

F-stat voxels in func
dataset that fall within the
TT left hippocampus ROI



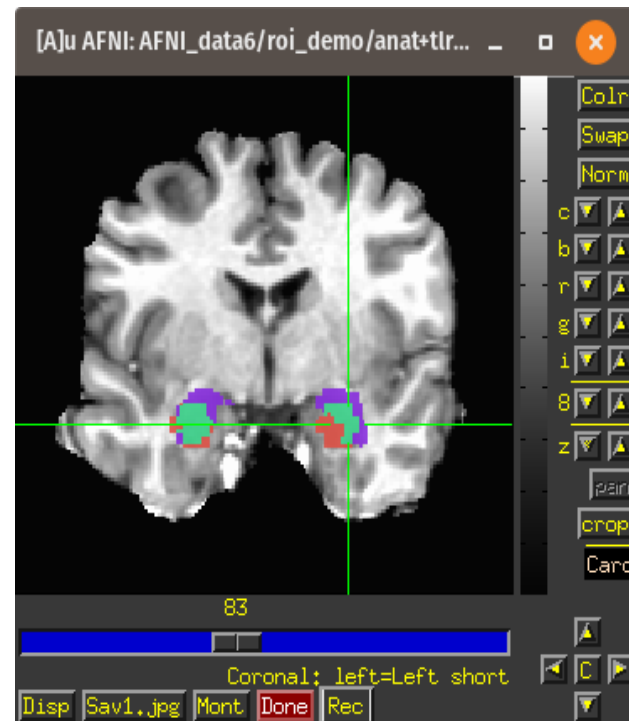
Compare the left and right amygdala between the Talairach atlas, and the CA_N27_ML atlas. The result will be **1** if a voxel is marked as amygdala in the **TT_Daemon** only, **2** if it is marked as amygdala in the **CA_N27_ML** only, and **3** where they overlap.

```
3dcalc -a 'TT_Daemon::amygdala' \
       -b 'CA_N27_ML::amygdala' \
       -expr 'step(a)+2*step(b)' \
       -prefix compare.maps
```

```
3drefit -cmap INT_CMAP compare.maps+tlrc
```



Note: `compare.maps+tlrc` displays the TT amygdala (value=1) in salmon, the N27 amygdala (value=2) in purple, and the overlap between the two atlases (value=3) in green.



Getting around with spheres

Another way to use cluster results –
make spheres from the cluster peaks or centers of mass

Try this:

```
adwarp -apar anat+tlrc -dpar func_slim+orig -dxyz 3
```

In afni GUI, select Underlay: anat, Overlay: func_slim
Switch view to Talairach

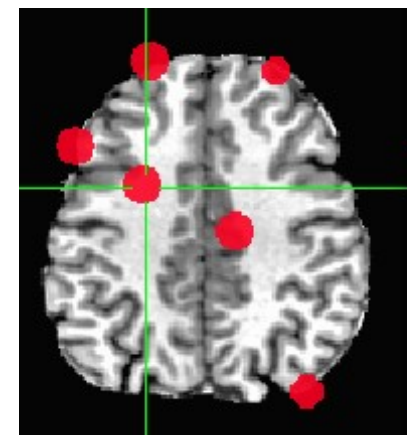
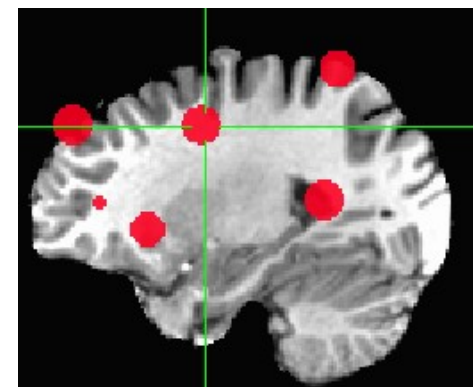
Overlay to vrel_coef, Threshold to vrel_tstat

Clusterize, and Save Table

```
1dcat Clust_table.1D'[4..6]' > Clust_PeakXYZ.1D
```

```
3dresample -orient RAI -prefix func_slim_RAI \  
-inset func_slim+tlrc
```

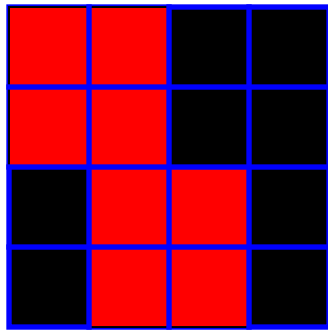
```
3dUndump -srad 7.5 -master func_slim_RAI+tlrc \  
-prefix clust_spheres -xyz Clust_PeakXYZ.1D
```



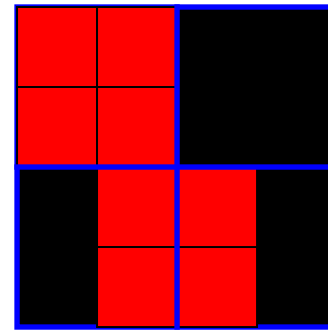
Using ROIs - Resampling

- ROIs are typically applied to functional datasets – low resolution
- Draw on anatomy or use atlas regions - high resolution

Hi-res voxel matrix



Low-res voxel matrix



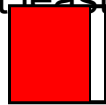
Each voxel inside the original ROI has a nonzero value

When the resolution is changed, what do you do with voxels that are only partially filled?

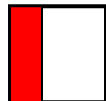
- 3dfractionize does this resolution conversion:

```
3dfractionize  -template low_res_dset+orig          \  
               -input ROI_high_res+orig           \  
               -clip 0.5 -preserve -prefix ROI_low_res
```

- ★ **-template** → The destination grid you want your ROI grid to be resampled to (we're going from high to low resolution here). Our output dataset `ROI_low_res+orig` will be written at the resolution of `func+orig`
 - (Also useful for transforming std space back to orig space with the `-warp dataset`)
- ▢ **-input** → Defines the input high-resolution dataset (that needs to be converted from high resolution to low resolution)
- ▢ **-clip 0.5** → Output voxels will only get a nonzero value if they are at least 50% filled by nonzero input voxels (you decide the percentage here). E.g., when going from high to low res, keep a label a voxel as part of the ROI if it is filled with at least 50% (or more) of the voxel value. For example:

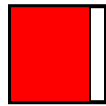


This voxel is 80% filled with the ROI value
-- keep it



This voxel is 30% filled with the ROI value
-- lose it

- ★ **-preserve** → once it has been determined that the output voxel will be part of the ROI, preserve the original ROI value of that voxel (and not some fraction of that value). This option also allows for "voting" – determine the ROI that would most fill that voxel. For example, if our ROI mask has values of "4":



This voxel is 80% filled with the ROI value -- keep it.

Without the `-preserve` option, this voxel would be given a value of "3.2" (i.e., 80% of "4").

With `-preserve`, it is labeled as "4"

- **`3dresample`** does conversion too but you have less controls for handling partial overlaps:

```
3dresample -master low_res_dset+orig \
           -prefix ROI_low_res \
           -inset ROI_high_res+orig \
           -rmode NN
```

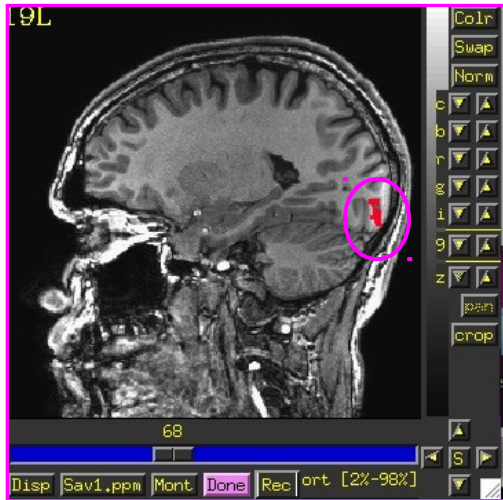
- **-master**: the destination grid we want our ROI mask resampled to
- **-prefix**: The output from `3dresample` -- in this example, a low resolution ROI mask that corresponds with the voxel resolution of our master dataset
- **-inset**: The ROI mask dataset that is being resampled from high to low resolution
- **-rmode NN**: If a voxel's "neighbor" is included in the ROI mask, include the voxel in question as well

Let's do a class example of 3dresample:

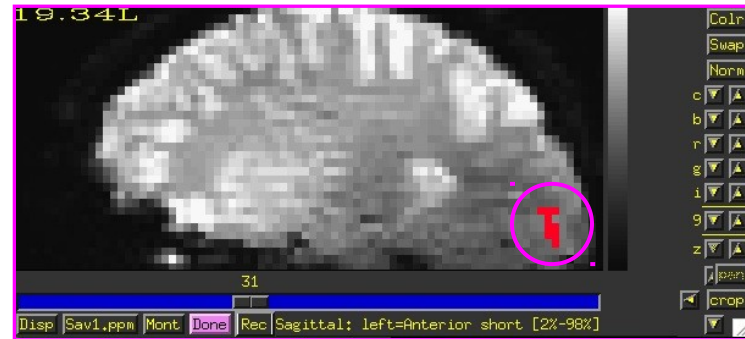
```
cd AFNI_data6/roi_demo
```

```
3dresample -master rall_vr+orig \
          -prefix anat_roi_resam \
          -inset anat_roi+orig \
          -rmode NN
```

In this class example, we want to take our ROI mask, which has a high voxel resolution of 0.9 x 0.9 x 1.2 mm, and resample to it the lower resolution of the time-series dataset, `rall_vr+orig` (2.75 x 2.75 x 3.0mm).



Before, overlay ROI is:
`anat_roi+orig`
0.9x0.9x1.2 mm voxel grid



After, overlay ROI is:
`anat_roi_resam+orig`
2.75x2.75x3.0 voxel grid

Moving ROIs - Back to the "Original" Standard Space to Native Subject Space

Useful for putting atlas regions into the native space

Affine transformations only (@auto_tlrc) – 3 ways to "inverse talairach":

3dAllineate

```
cat_matvec -ONELINE anat+tlrc::WARP_DATA > tlrc.aff12.1D
3dAllineate -1Dmatrix_apply tlrc.aff12.1D -prefix invtlrc3dAl+orig \
  -source anat+tlrc -master anat+orig
```

3dWarp

```
cat_matvec anat+tlrc::WARP_DATA > tlrc.1D
3dWarp -matvec_out2in tlrc.1D -prefix invtlrc_3dWarp+orig \
  -gridset anat+orig anat+tlrc
3drefit -view orig invtlrc_3dWarp+tlrc.
```

3dfractionize - slow but useful voting option for multiple ROIs

and manual Talairach transformations

```
3dfractionize -input anat+tlrc -warp anat+tlrc -preserve \
  -prefix invtlrc_3dffrac -template anat+orig
```

Moving ROIS - Back to the "Original" 2 Standard Space to Native Subject Space

Nonlinear and Affine transformation combinations – 3 ways to "inverse talairach":

```
# getting data to a standard space with @auto_tlrc and auto_warp.py
```

```
# affinely align to template with @auto_tlrc
```

```
@auto_tlrc -base TT_N27+tlrc -input strip_shift+orig. -no_ss \  
-init_xform AUTO_CENTER
```

```
# nonlinearly align to template
```

```
auto_warp.py -skip_affine -base TT_N27+tlrc -input strip_shift+tlrc
```

```
# 3dNwarpApply
```

```
cat_matvec -ONELINE "strip_shift+tlrc::WARP_DATA" > at_shift.1D
```

```
# one step concatenate and apply
```

```
3dNwarpApply -prefix tw3 \  
-
```

```
-nwarp 'at_shift.1D INV(awpy/anat.un.aff.qw_WARP.nii)' \  
-
```

```
-source awpy/strip_shift.aw.nii \  
-
```

```
-master strip_shift+orig. \  
-
```

```
-ainterp NN <===== Nearest neighbor interpolation for ROIs
```

Moving ROIs - Back to the "Original" 2b Standard Space to Native Subject Space

Nonlinear and Affine transformation combinations –

3dNwarpCat

```
3dNwarpCat -prefix anat_total_WARPINV2 \  
-warp2 'INV(anat_qw9_WARP+tlrc)' -warp1 'at.1D'
```

```
3dNwarpApply -prefix anat_backtoorig2 \  
-nwarp anat_total_WARPINV2+tlrc. \  
-source anat_qw9+tlrc -master anat+orig
```

3dNwarpCalc

```
3dNwarpCalc "&readwarp(anat_qw9_WARP+tlrc.)" "&invert" \  
"&read4x4(at_matrix.1D)" "&compose" "&write(combo_warp3)"  
3dNwarpApply -prefix anat_backtoorig5 -nwarp combo_warp3+tlrc. \  
-source anat_qw9+tlrc -master anat+orig
```

That's All for Now



STAY
TUNED FOR
MORE!

KEEP
CALM
AND
WASH
YOUR
HANDS



AFNI!



SUMA!



FAT
CAT