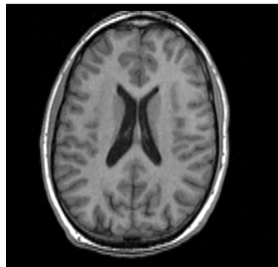
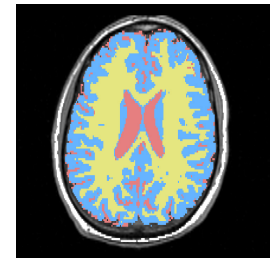


Automatic Segmentation of Brain Structures

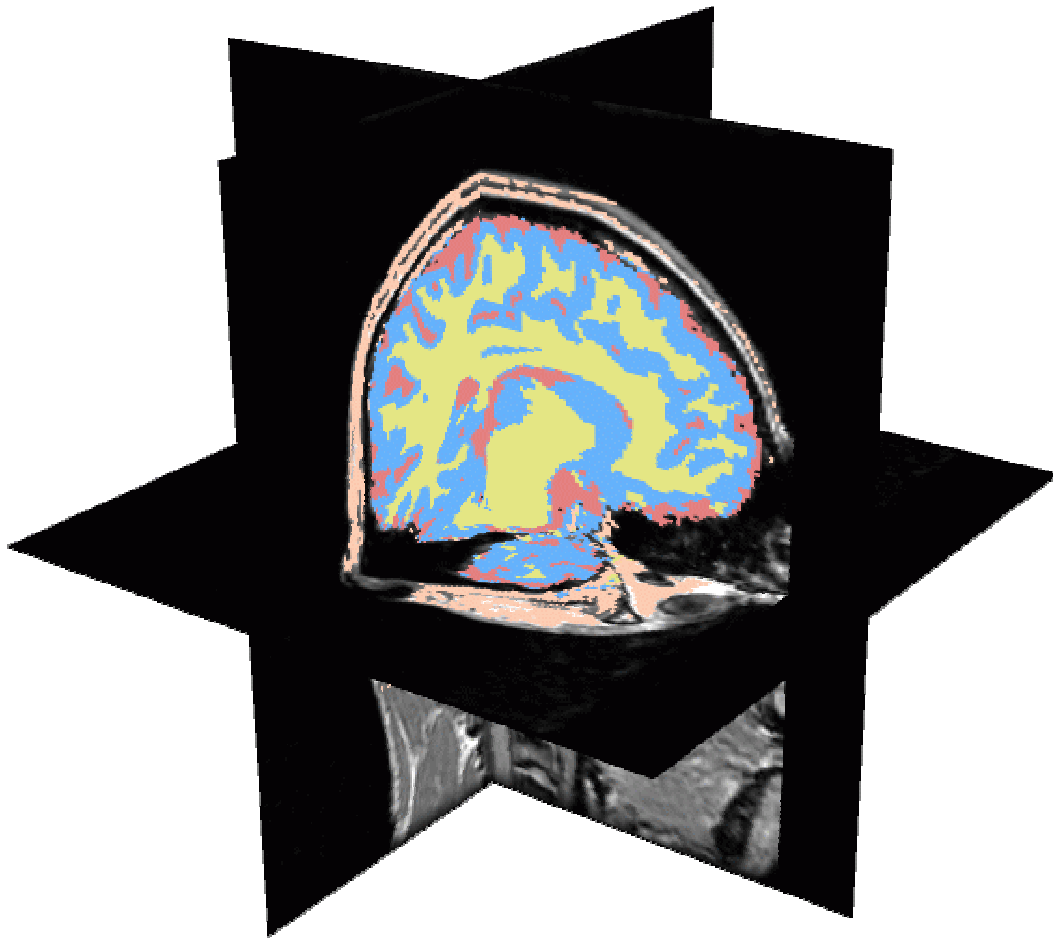


Sonia Pujol, Ph.D.

Surgical Planning Laboratory
Harvard Medical School



Goal of the course



Guiding you step by step through the process of using the Expectation-Maximization algorithm to **automatically segment brain structures from MRI data.**

Algorithm History

12-year of algorithm development

- 1996: Williams Wells et al.
EM framework for simultaneous estimation of bias field a label map. IEEE Transactions on Medical Imaging.
- 1999: Kapur et al.
Model noise via Markov Random Field . MIT PhD Thesis
- 2002: Van Leemput et al.
Non-spatial tissue priors. IEEE Transactions on Medical Imaging
- Since 2002: Pohl et al.
Deformable registration to align atlas (MICCAI)
Hierarchical framework to model anatomical dependencies (ISBI)
- 2007: Brad Davis et al.: EMSegmenter in Slicer3

Material

- **Slicer 3.4 Software**

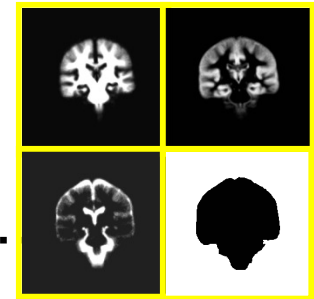
./Slicer.exe (Windows) or ./Slicer (Linux/Mac)

- **AutomaticSegmentation.zip** dataset

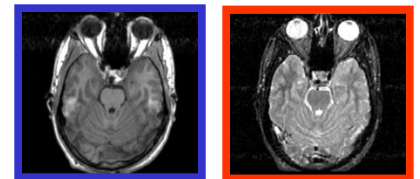
Disclaimer: It is the responsibility of the user of 3DSlicer to comply with both the terms of the license and with the applicable laws, regulations and rules.

Tutorial dataset

- Pre-computed **generic atlas** of the brain.....



- **T1** and **T2** volumes



Anatomical Tree

- The **anatomical tree** defines the hierarchy of structures that will be segmented.
- In this course, we focus on the following hierarchy
 - Intracranial Cavity
 - White Matter (WM)
 - Grey Matter (GM)
 - Cerebrospinal Fluid (CSF)
 - Background
 - Air
 - Skull

EM Pipeline

Step 1: Pre-processing



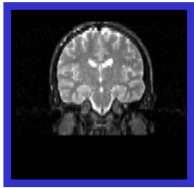
Step 2: Patient-specific atlas generation



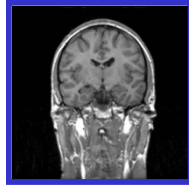
Step 3: Automatic segmentation

EM Pipeline: Preprocessing

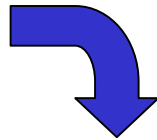
Patient data



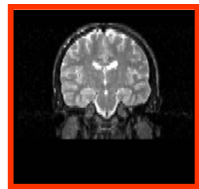
t2



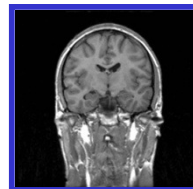
t1



Intensity Normalization
Normalize the intensity of t1
and t2

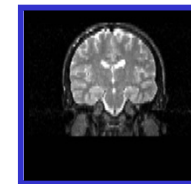
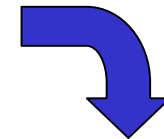


t2 normalized

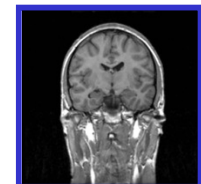


t1 normalized

Target to Target Registration
Align t2 to t1

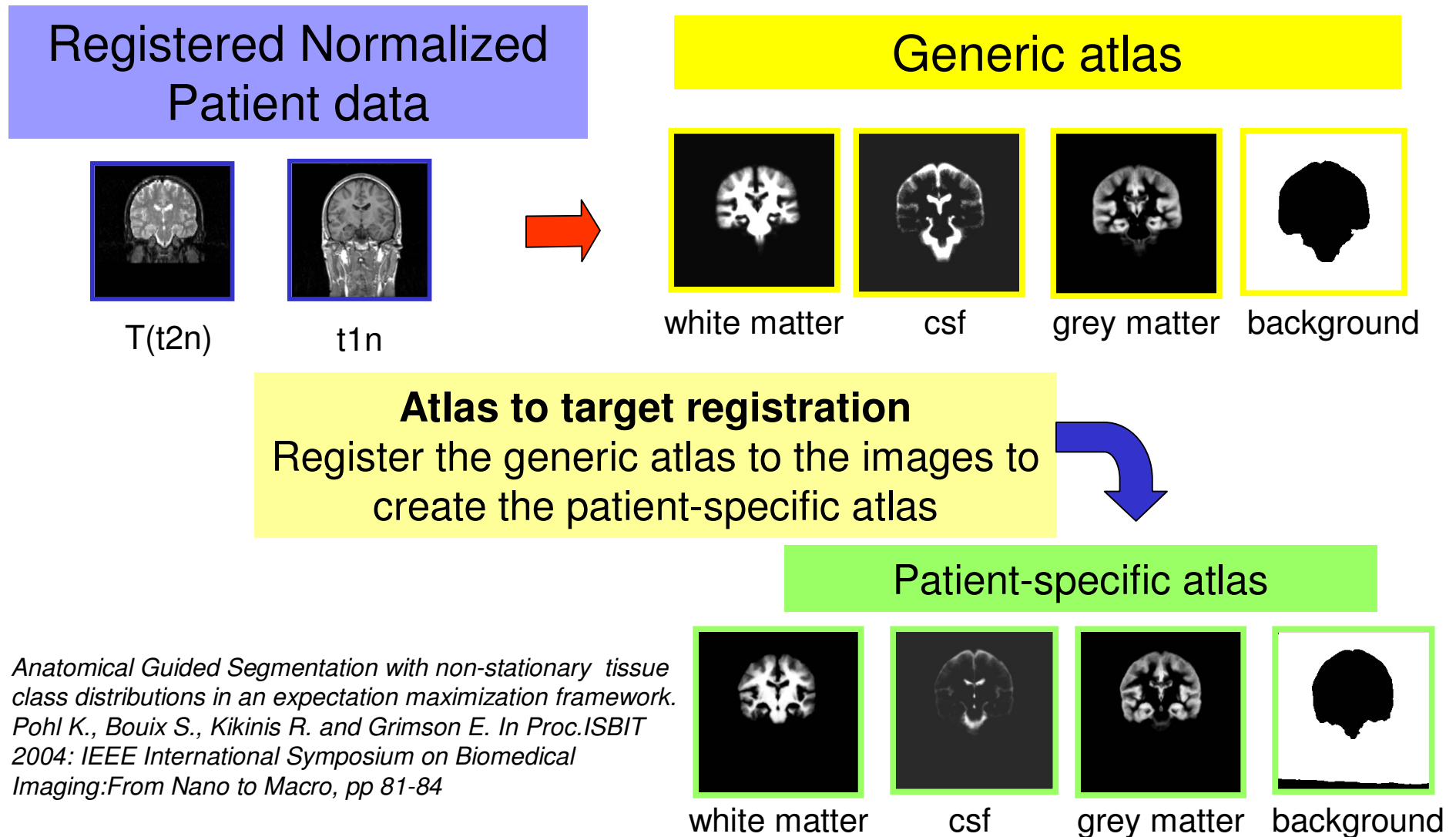


T(t2n)



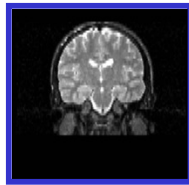
t1n

EM Pipeline: Patient-Specific Atlas Generation

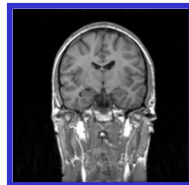


EM Pipeline: Segmentation

Normalized
Patient data



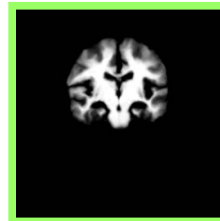
T(t2)
normalized



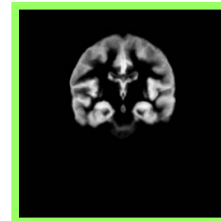
T1
normalized



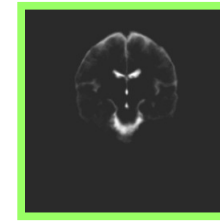
Patient-specific atlas



white matter



csf



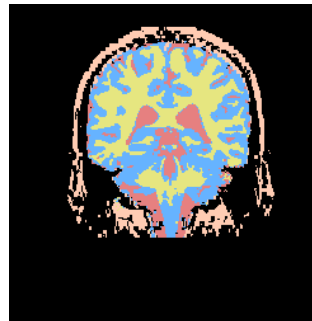
grey matter



background



Anatomical Guided Segmentation with non-stationary tissue class distributions in an expectation maximization framework. Pohl K., Bouix S., Kikinis R. and Grimson E. In Proc.ISBIT 2004: IEEE International Symposium on Biomedical Imaging: From Nano to Macro, pp 81-84

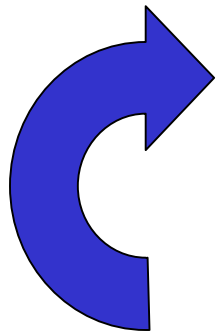


Segment using
the Expectation
Maximization
algorithm

EP Pipeline: Segmentation Algorithm

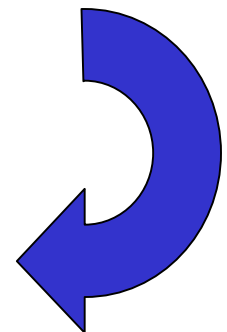
Expectation Step

classifies the MR voxels in tissue classes
(Gray Matter, White Matter, CSF)



Maximization Step

applies the intensity correction as a
function of the tissue class



Running Slicer3

Mac/Linux

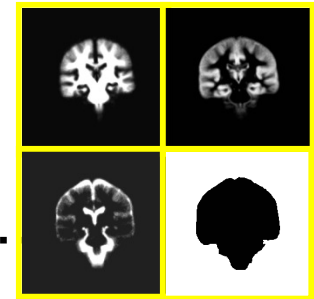
Run **./Slicer3** in Slicer3-build/

Windows

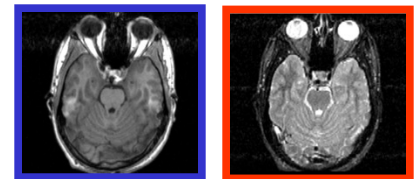
Run **./Slicer3.exe** in Slicer3-build/

Tutorial dataset

- Pre-computed **generic atlas** of the brain.....



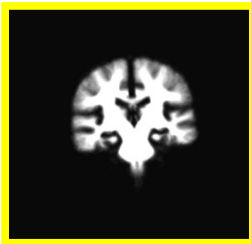
- **T1** and **T2** volumes



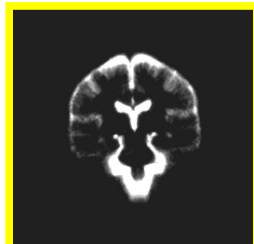
Generic Brain Atlas

The Generic **Brain Atlas** is composed of four grey-levels volumes which correspond to the structures that will be automatically segmented in the MRI example datasets.

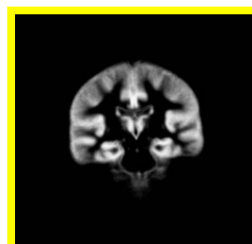
White Matter
(WM)



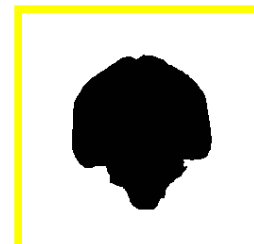
Cerebrospinal
Fluid (CSF)



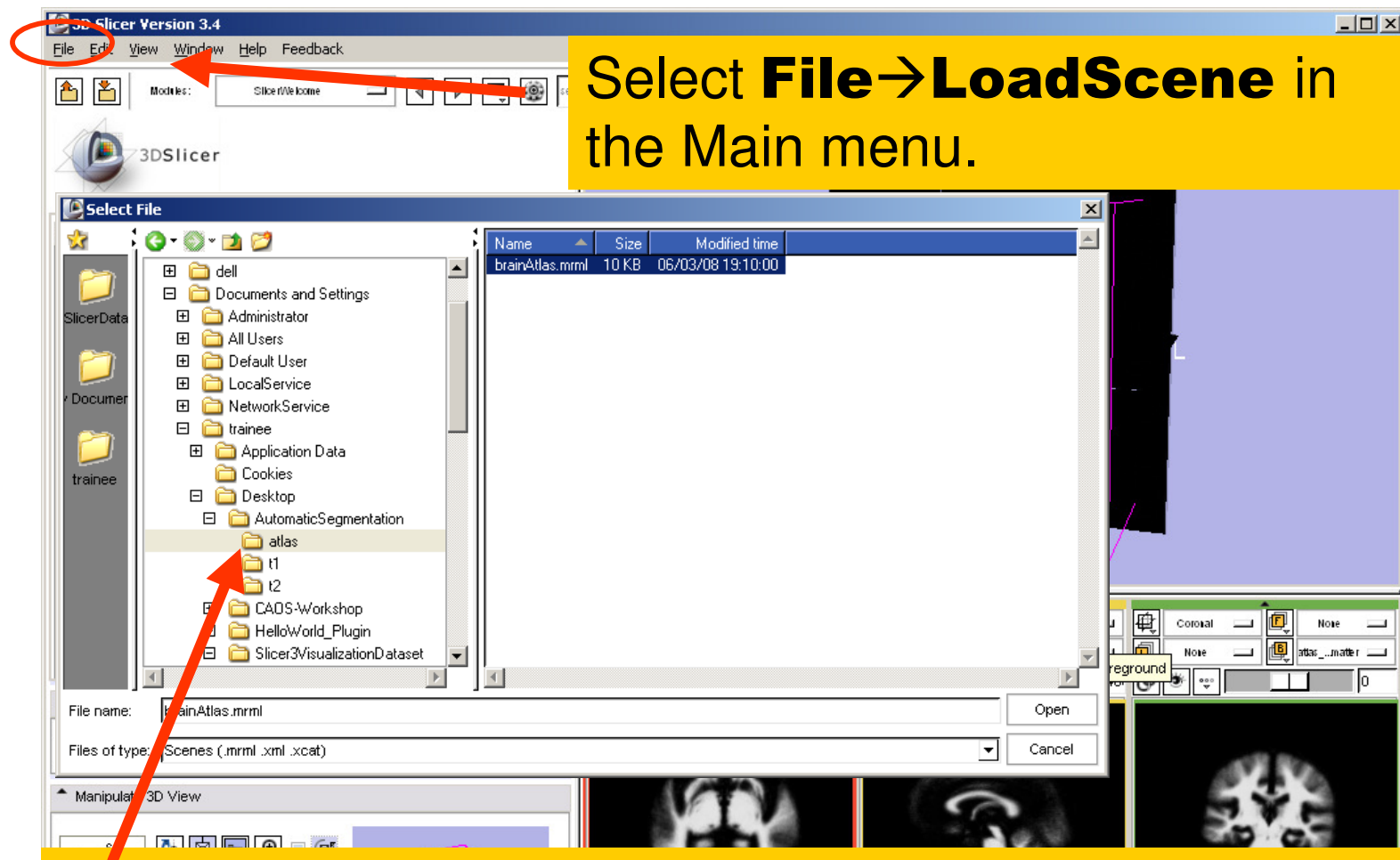
Grey Matter
(WM)



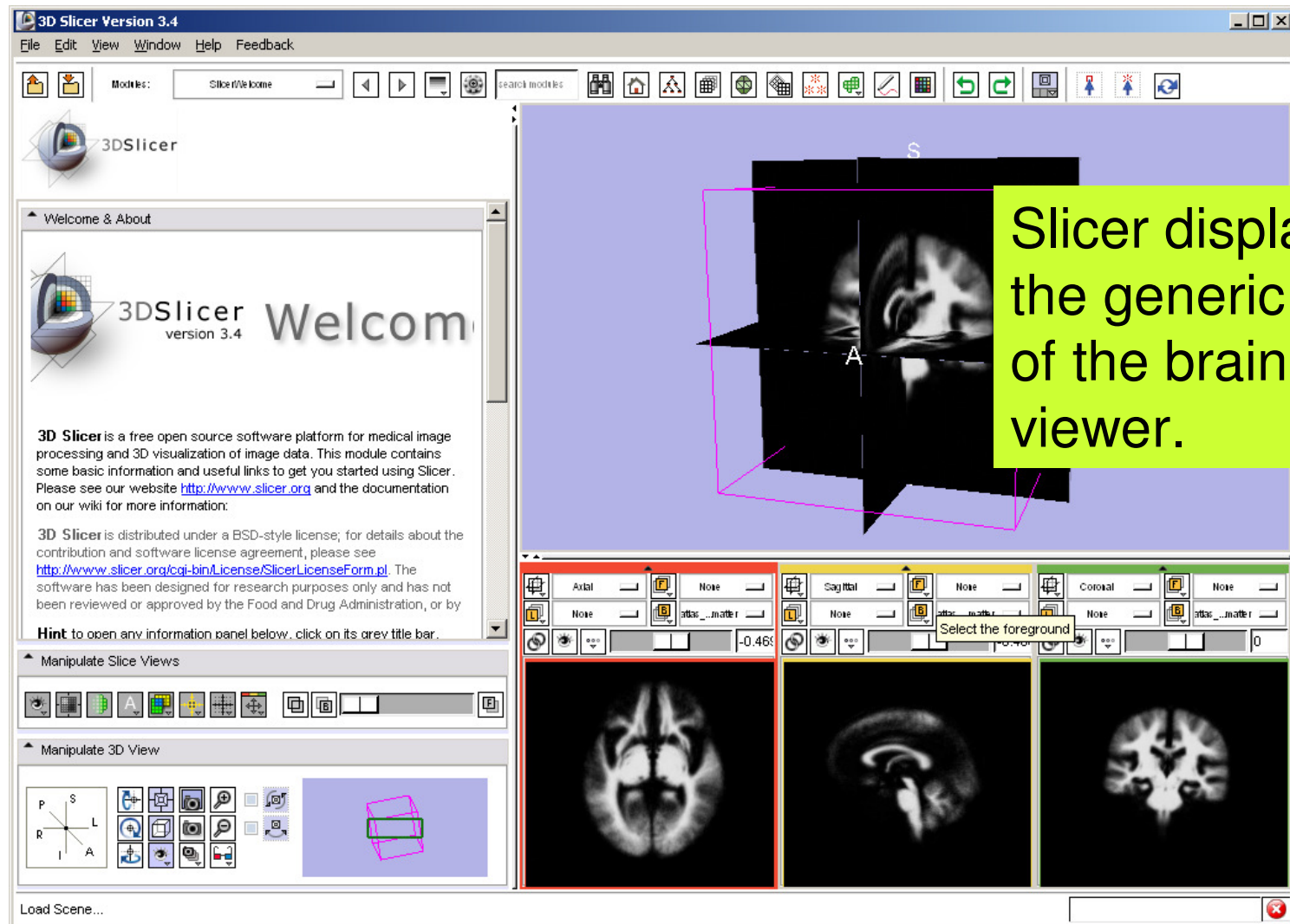
Background
(Bg)



Loading the generic atlas of the brain

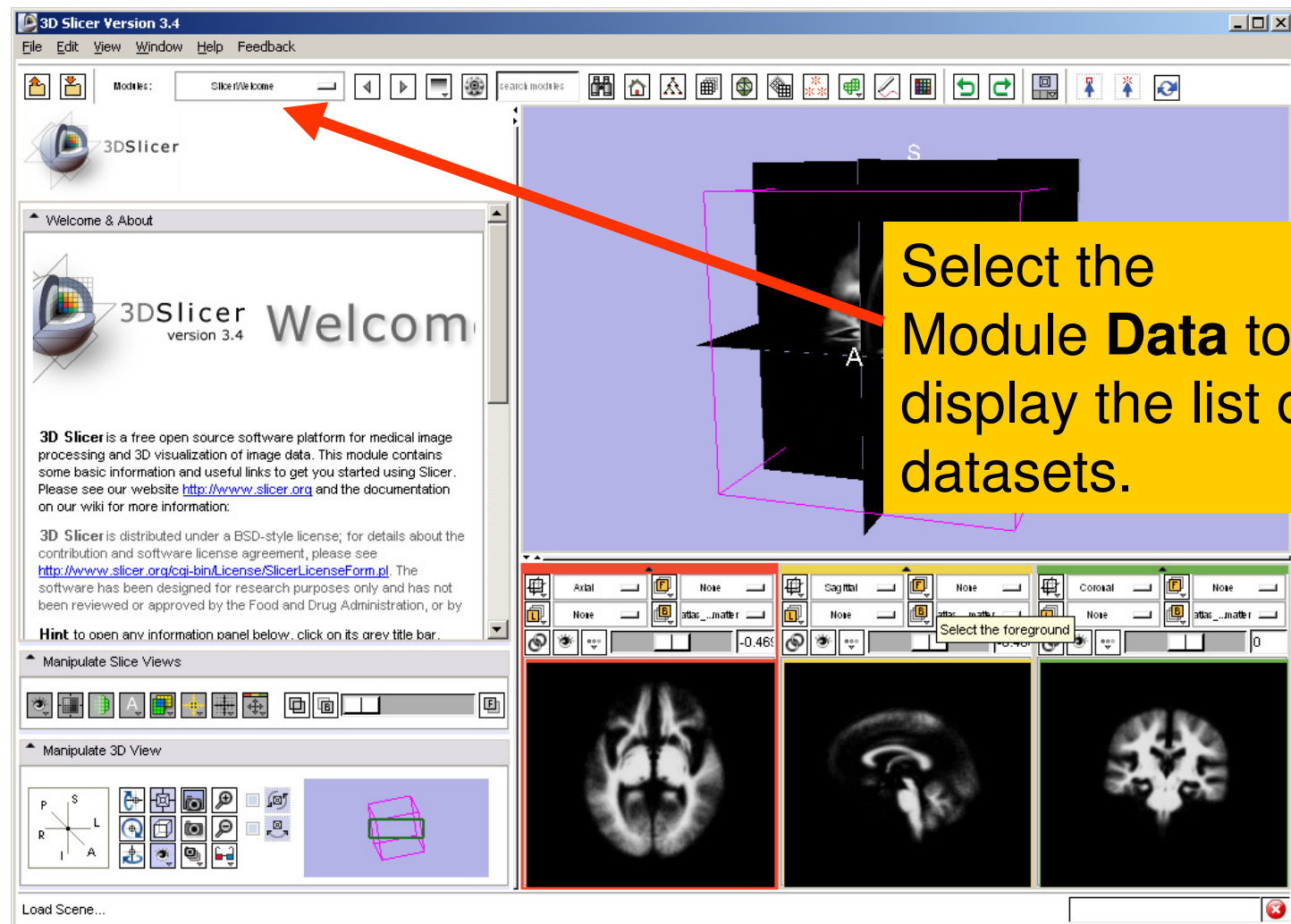


Viewing the generic atlas of the brain

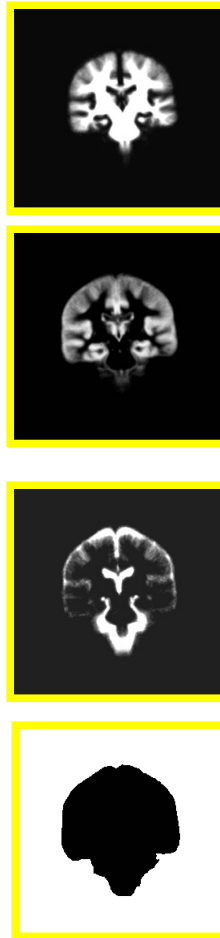
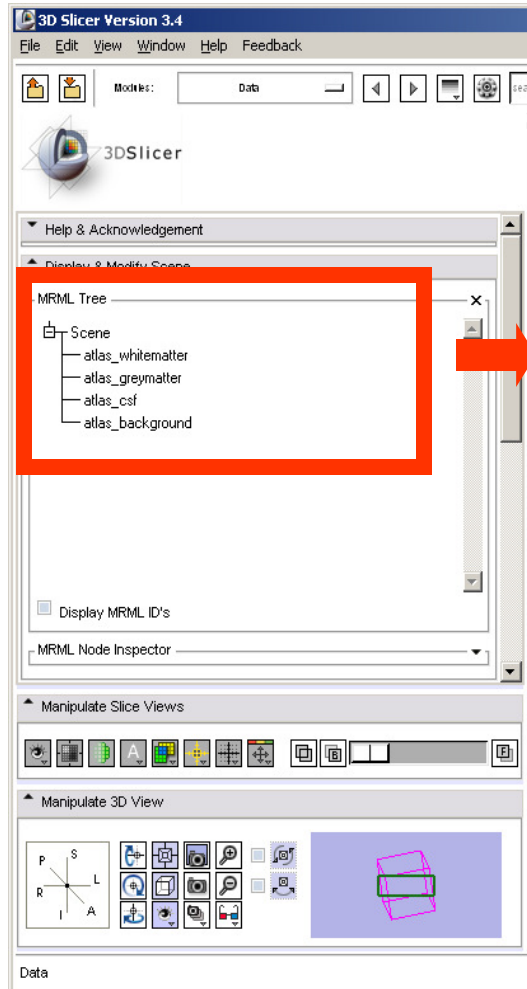


*Automatic Segmentation. Sonia Pujol, Ph.D., Harvard Medical School
National Alliance for Medical Image Computing*

Viewing the generic atlas of the brain



Viewing the generic atlas of the brain

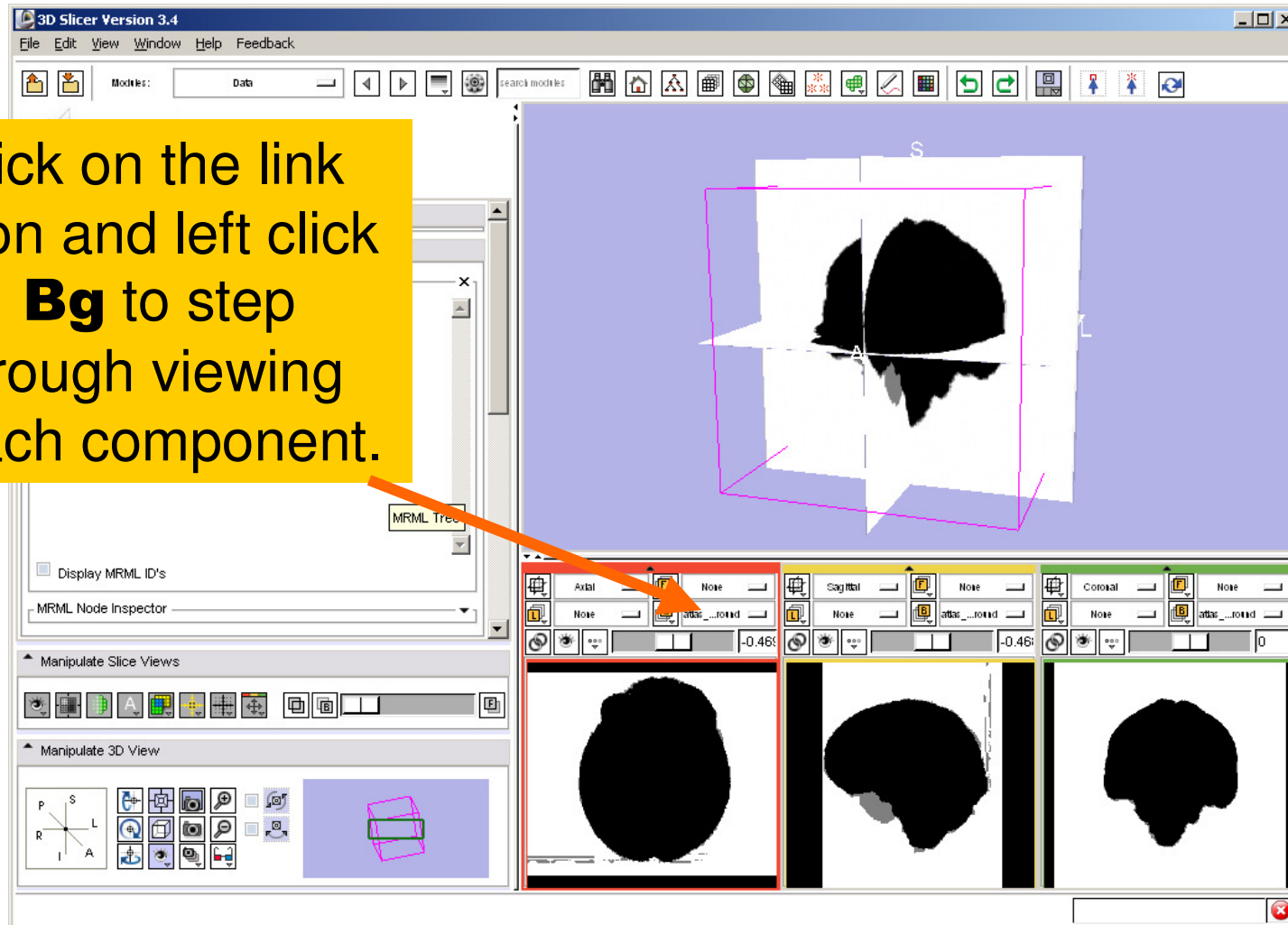


The generic atlas is composed of 4 volumes:

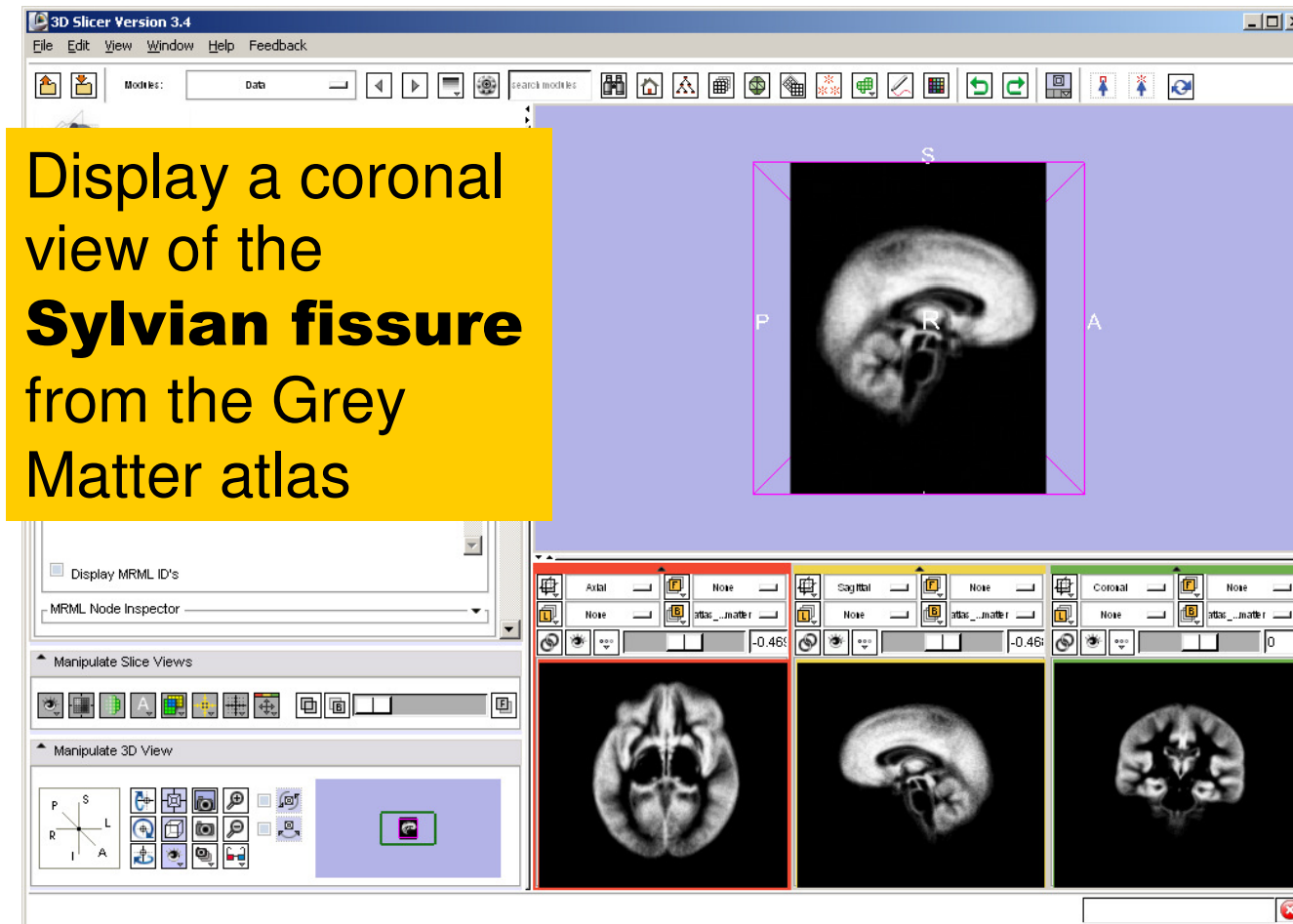
- White Matter
- Grey Matter
- CSF
- Background

Loading the generic atlas of the brain

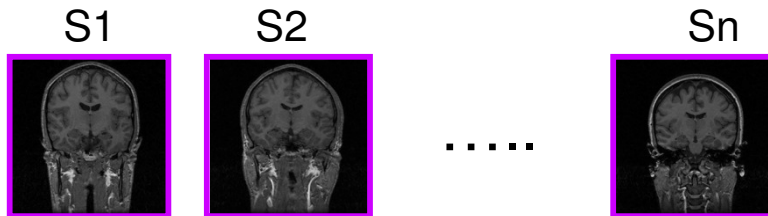
Click on the link icon and left click on **Bg** to step through viewing each component.



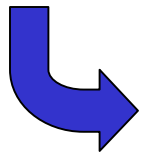
Viewing the generic atlas of the brain



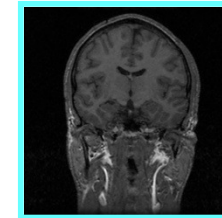
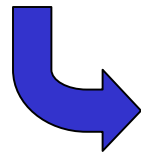
Generic Atlas Generation (Step 1)



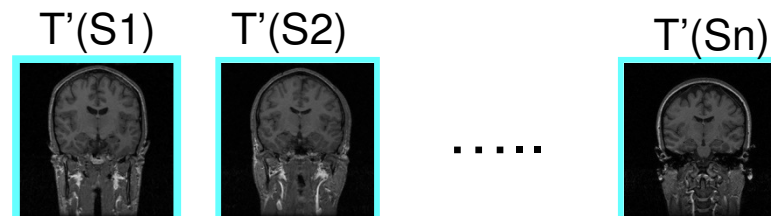
n=82 healthy subjects, ages 25-40



Register all the
subjects to the
training subject



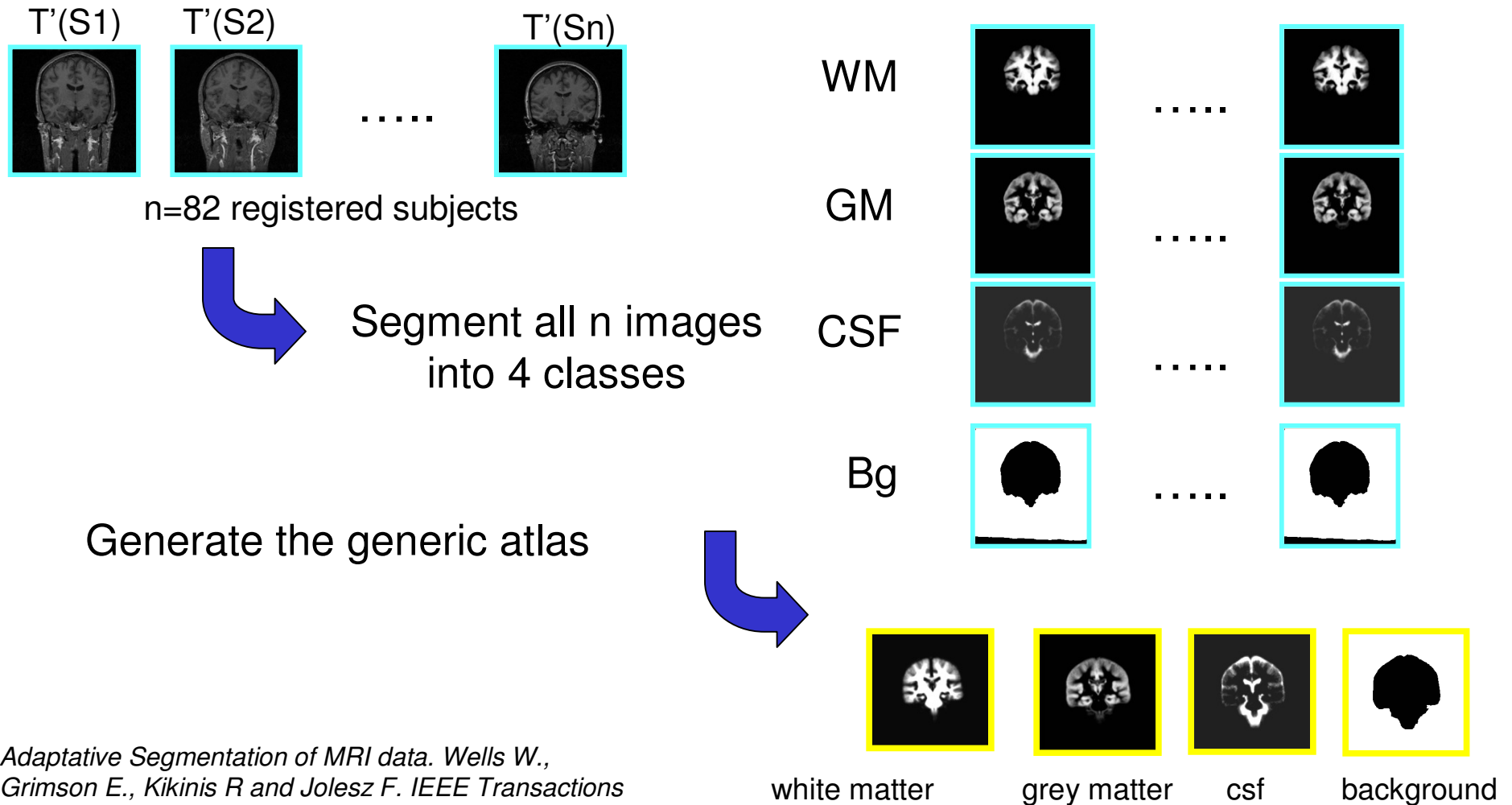
Training subject (randomly chosen)



n=82 registered subjects

A Binary Entropy Measure to Assess Non-rigid Registration Algorithms. S. Warfield, J. Rexilius, P. Huppi, T. Inder, E. Miller, W. Wells, G. Zientara, F. Jolesz, R. Kikinis. In Proc. MICCAI 2001: Medical Image Computing and Computer-Assisted Interventions, pp 266-274.

Generic Atlas Generation (Step 2)

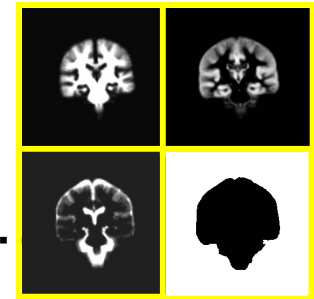


Adaptive Segmentation of MRI data. Wells W., Grimson E., Kikinis R and Jolesz F. IEEE Transactions on Medical Imaging, vol.15, p 429-442, 1996.

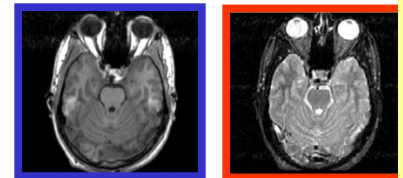
*Automatic Segmentation. Sonia Pujol, Ph.D., Harvard Medical School
National Alliance for Medical Image Computing*

Tutorial dataset

- Pre-computed **generic atlas** of the brain.....

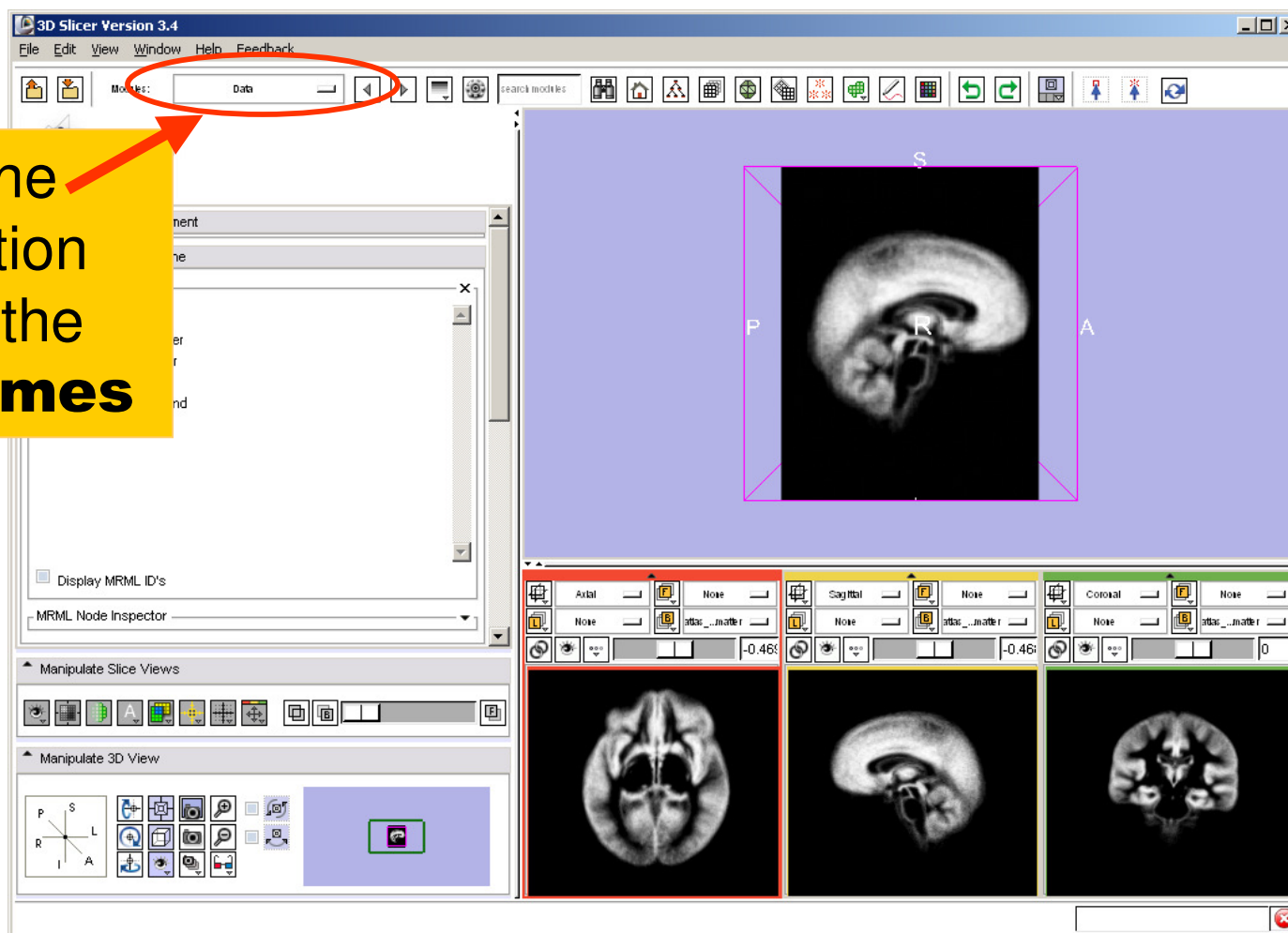


- **T1** and **T2** volumes



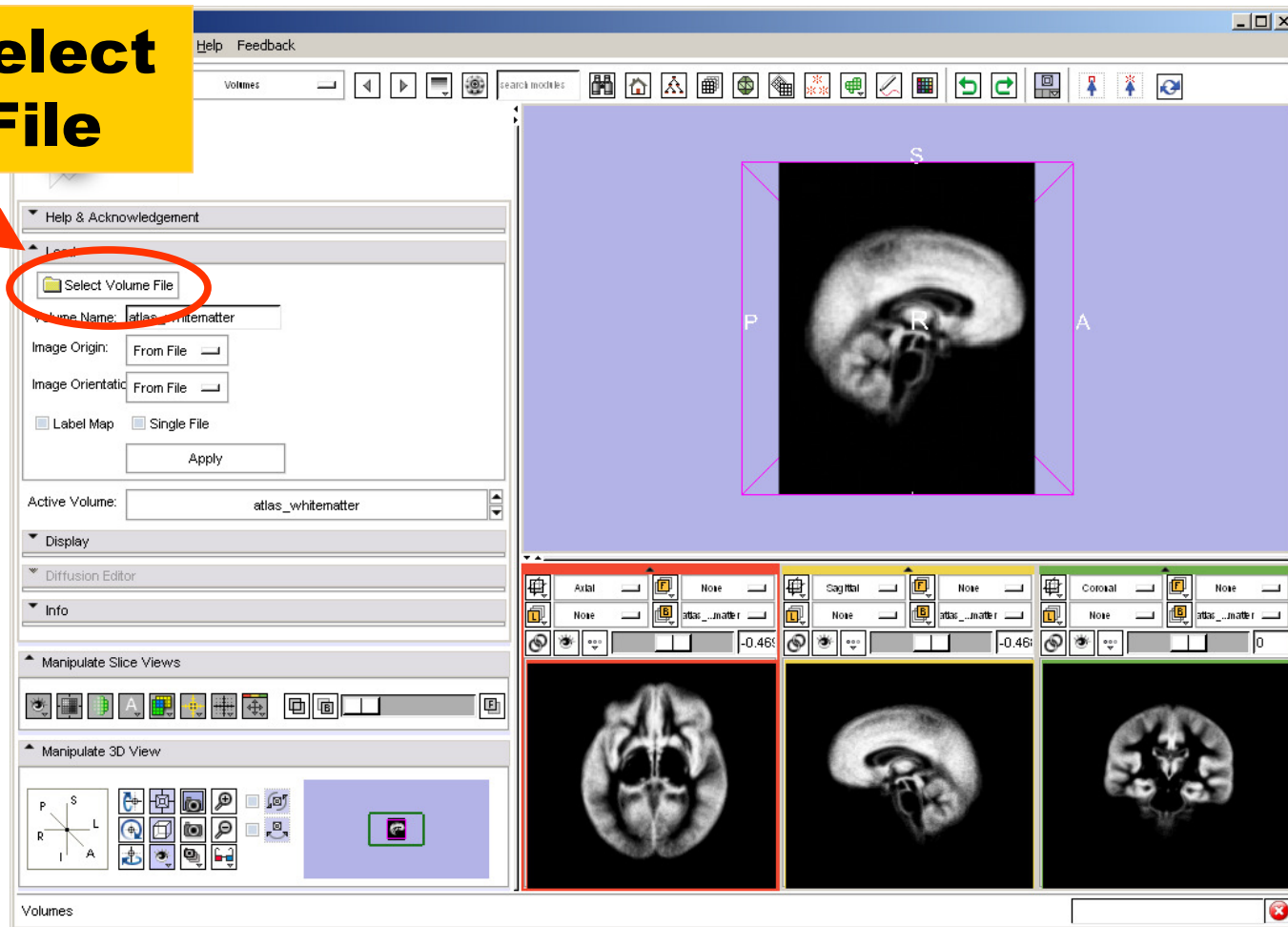
Loading T1 Volume

Left-click on the
module selection
menu to load the
module **Volumes**

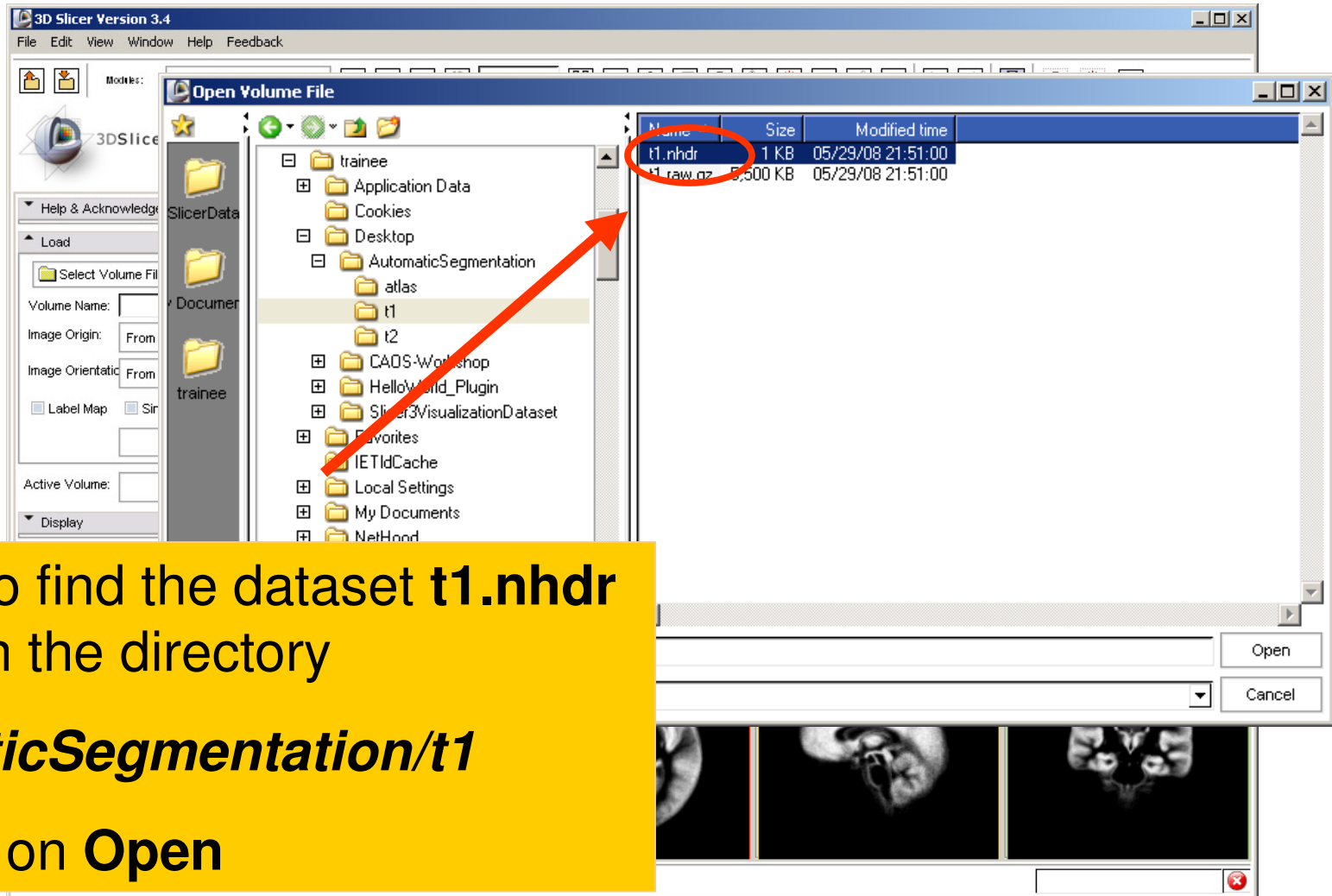


Loading T1 Volume

Click on **Select Volume File**



Loading T1 Volume

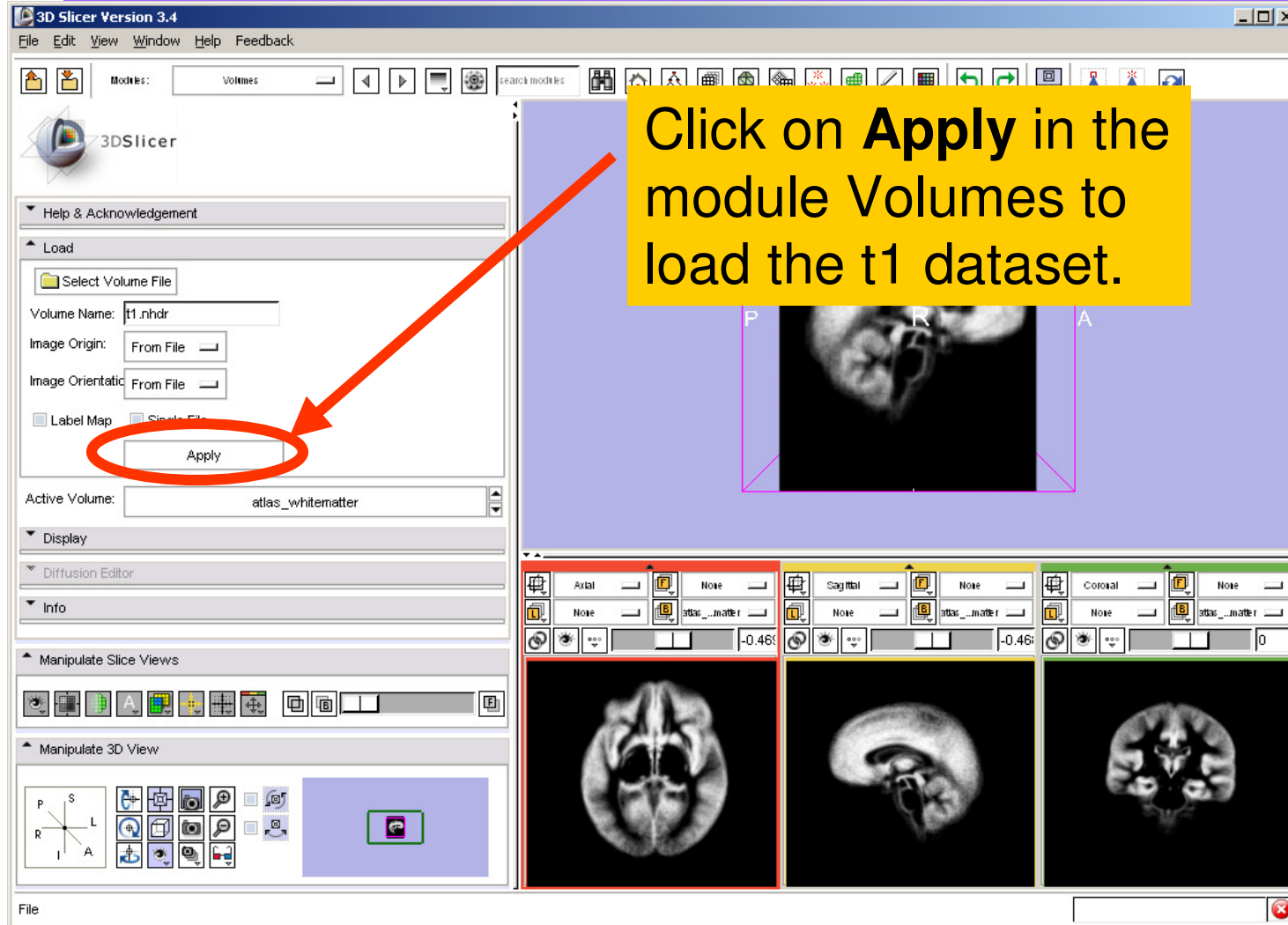


Browse to find the dataset **t1.nhdr** located in the directory

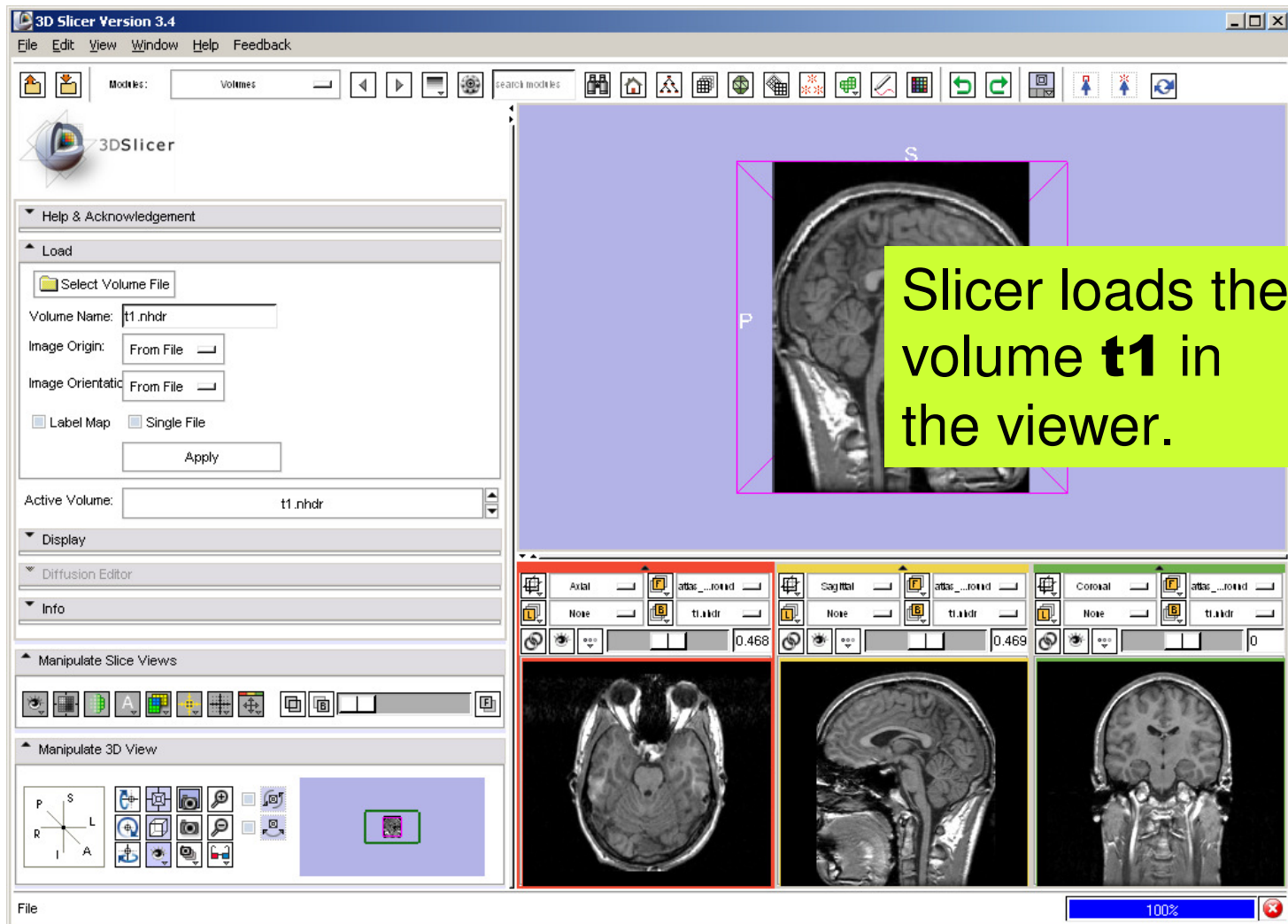
AutomaticSegmentation/t1

and click on **Open**

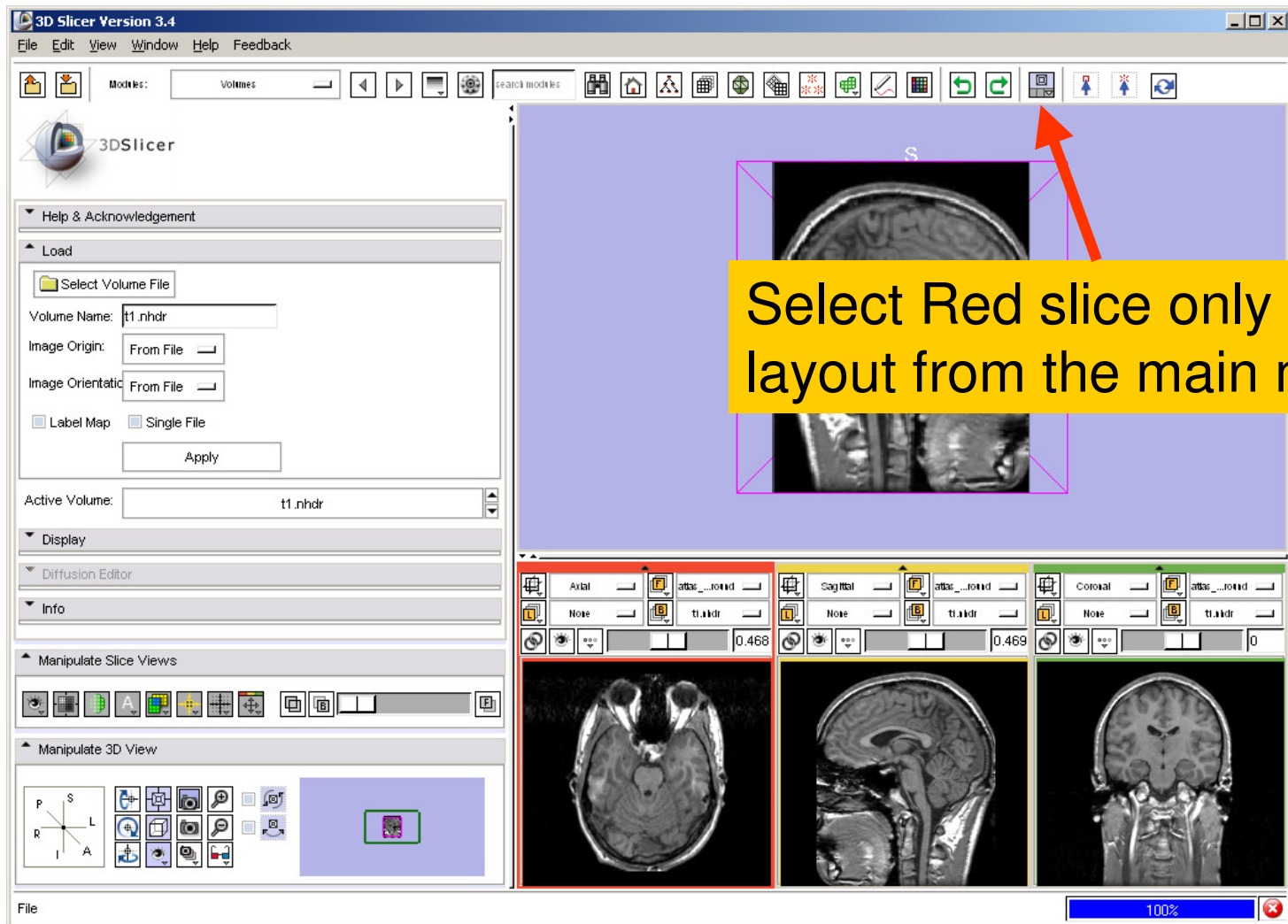
Loading T1 volume



Loading T1 volume

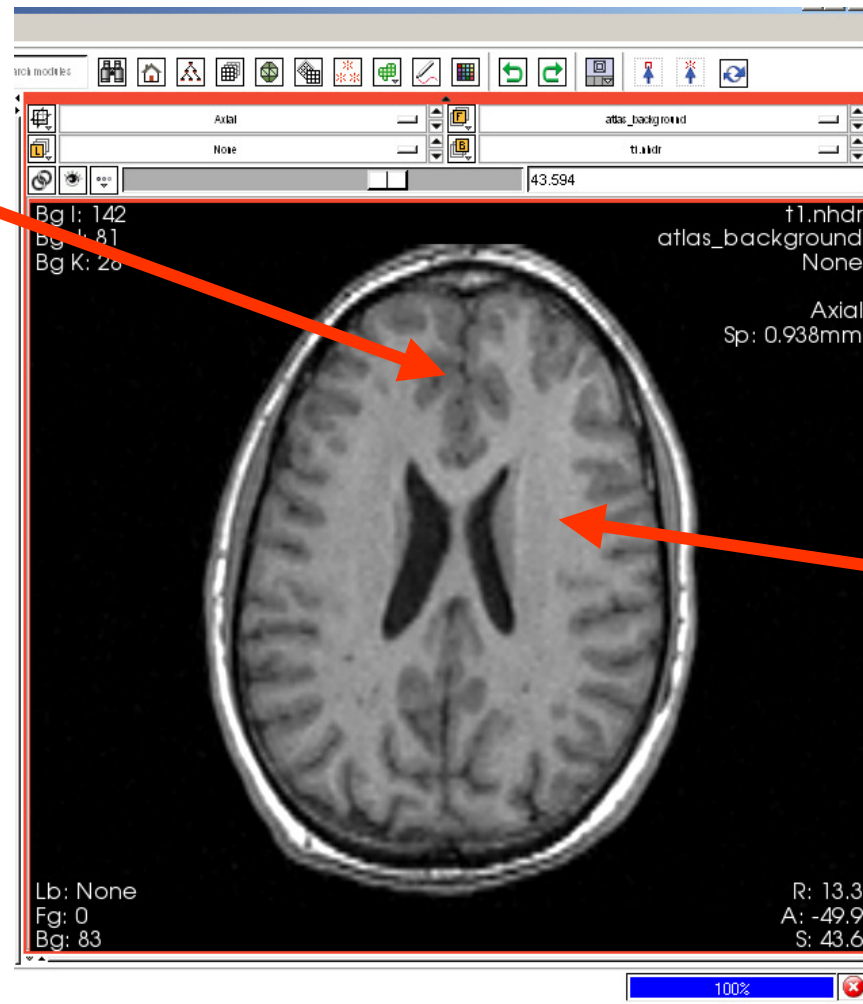


Loading T1 volume



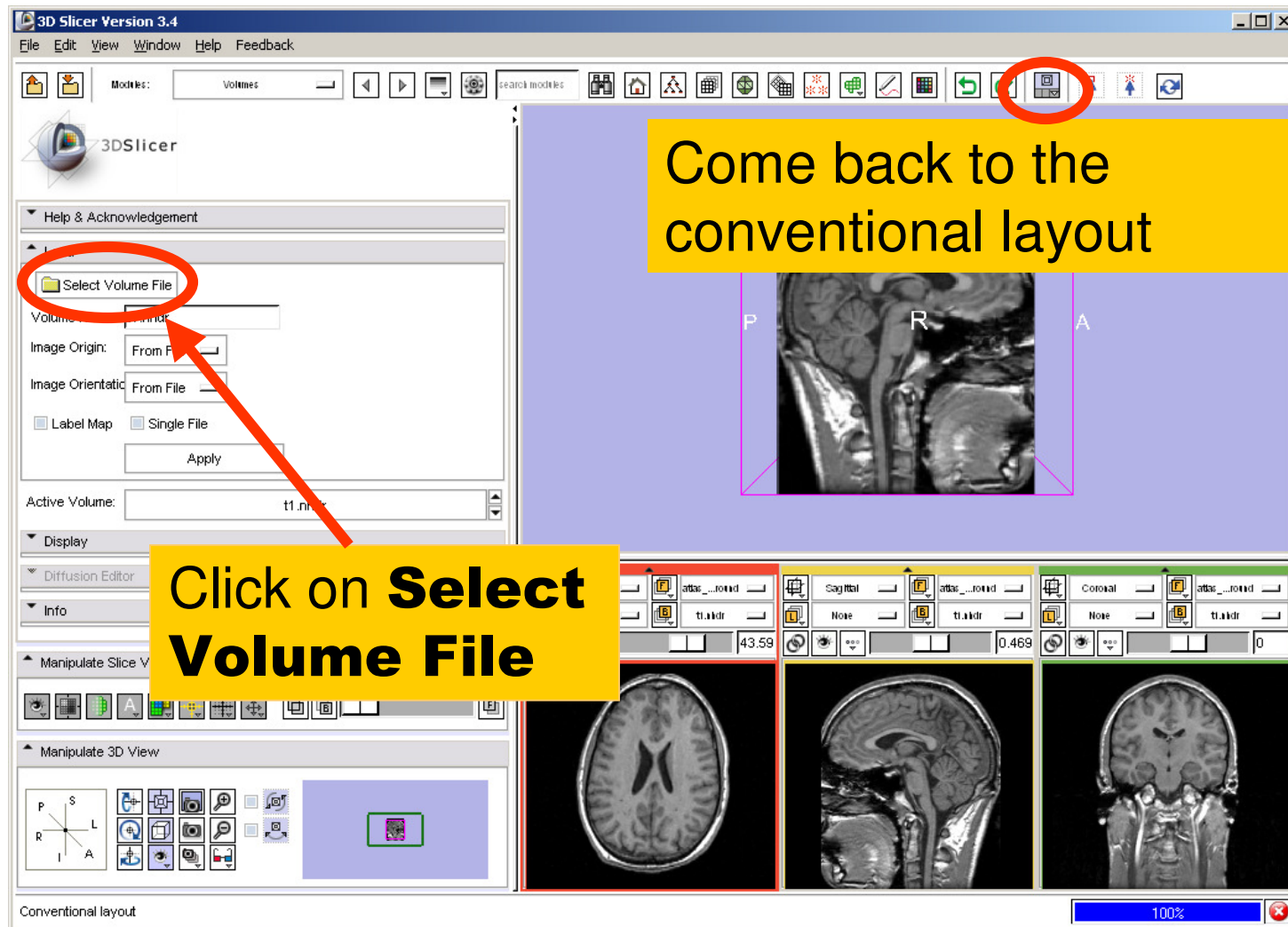
T1 contrast

Grey
Matter

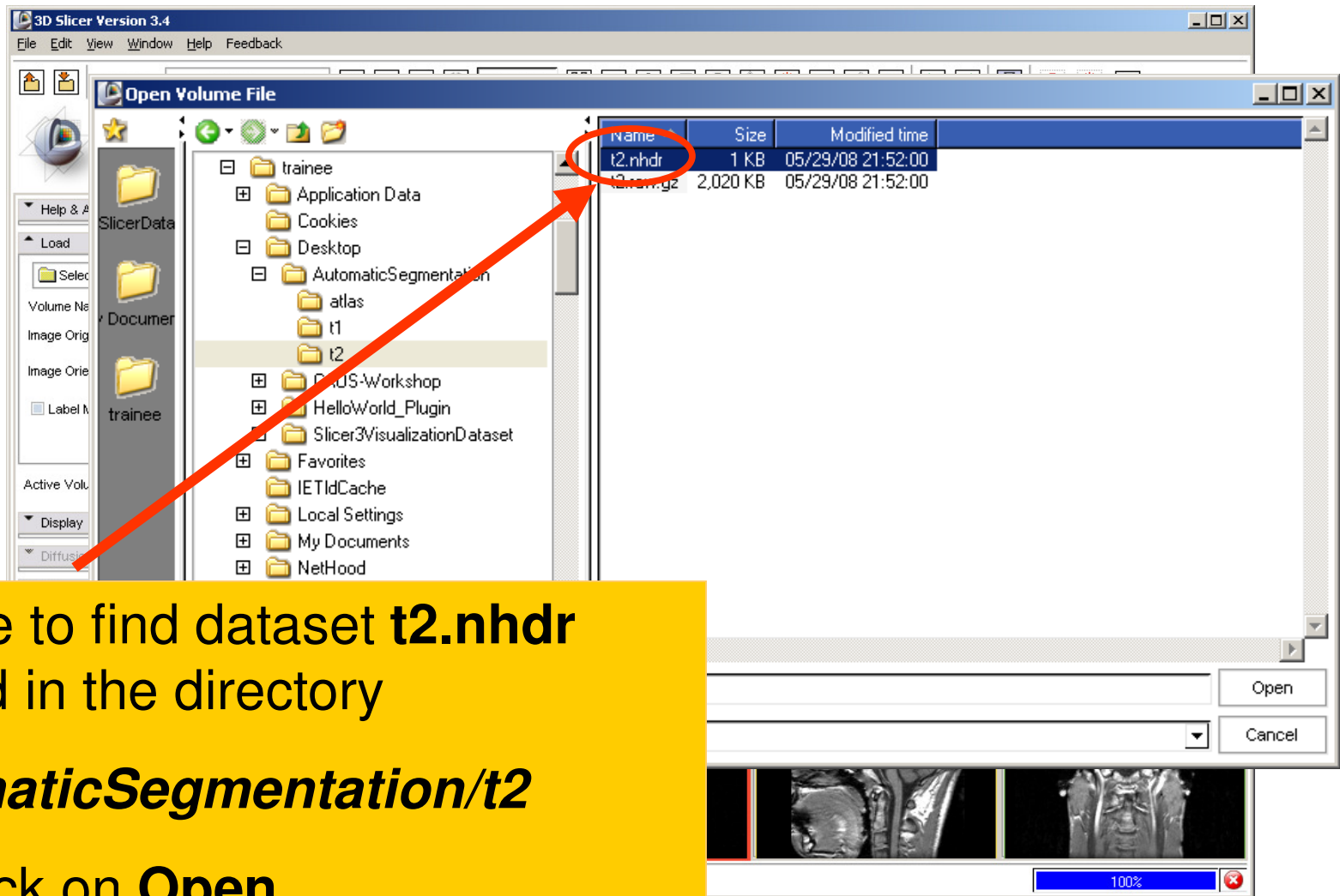


White
Matter

Loading T2 volume



Loading T2 volume



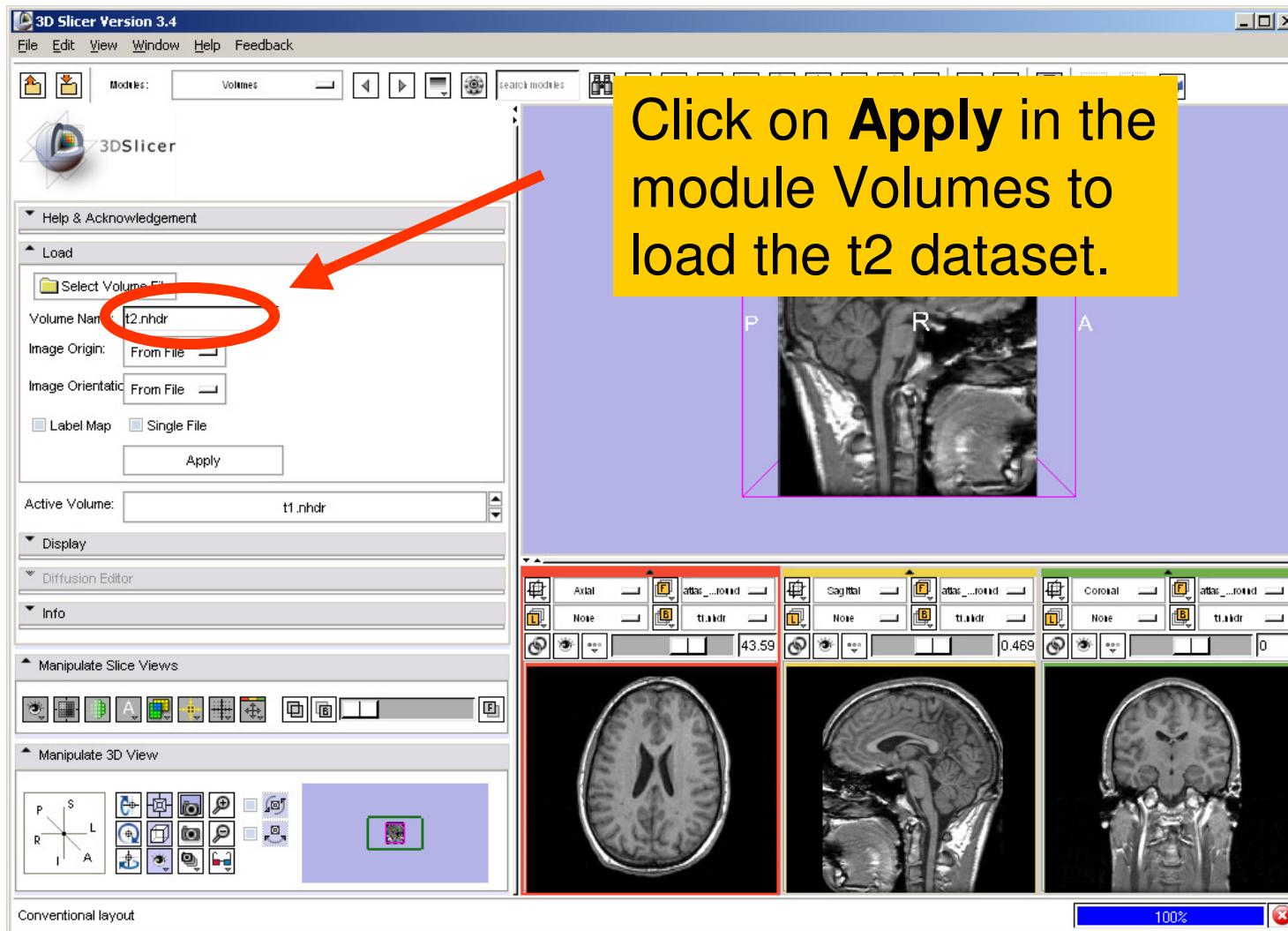
Browse to find dataset **t2.nhdr**
located in the directory

AutomaticSegmentation/t2

and click on **Open**

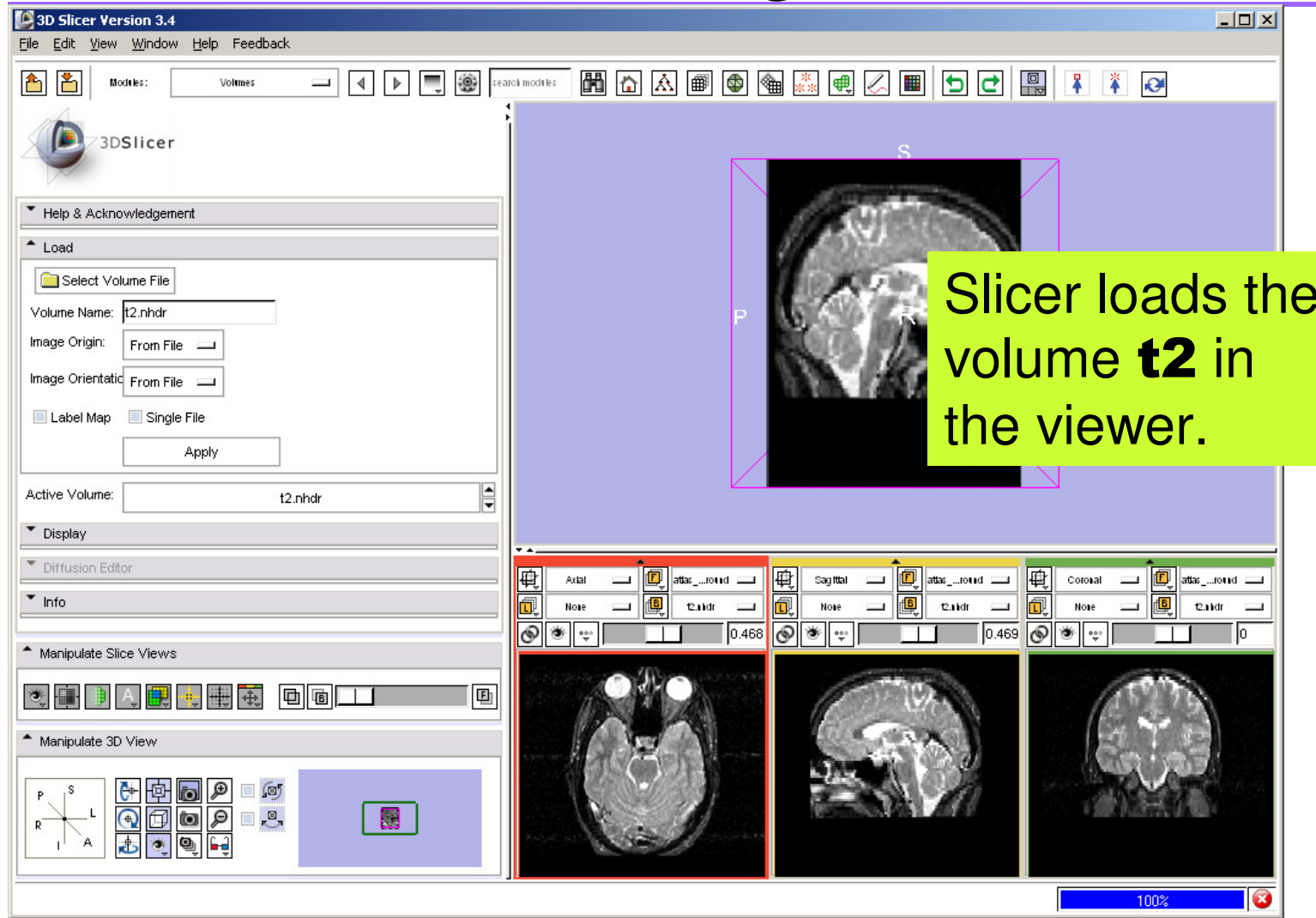
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National Alliance for Medical Image Computing*

Loading T2 volume

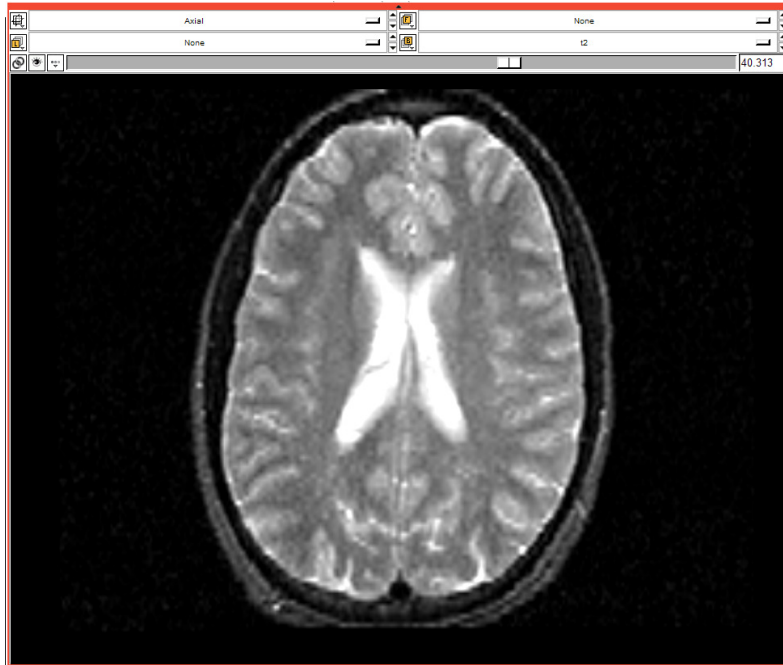


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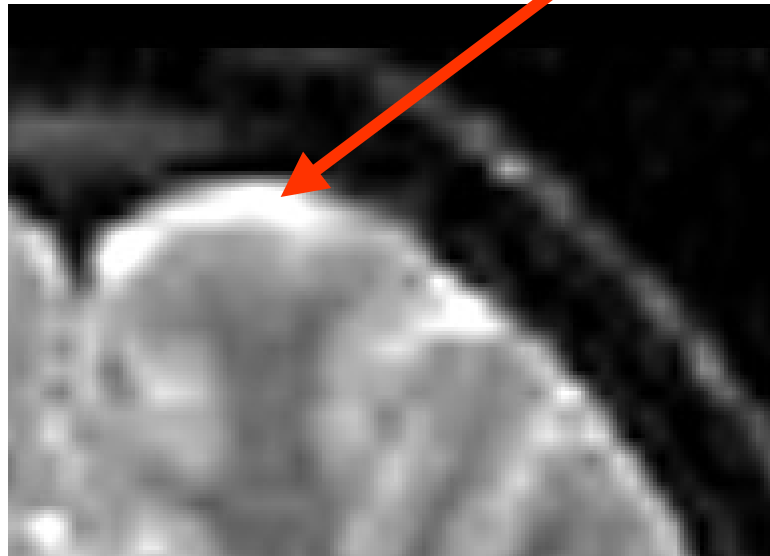
Loading T2 volume



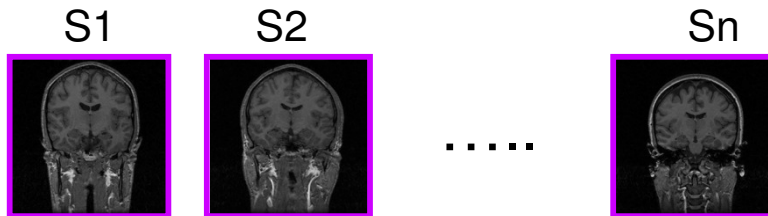
T2 contrast



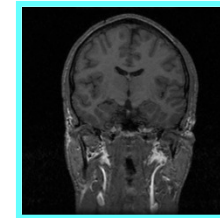
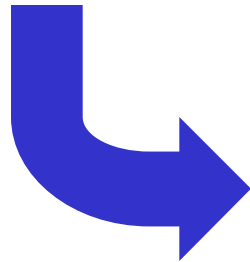
CSF



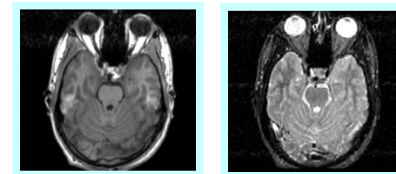
Generic Atlas



n=82 healthy subjects, ages 25-40

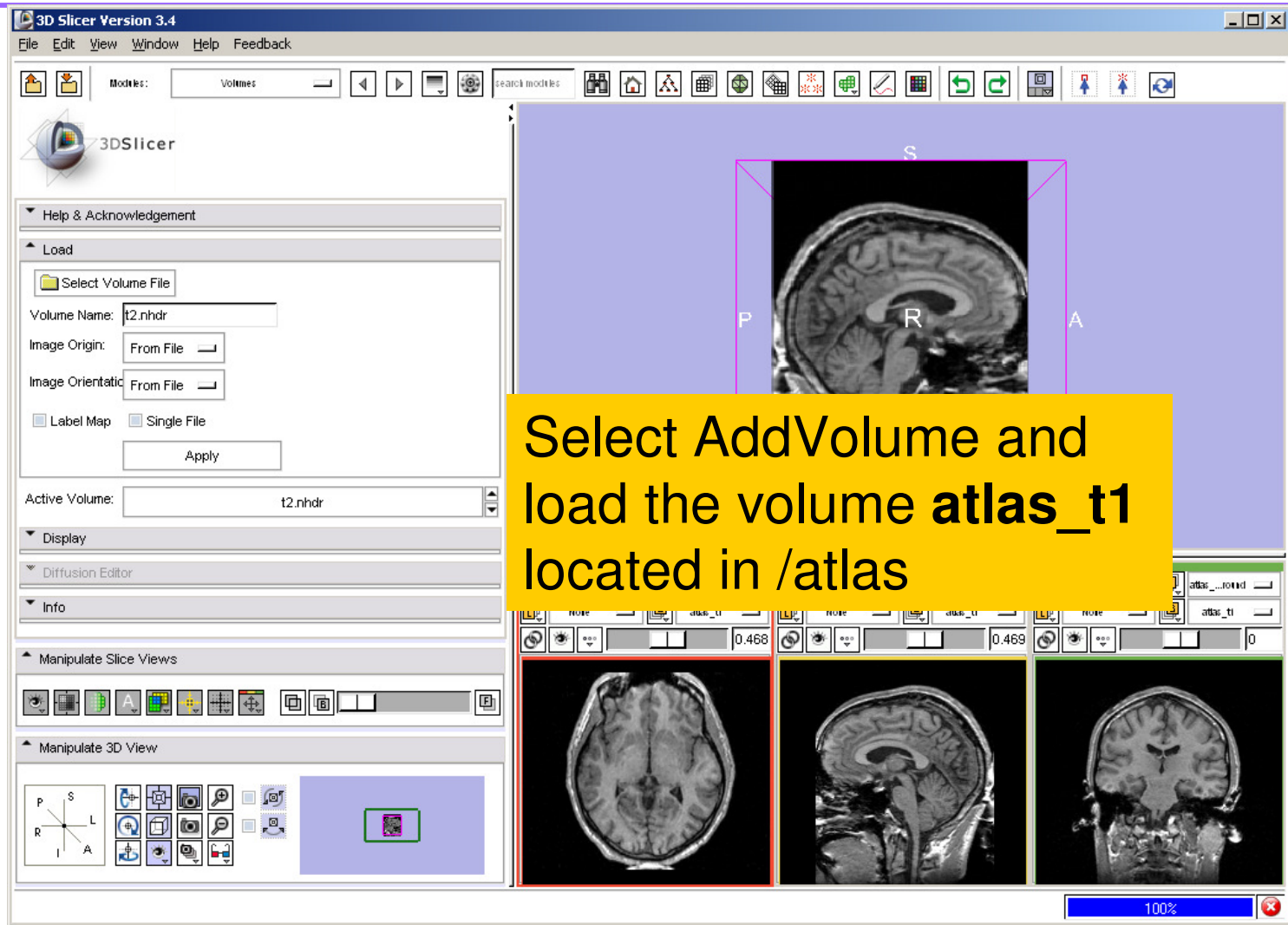


Training subject (randomly chosen)

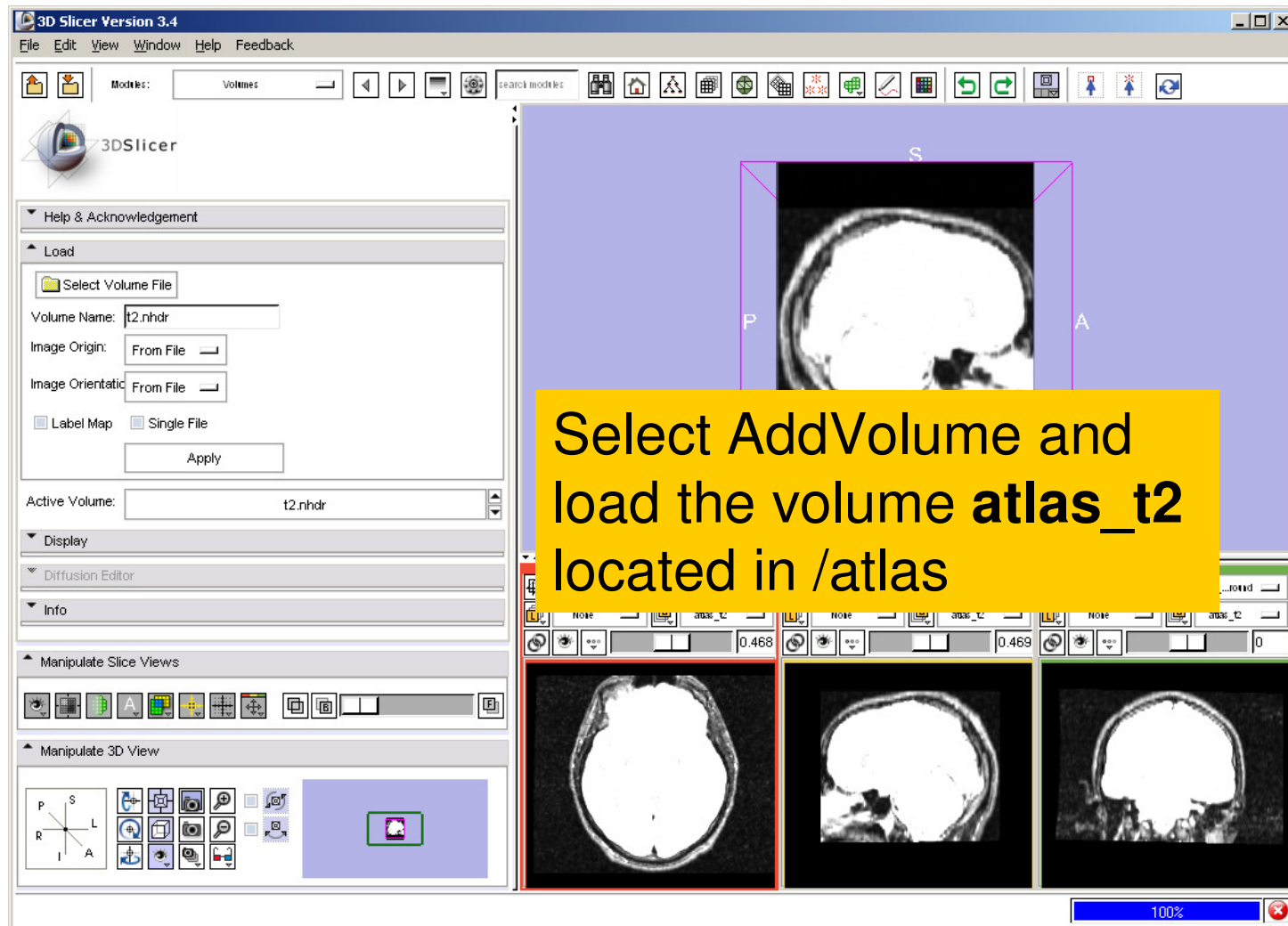


t1 and t2 of the training subject

Atlas T1

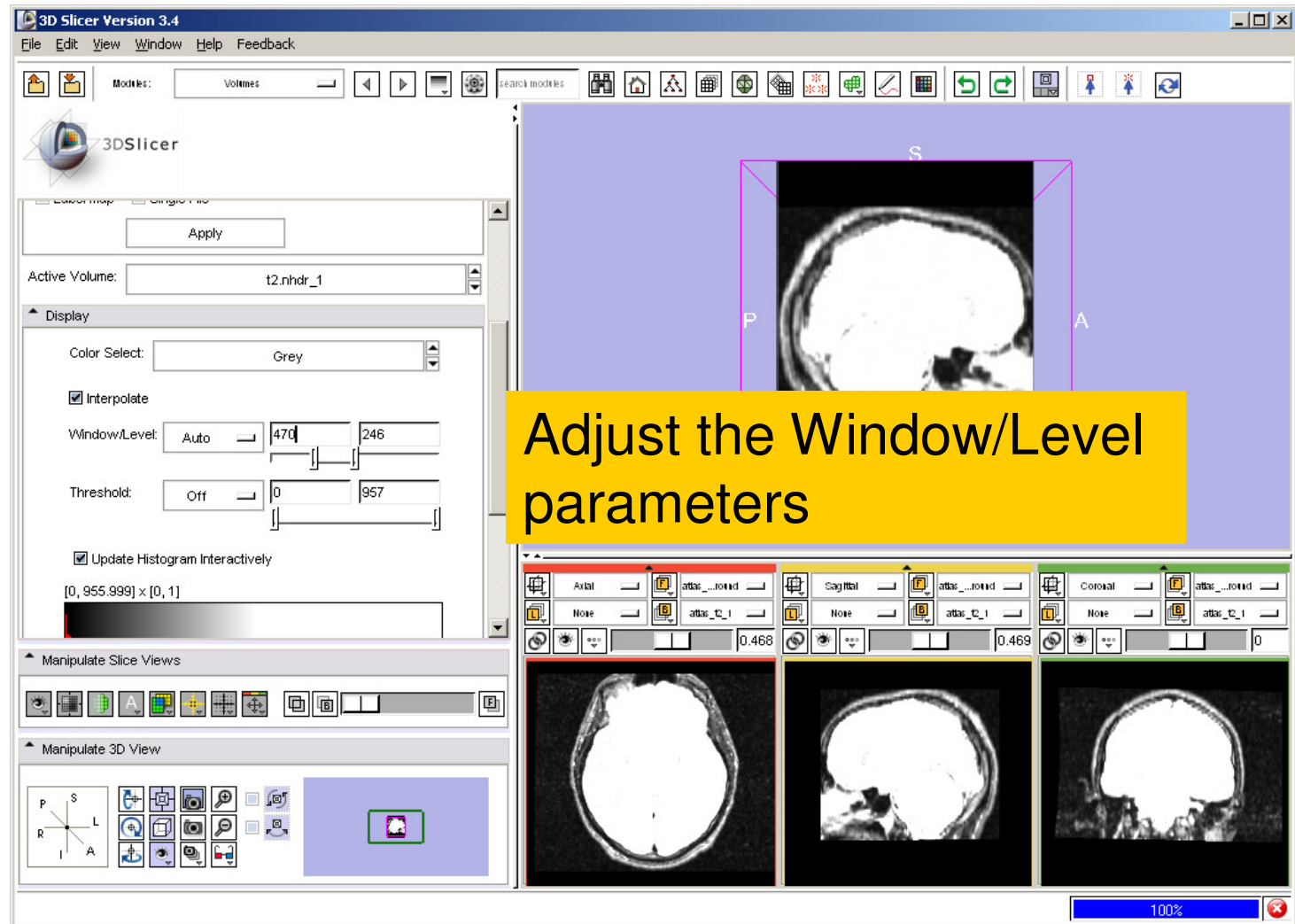


Atlas T2



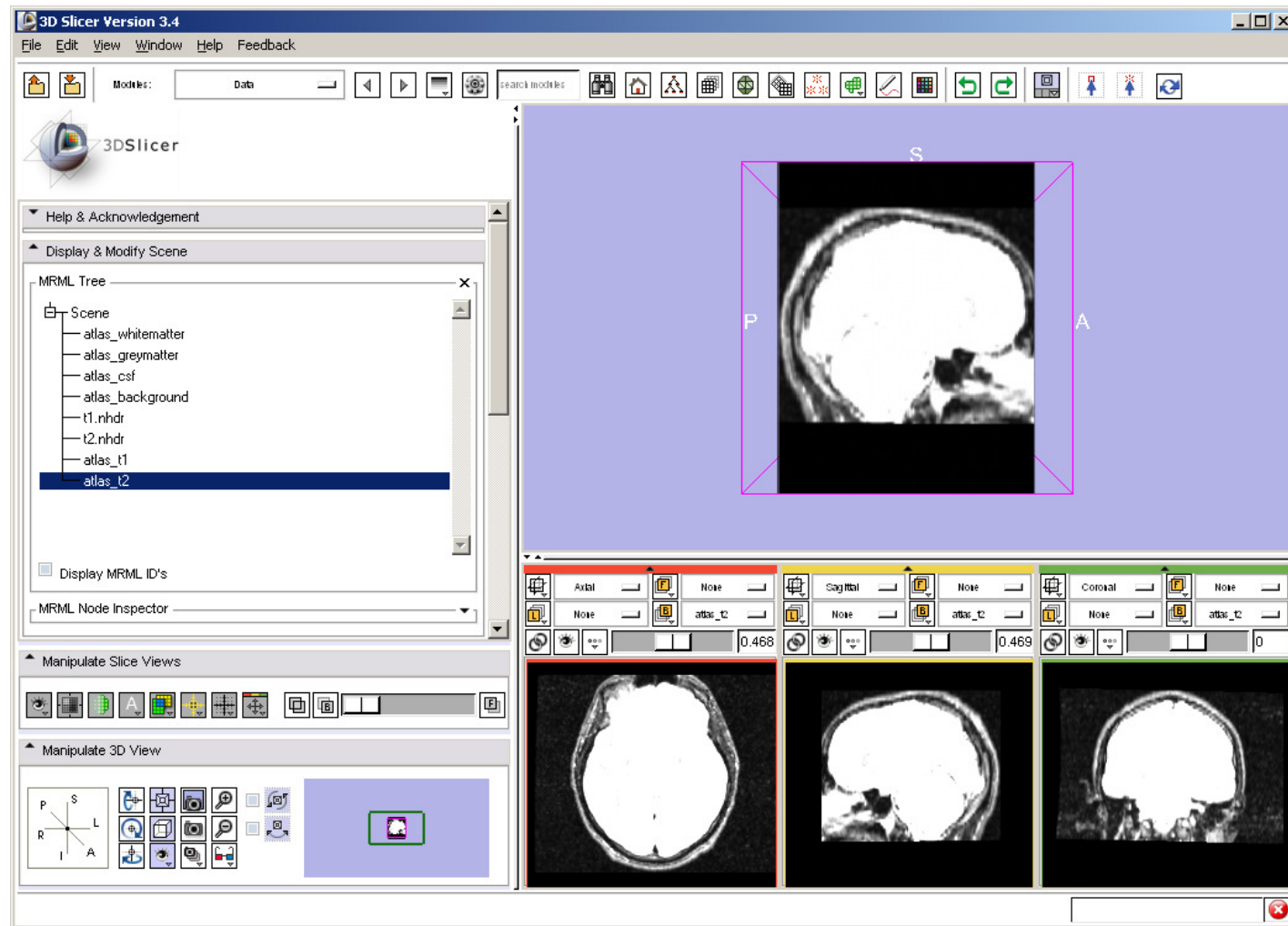
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Atlas T2



*Automatic Segmentation. Sonia Pujol, Ph.D., Harvard Medical School
National Alliance for Medical Image Computing*

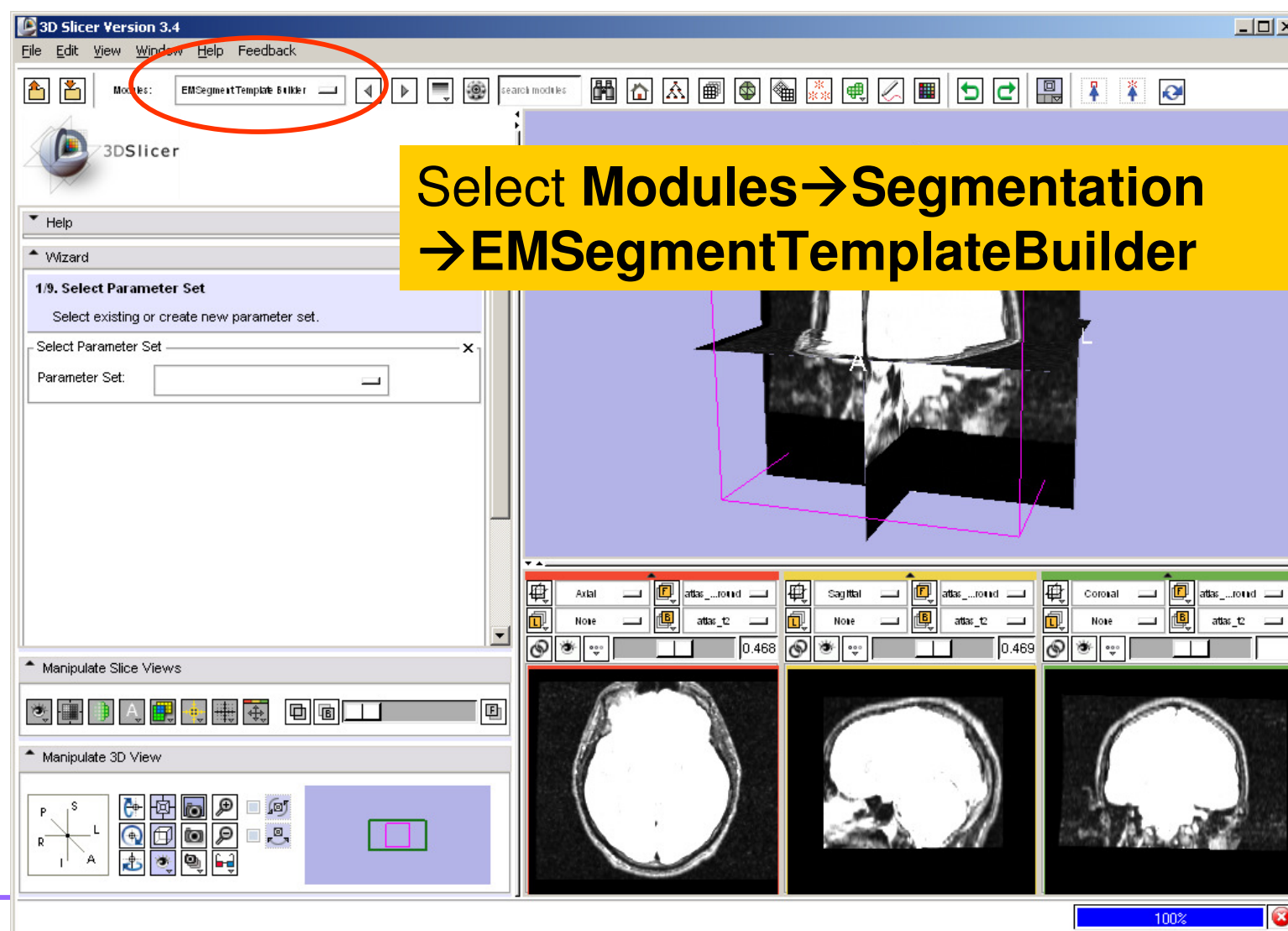
Training Datasets



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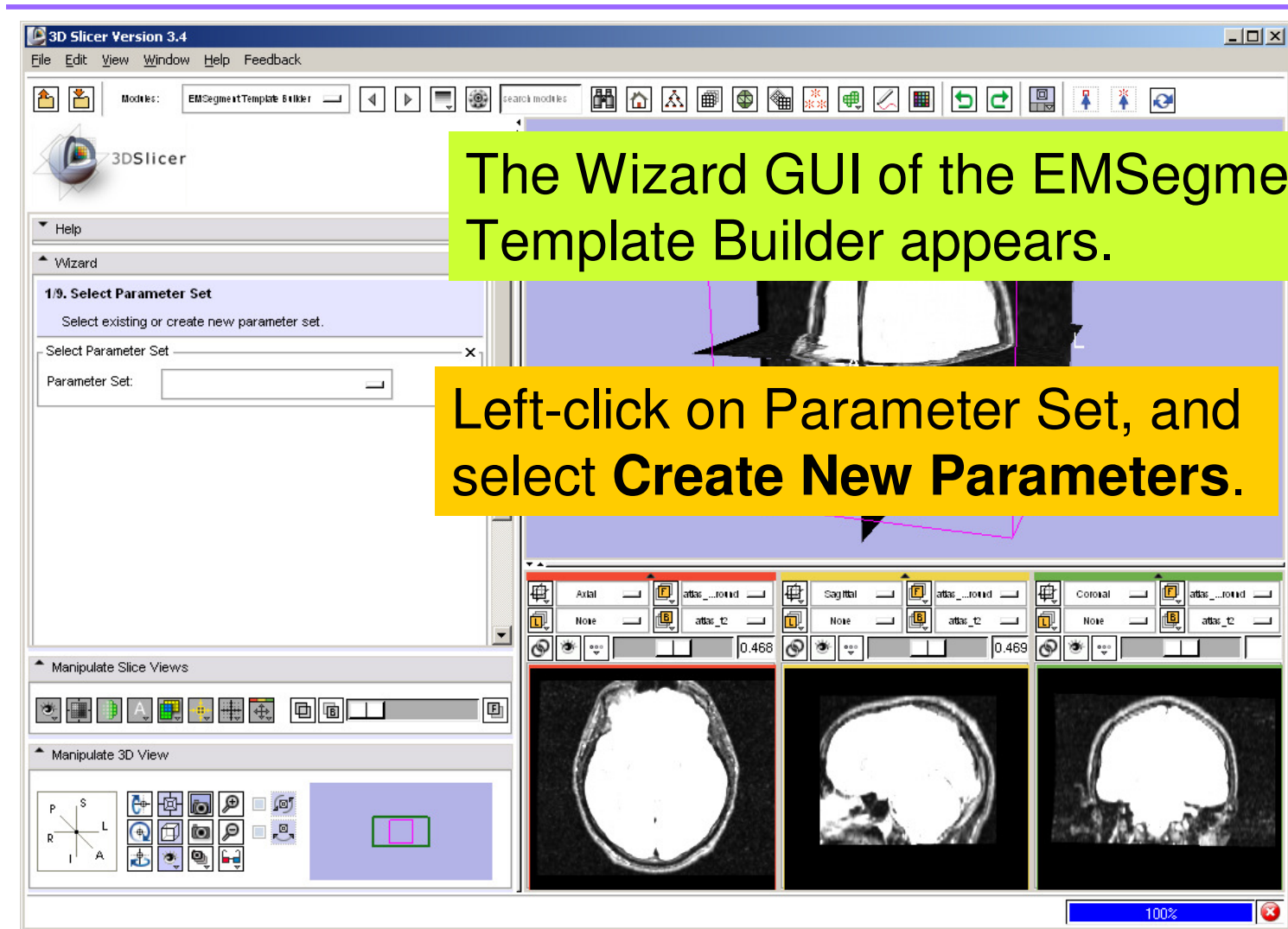
Template Builder: Parameters Settings

Template Builder

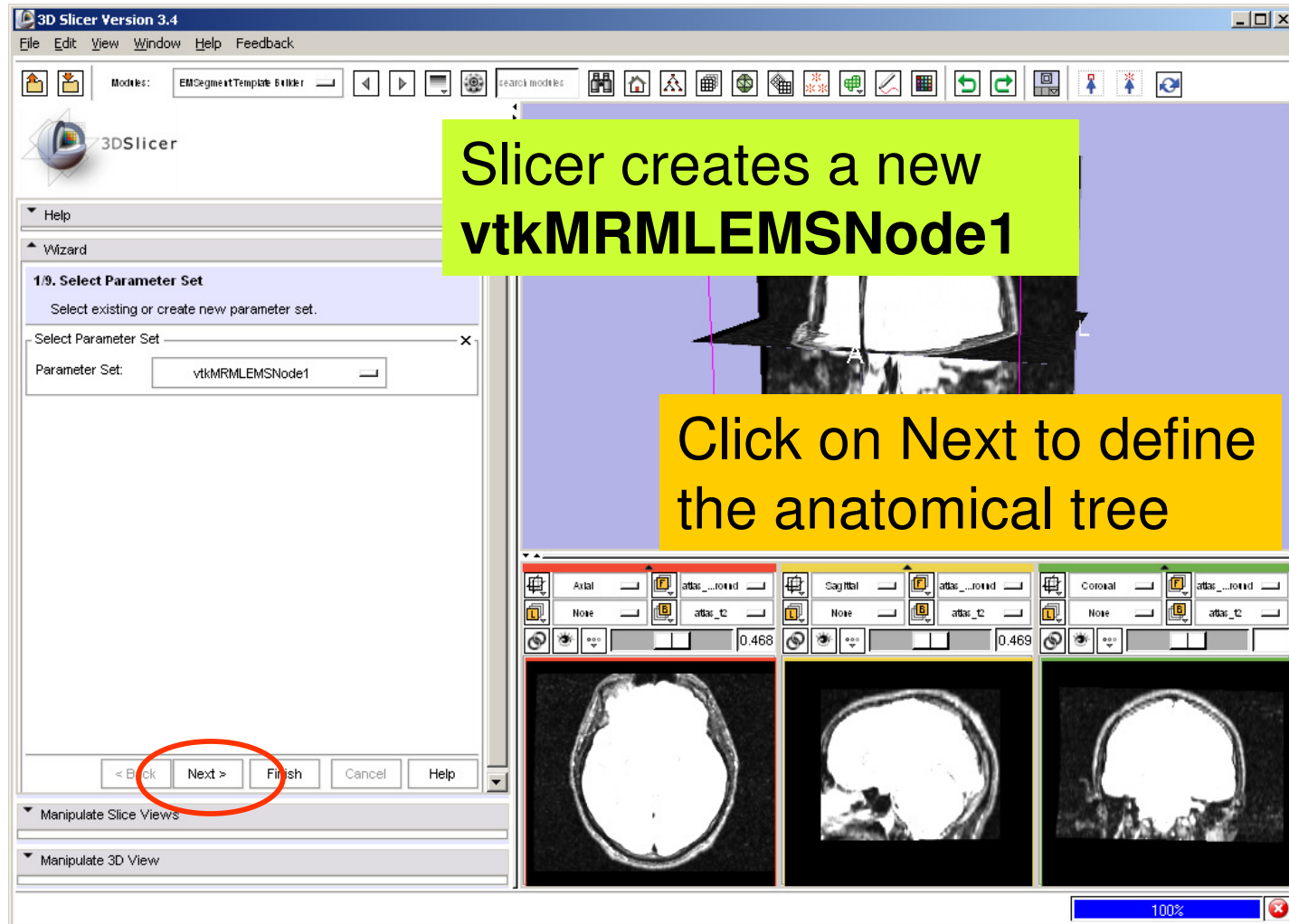


Automatic Segmentation: Sonia Fajol, Ph.D., Harvard Medical School
National Alliance for Medical Image Computing

Template Builder

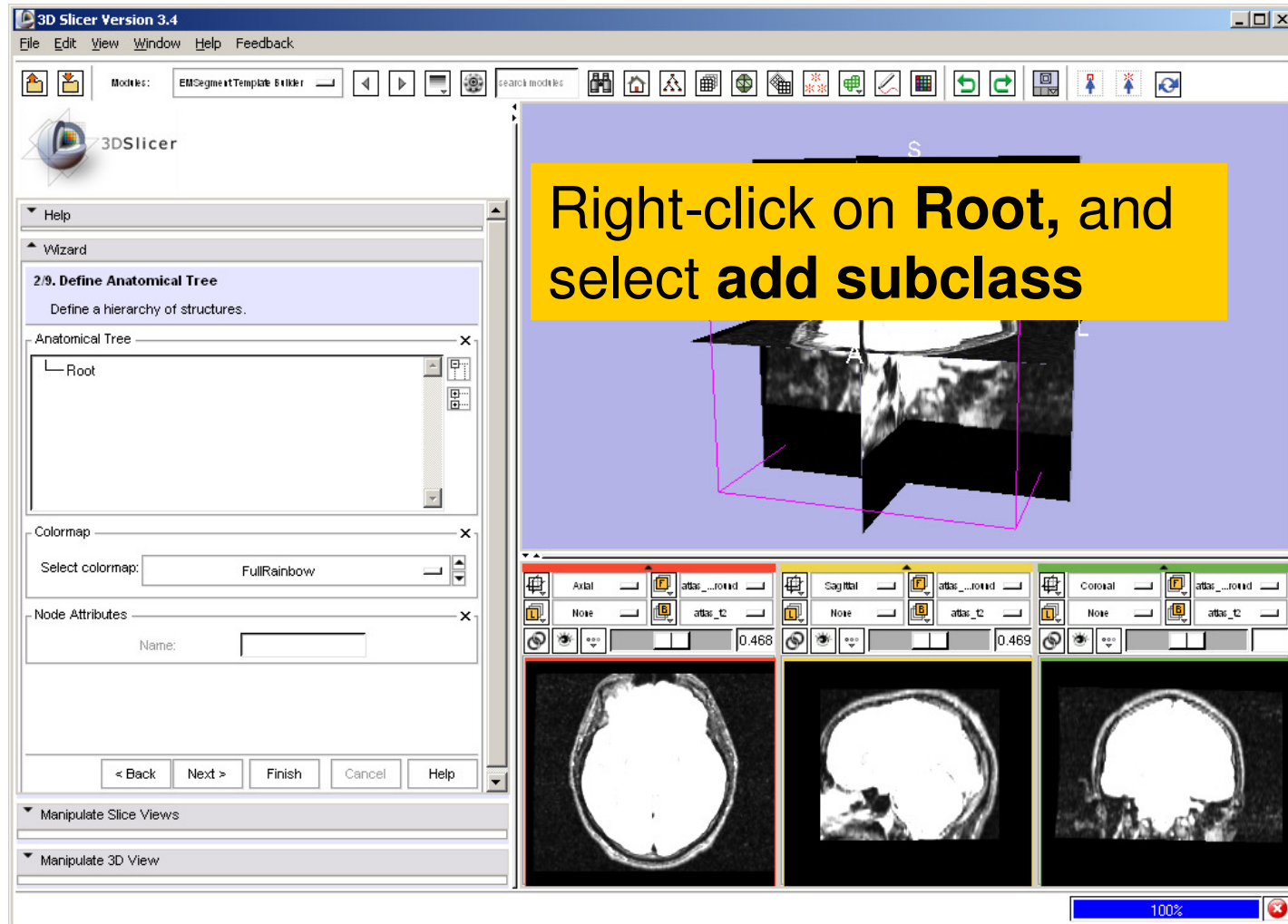


Template Builder



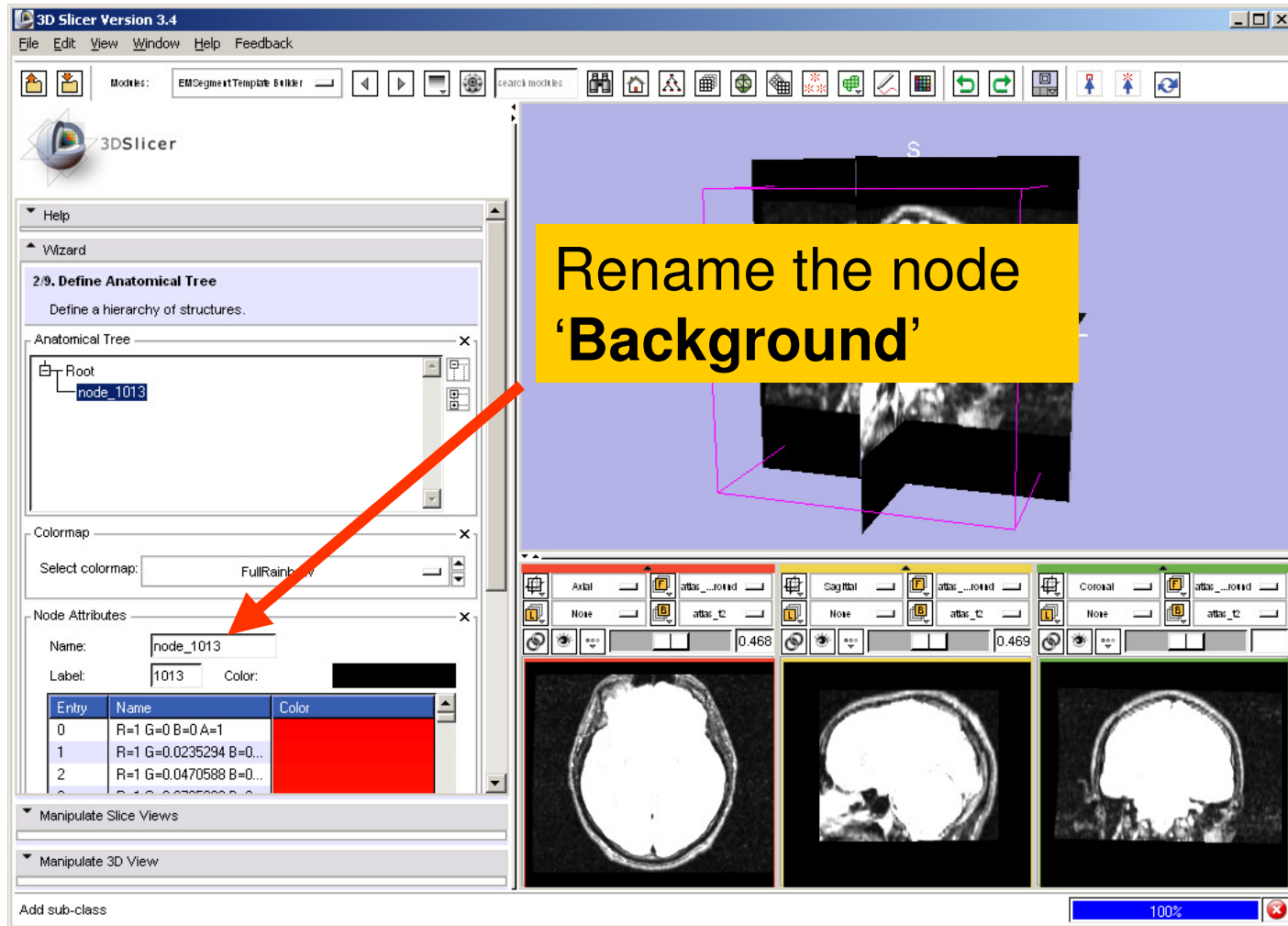
Anatomical Tree

Anatomical Tree

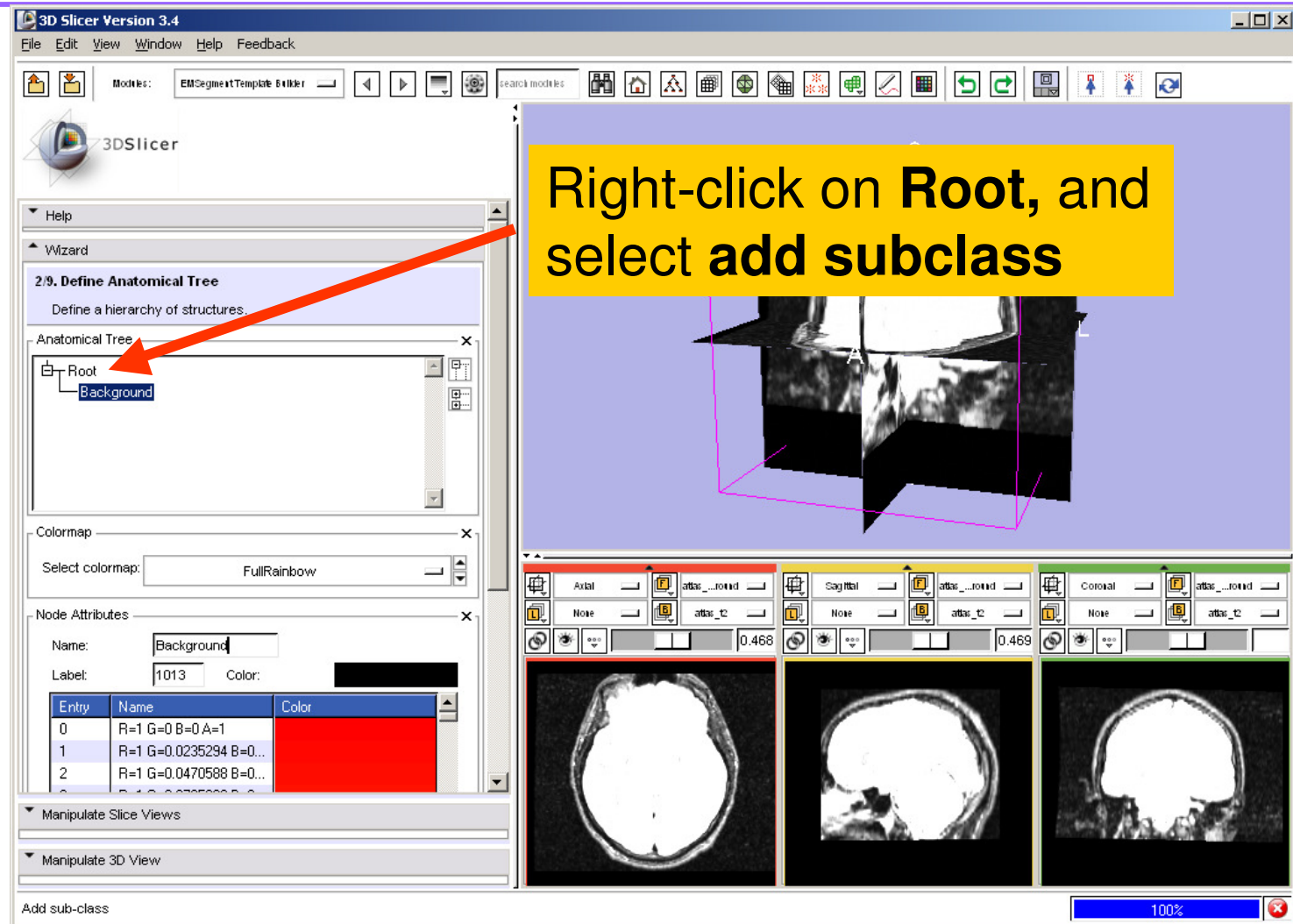


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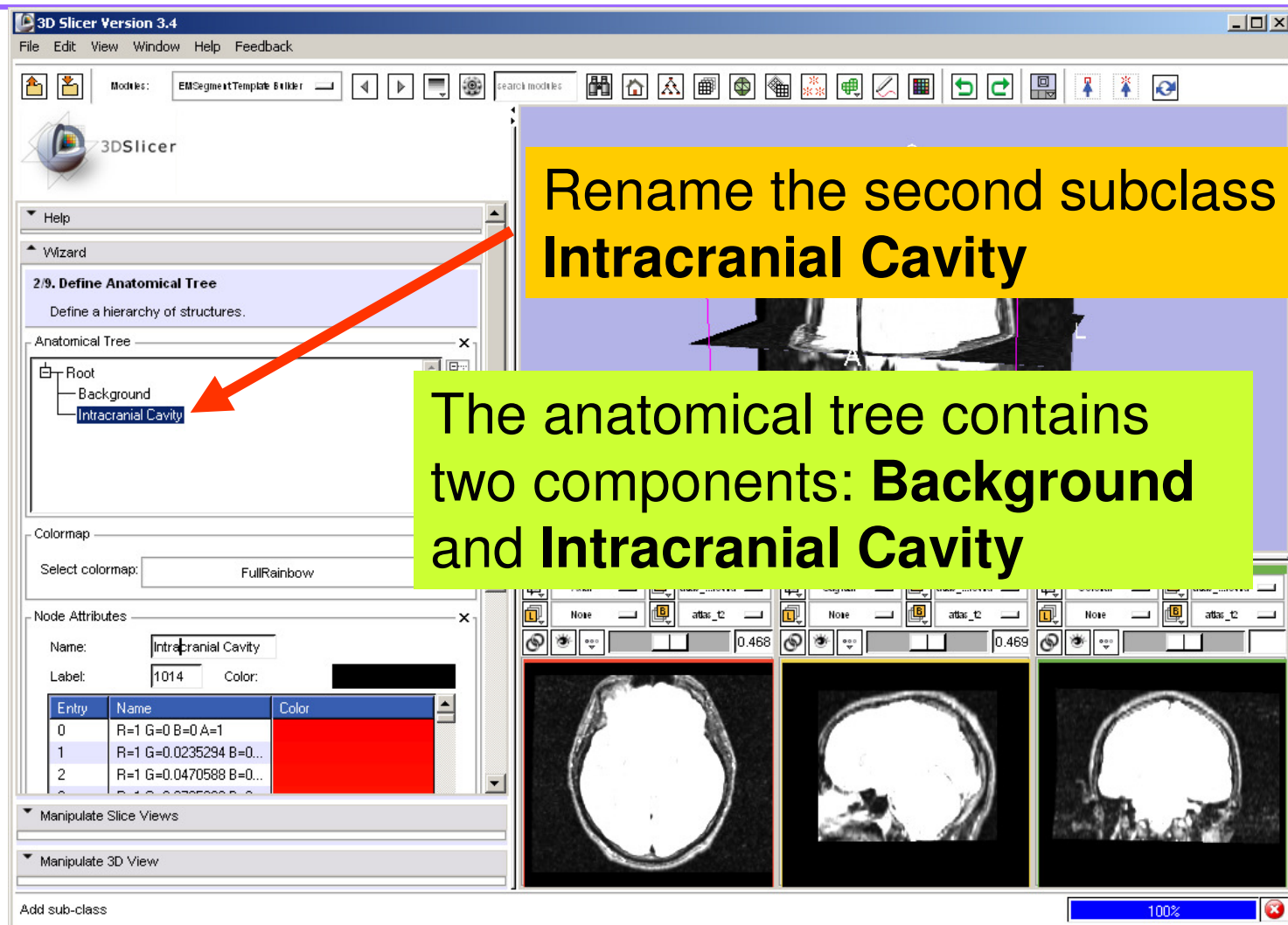
Anatomical Tree



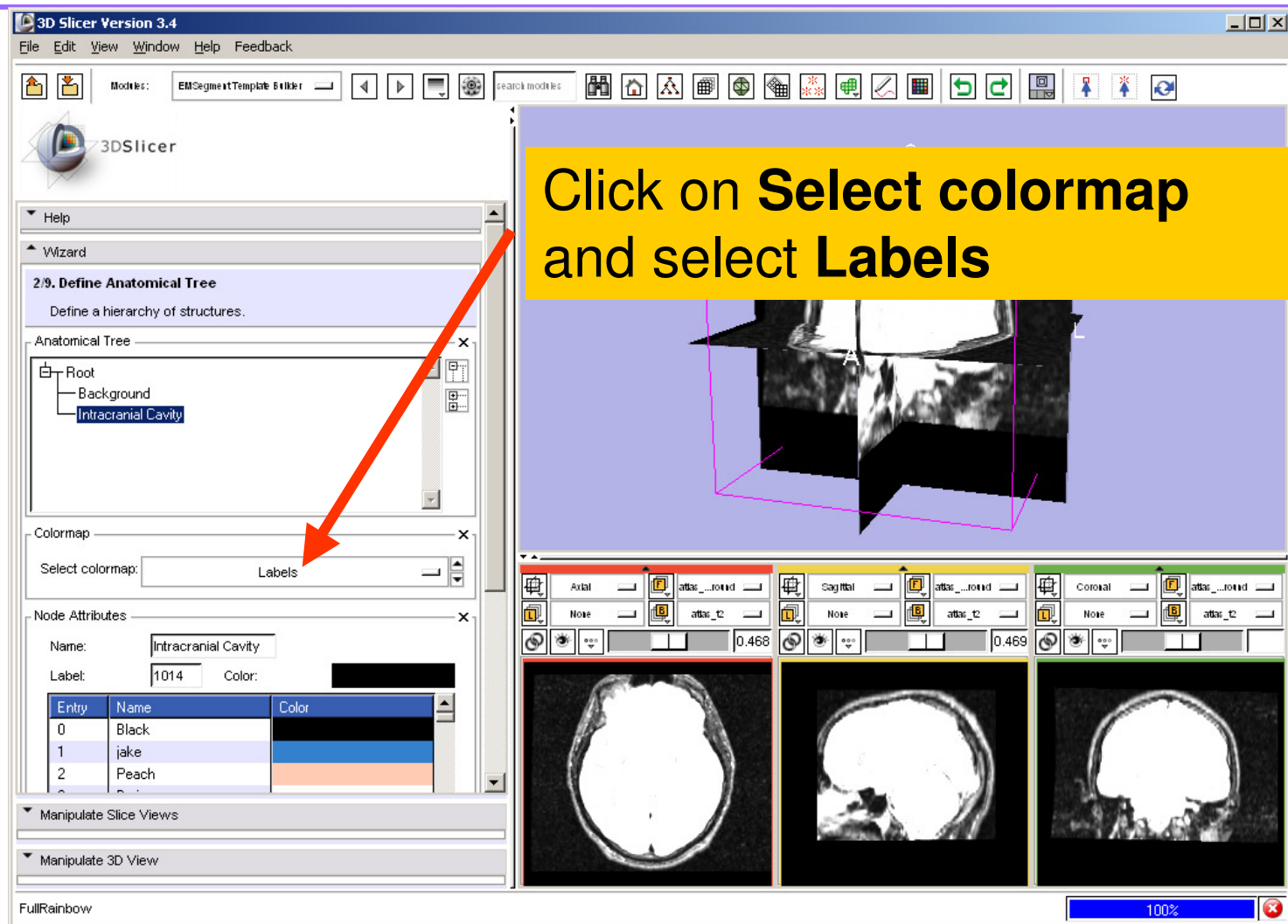
Anatomical Tree



Anatomical Tree



Anatomical Tree



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Anatomical Tree

3D Slicer Version 3.4

File Edit View Window Help Feedback

Modules: EMSegmentTemplate Builder

2.9. Define Anatomical Tree

Define a hierarchy of structures.

Anatomical Tree

- Root
 - Background
 - Air
 - Intracranial Cavity

Colormap

Select colormap: Labels

Node Attributes

Name: Air

Label: 0

Entry	Name	Color
0	Black	
1	jake	
2	Peach	

Manipulate Slice Views

Manipulate 3D View

Add sub-class

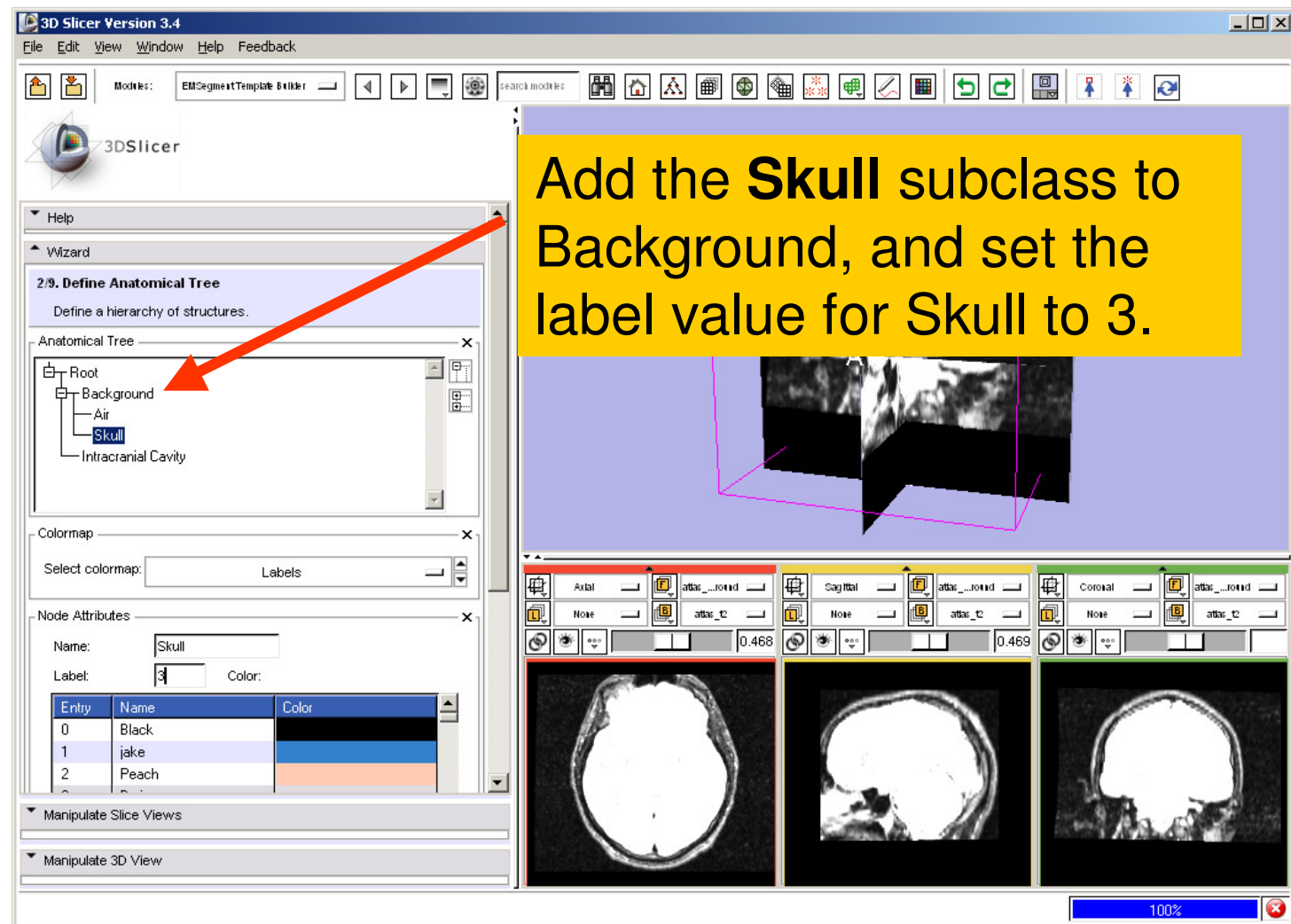
100%

Right-click on **Background**, and select add subclass.

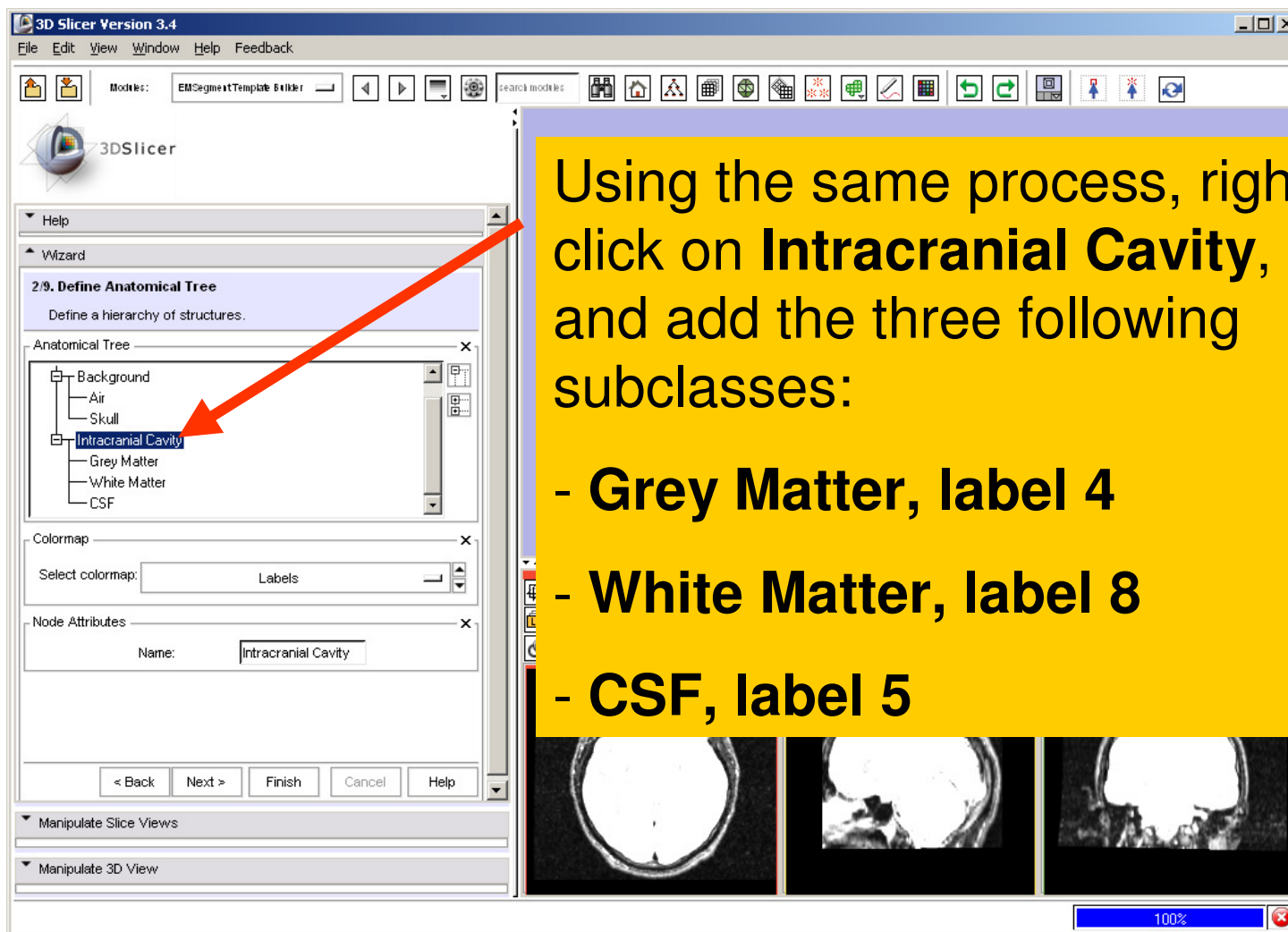
Add the subclass **Air** to **Background**.

Set the label value for **Air** to 0.

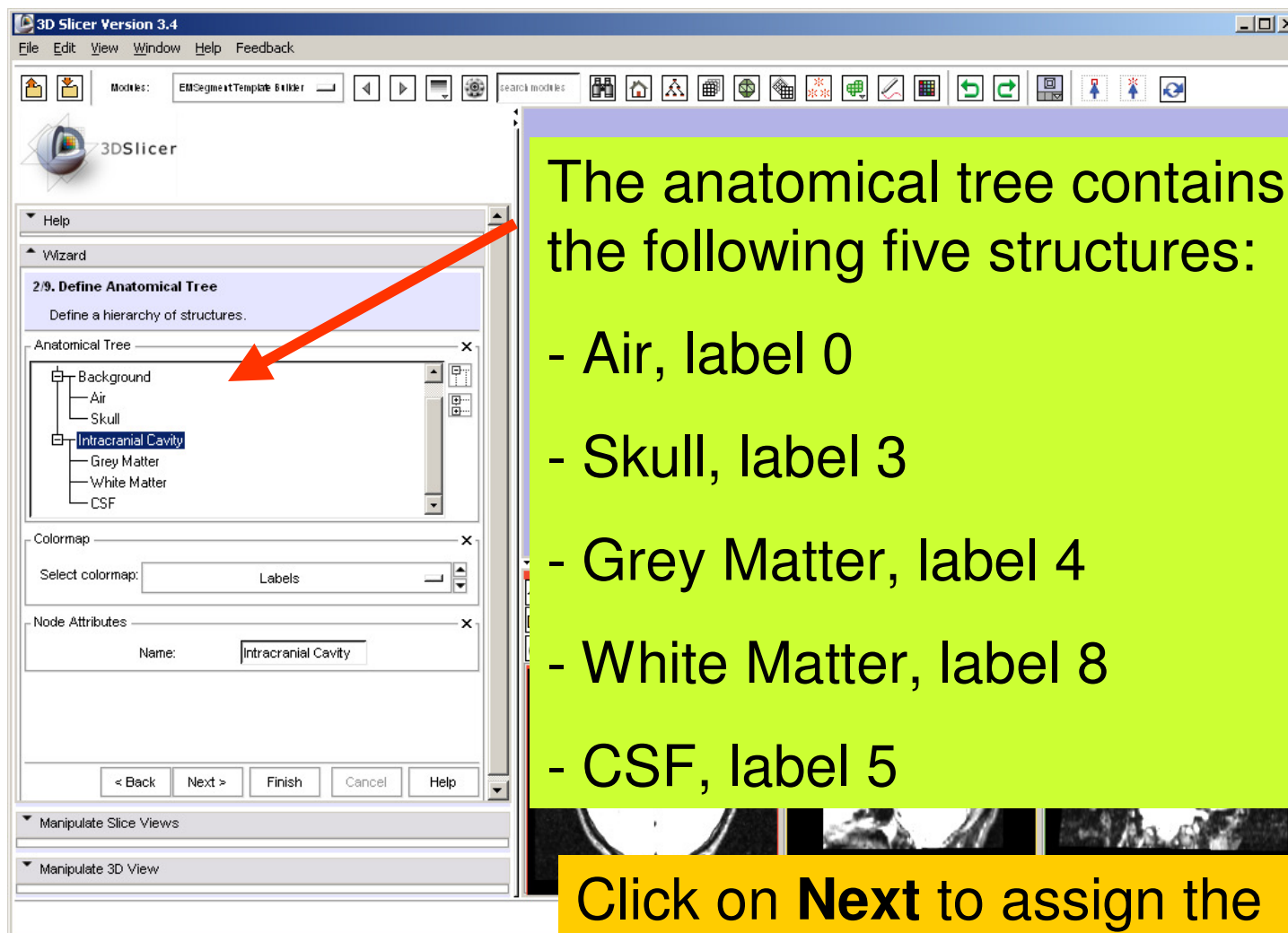
Anatomical Tree

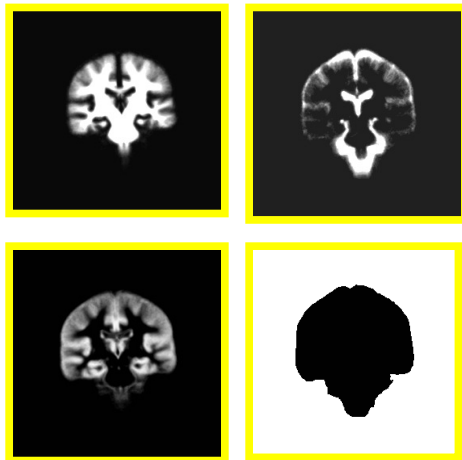


Anatomical Tree



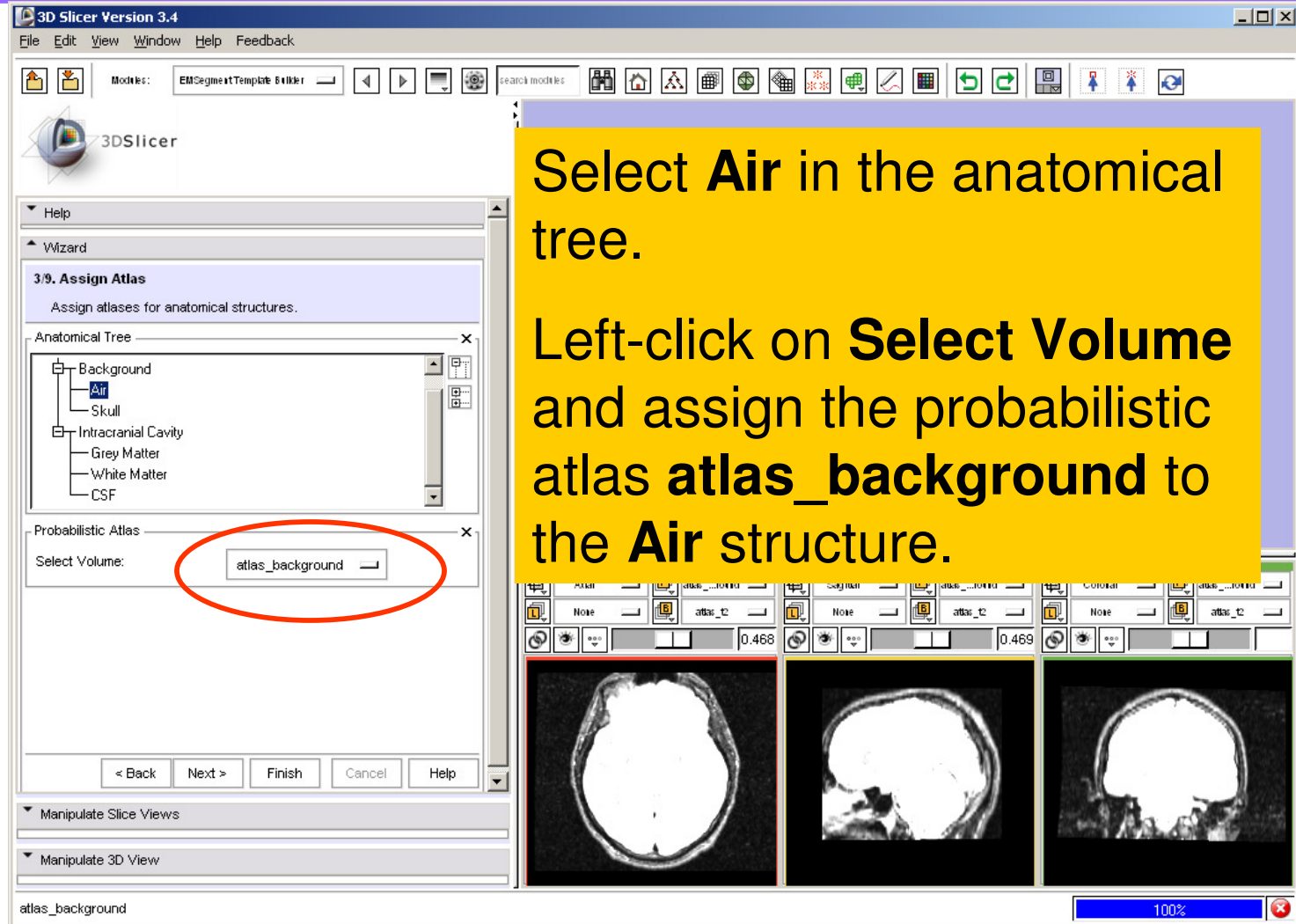
Anatomical Tree



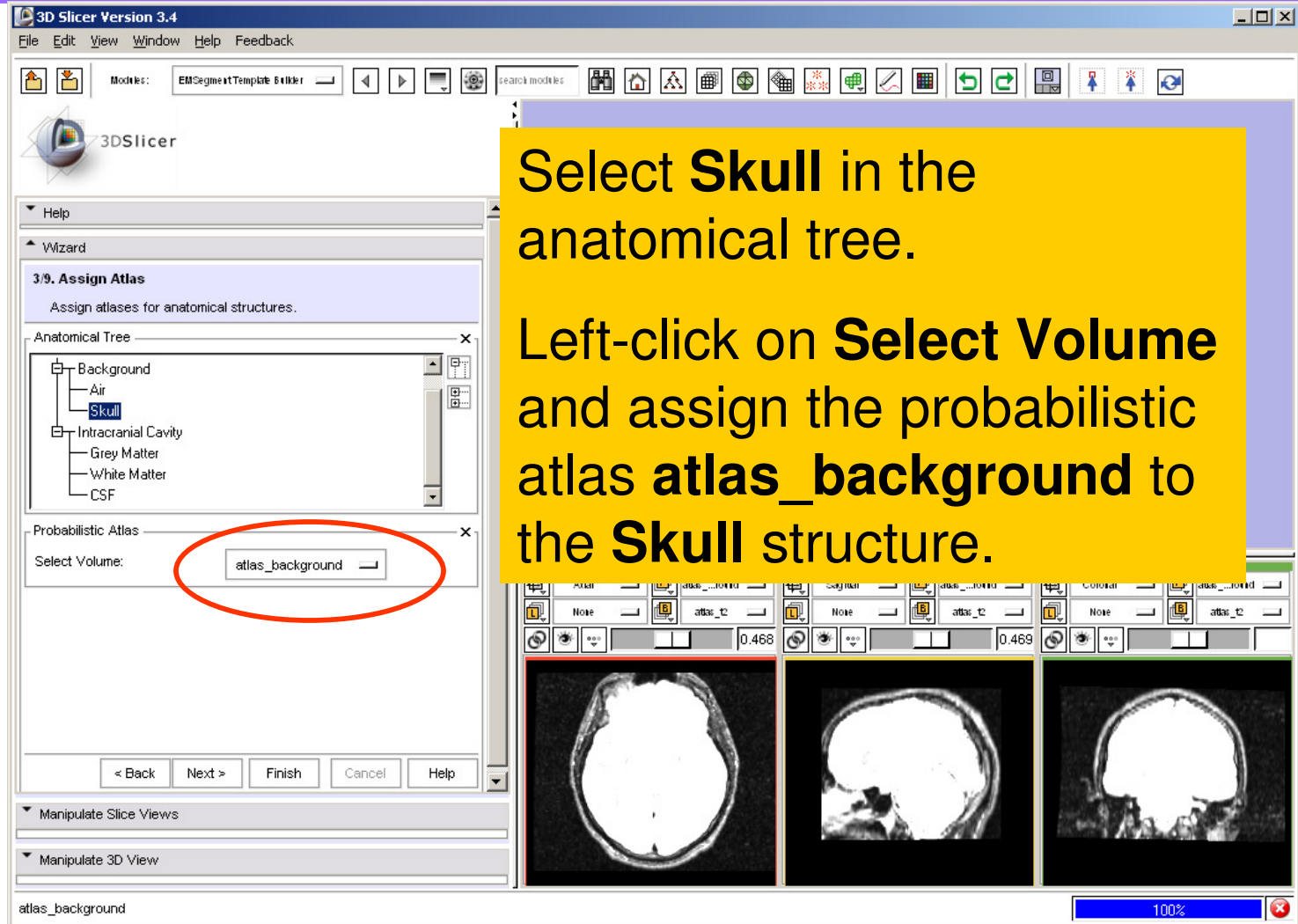


Atlas Assignment

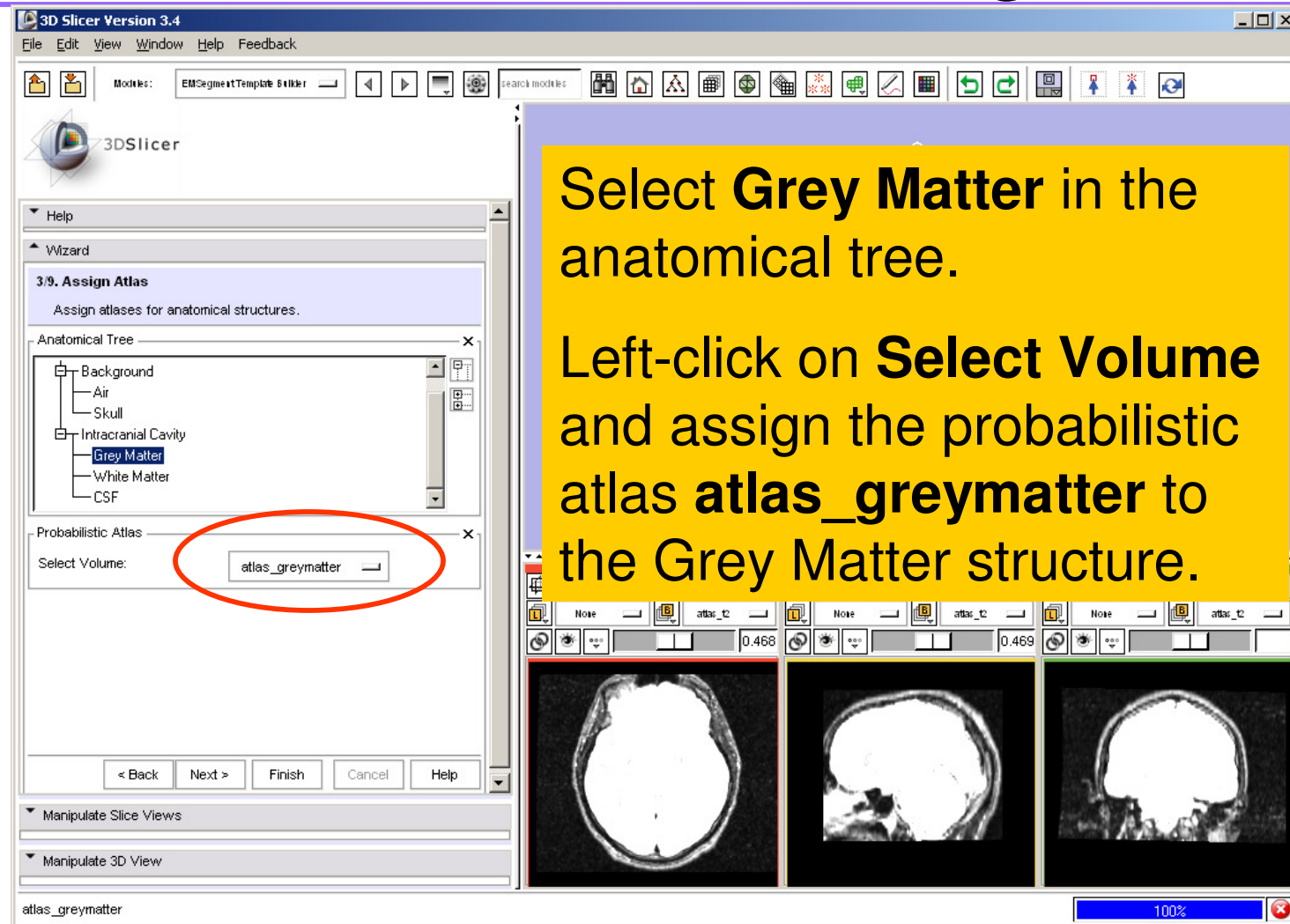
Atlas Assignment



Atlas Assignment

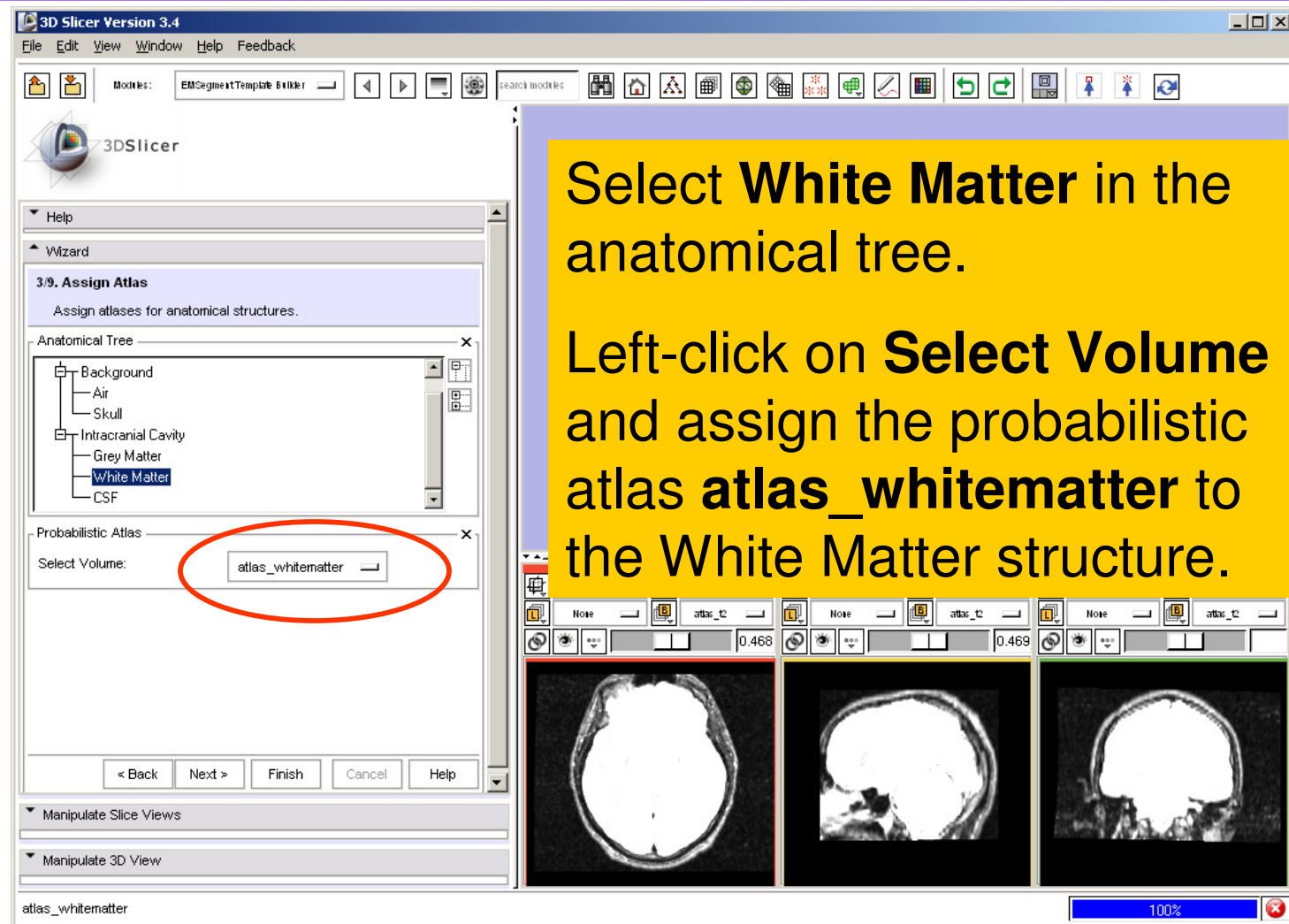


Atlas Assignment

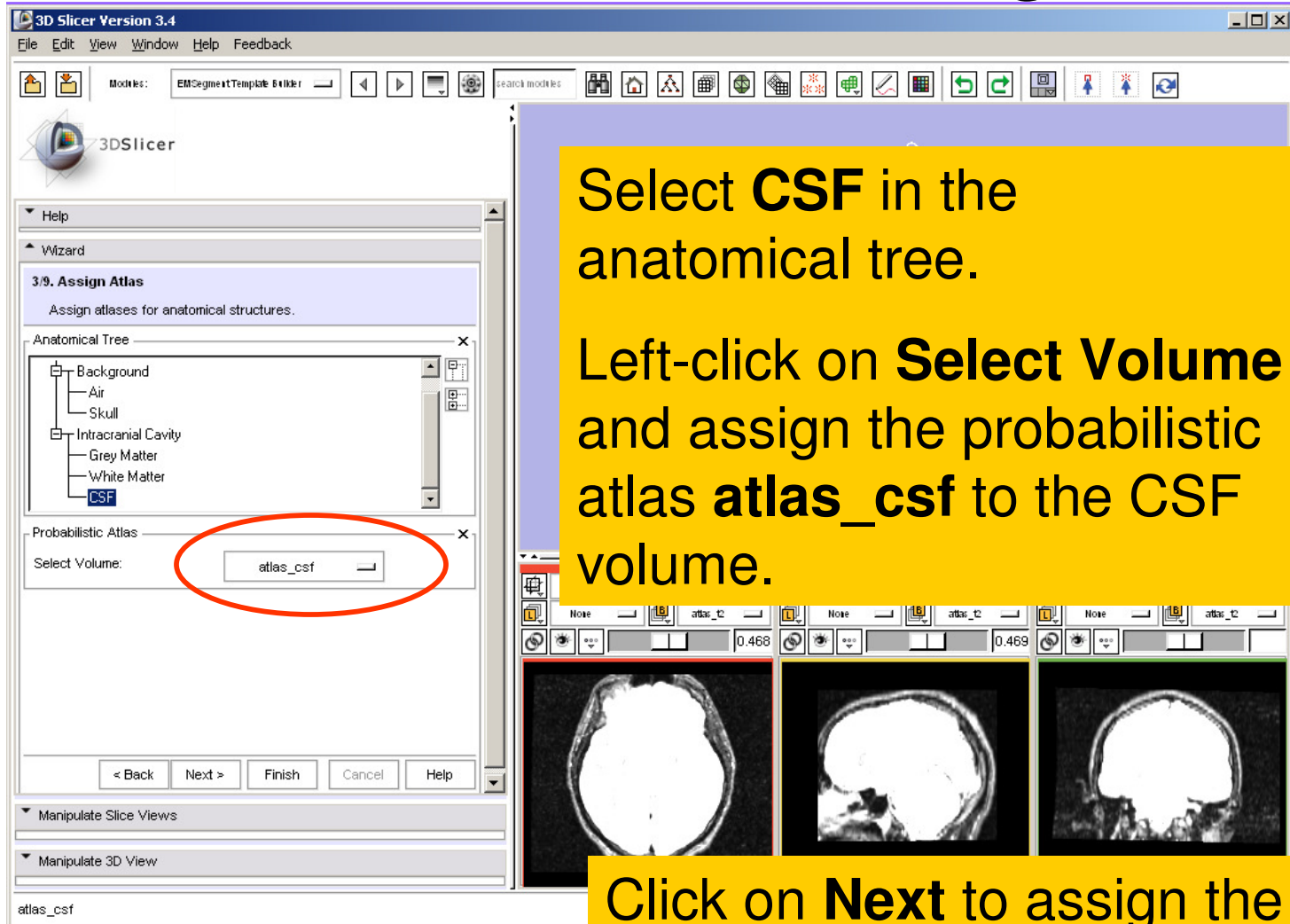


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Atlas Assignment



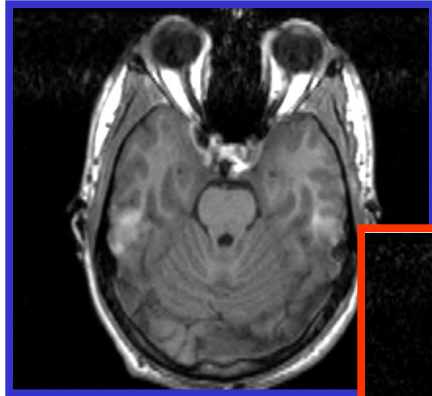
Atlas Assignment



Select **CSF** in the anatomical tree.

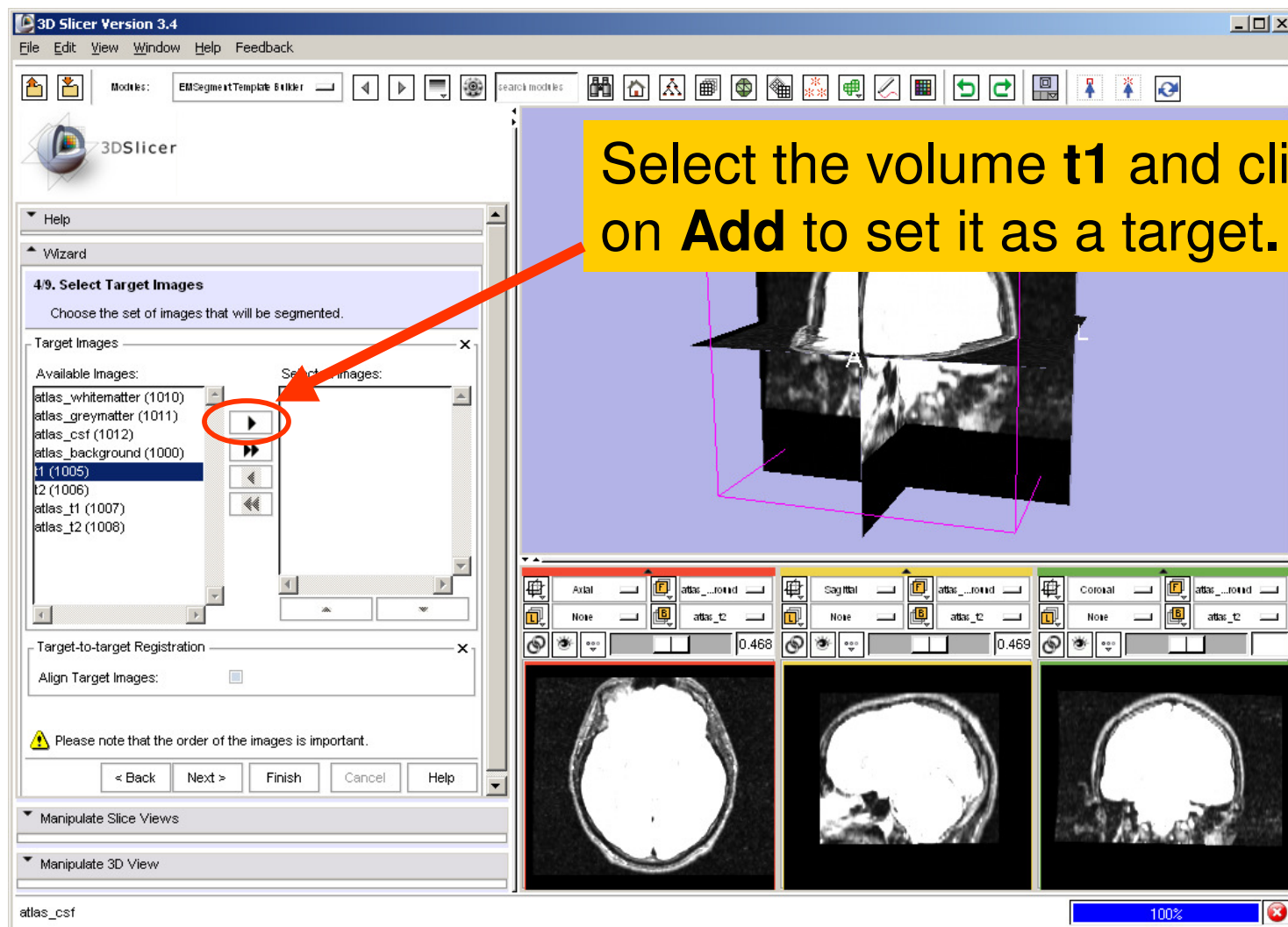
Left-click on **Select Volume** and assign the probabilistic atlas **atlas_csf** to the CSF volume.

Click on **Next** to assign the atlas to select the target images.



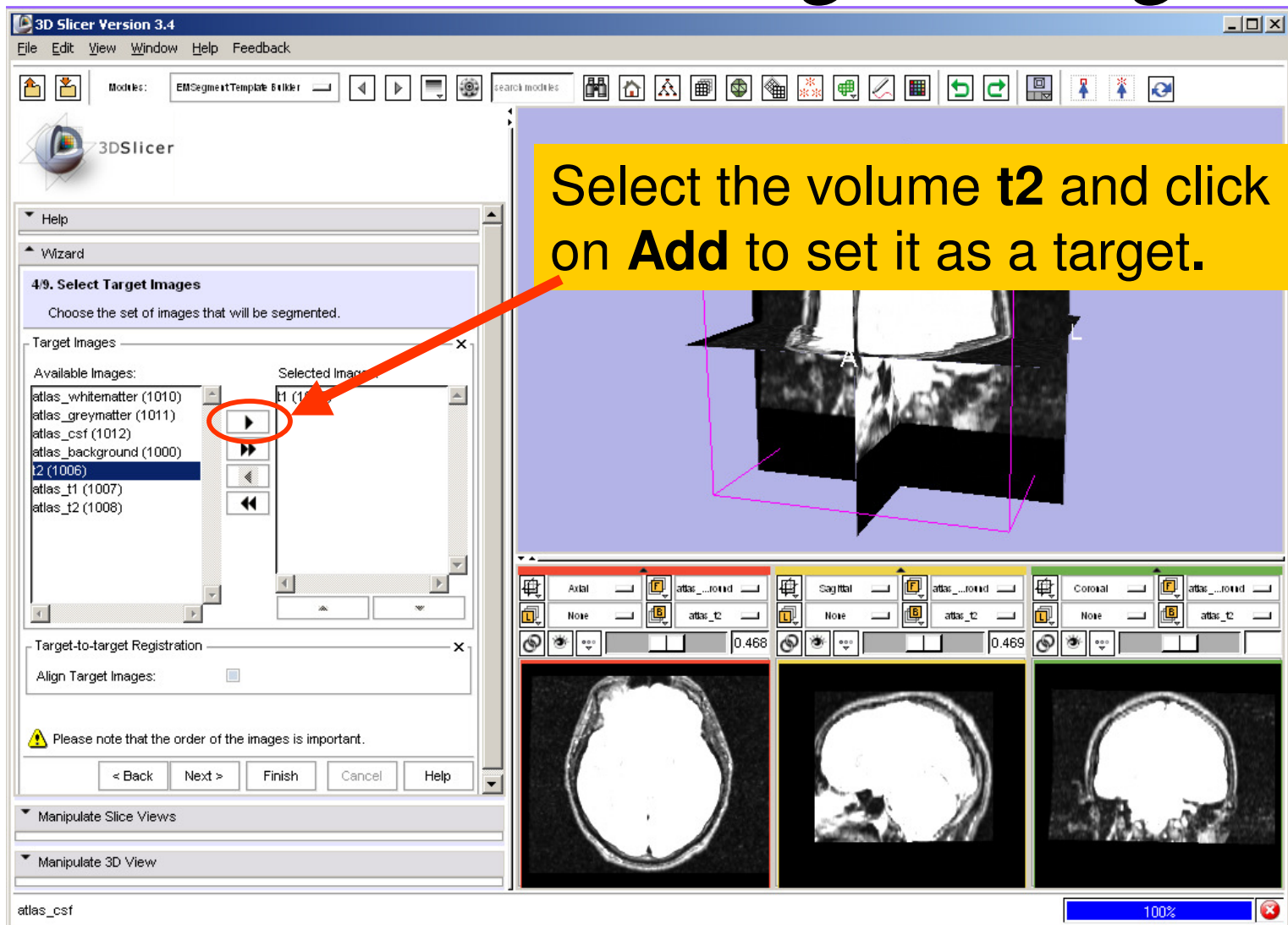
Target Images

Target Images



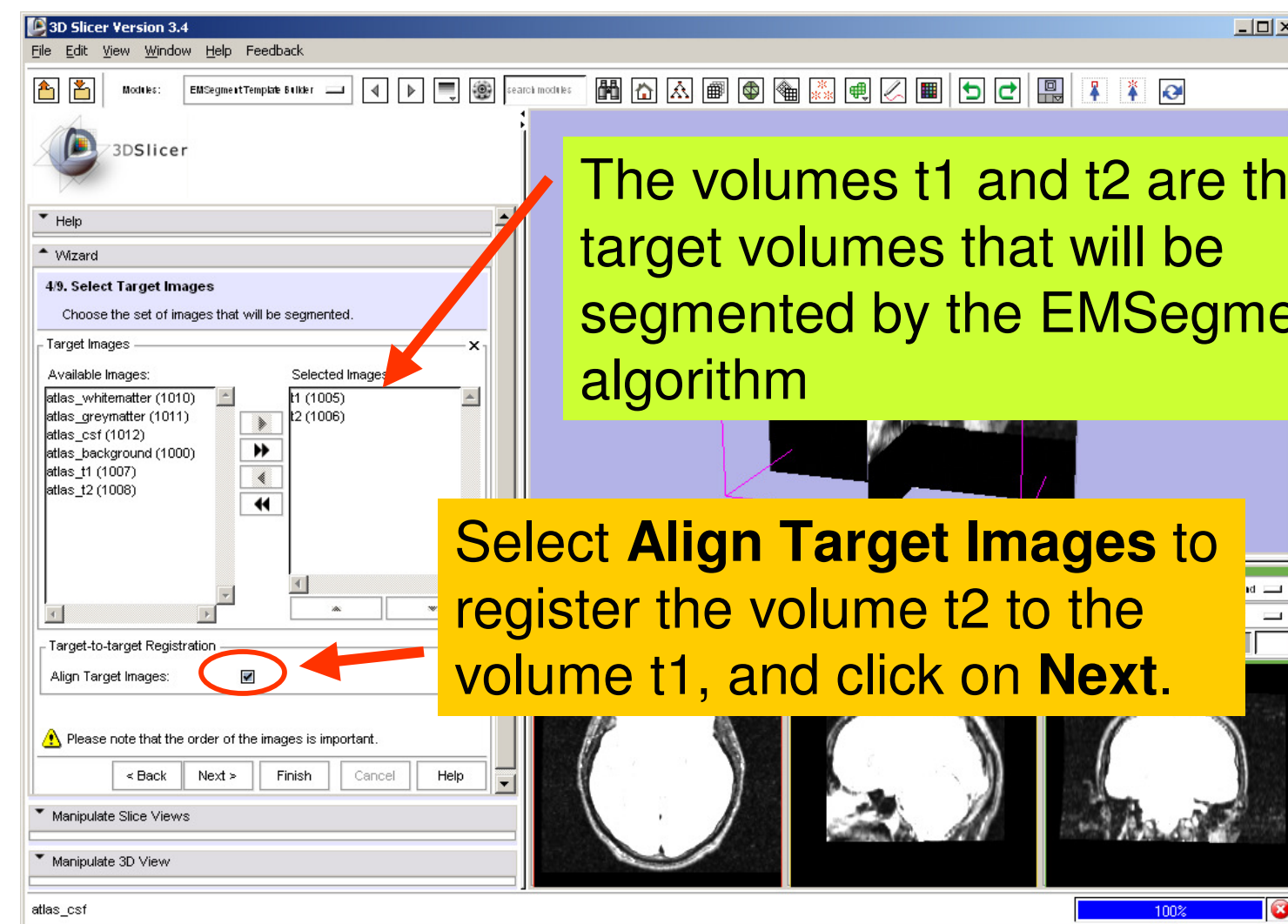
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Target Images

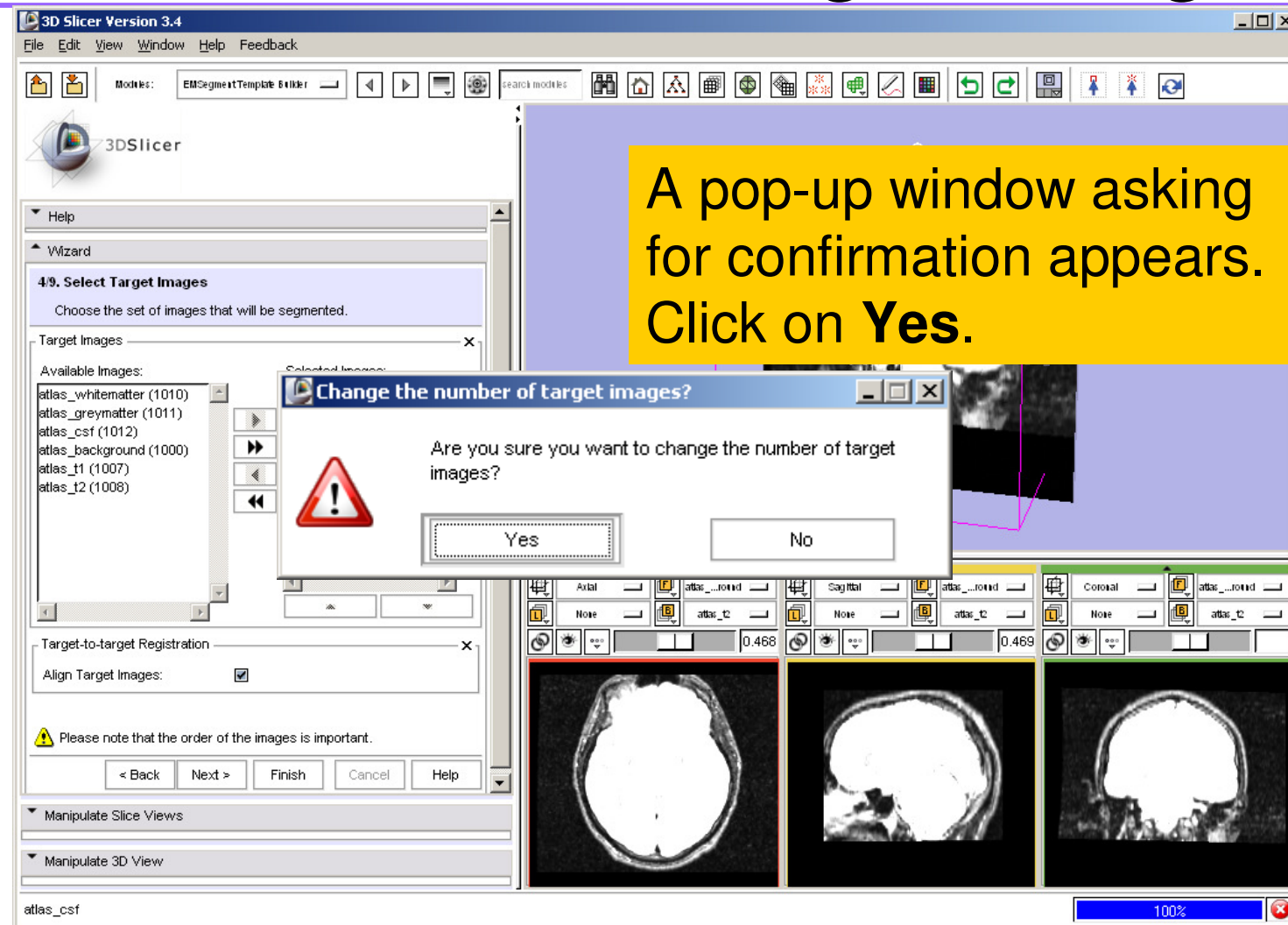


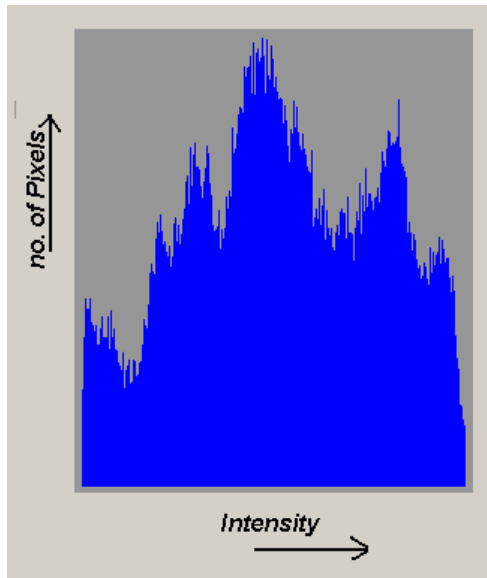
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Target Images



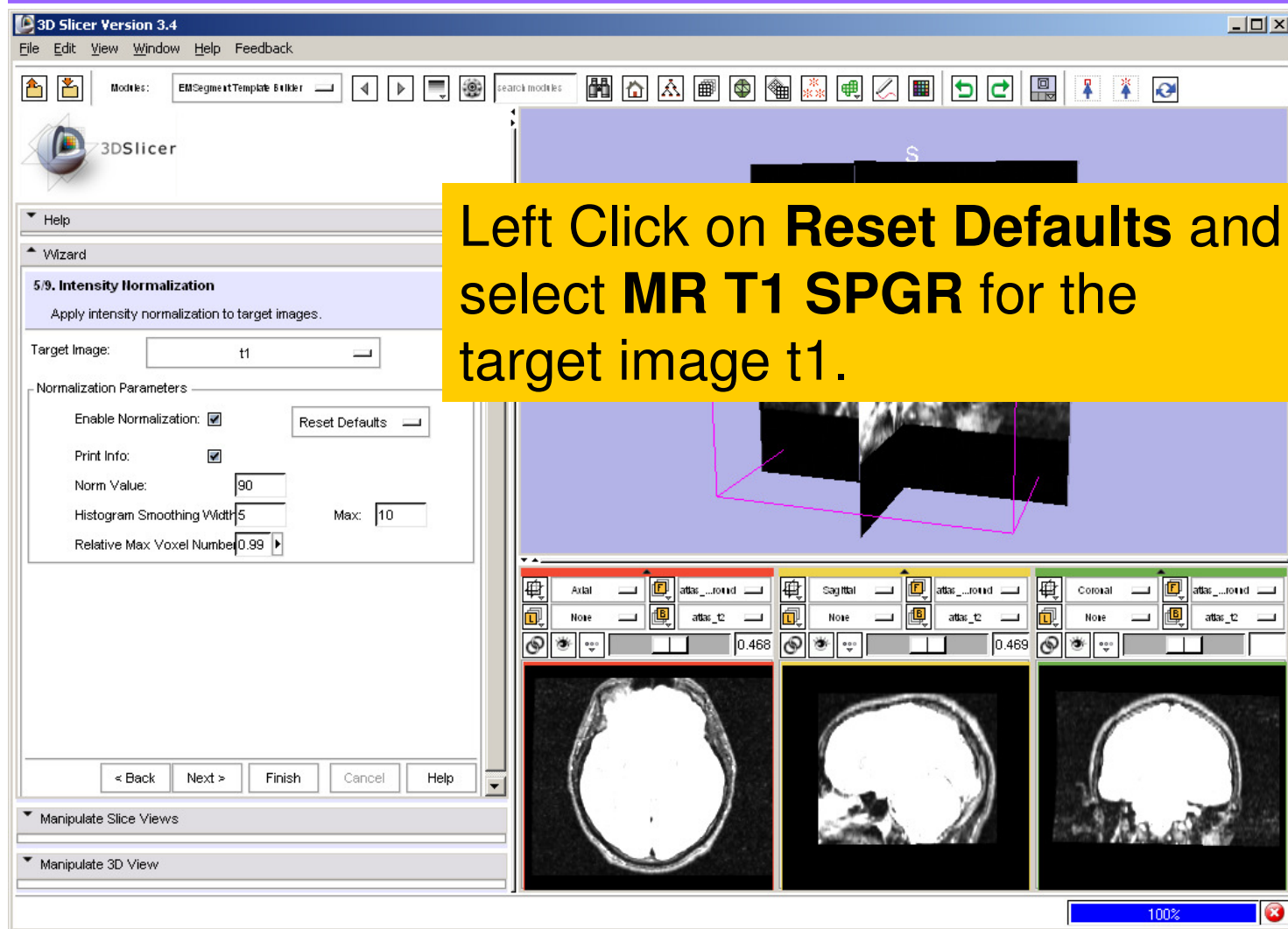
Target Images





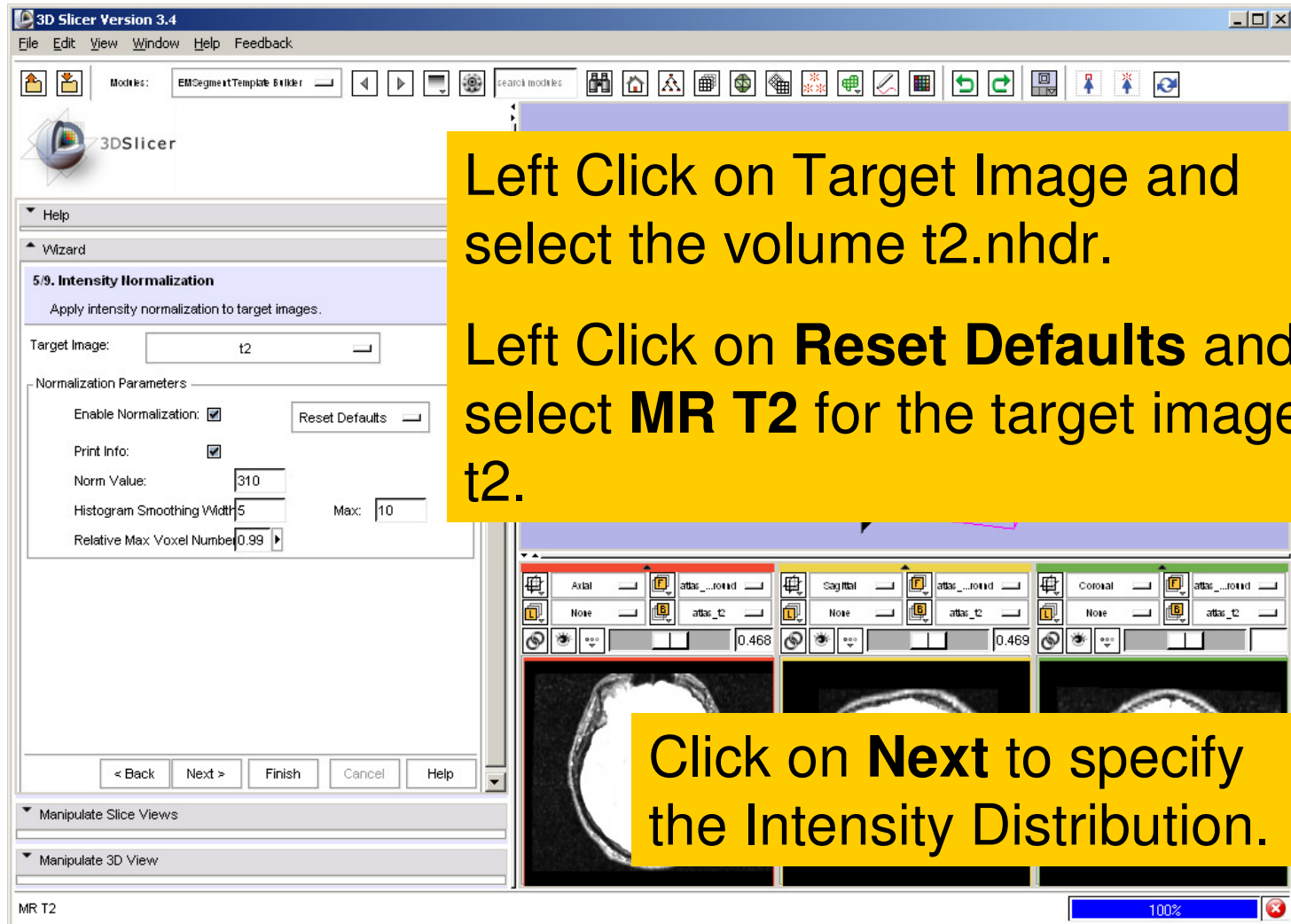
Intensity Normalization

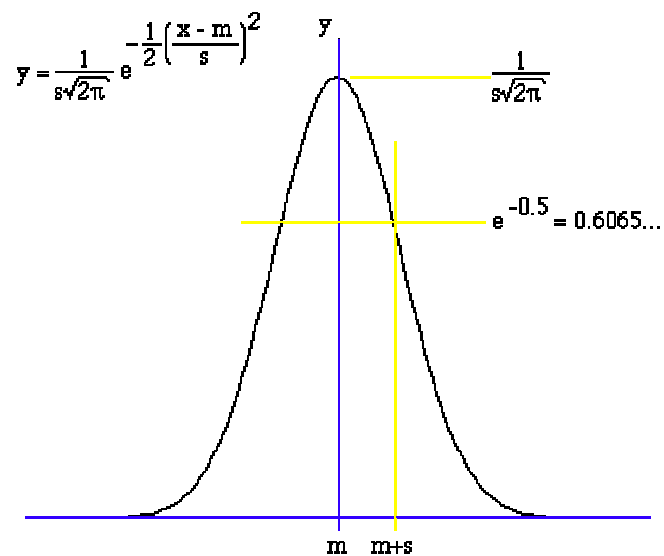
Intensity Normalization



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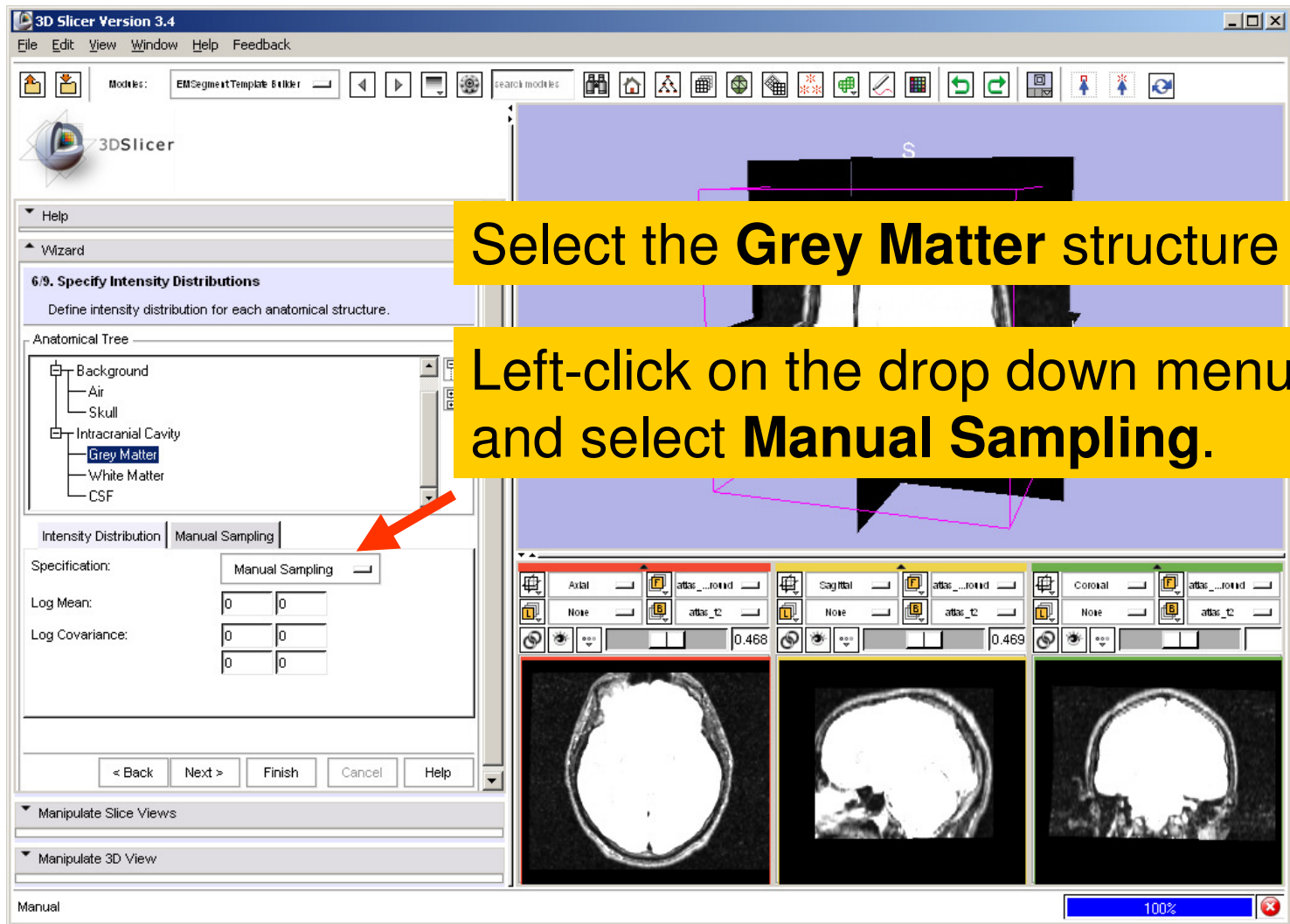
Intensity Normalization





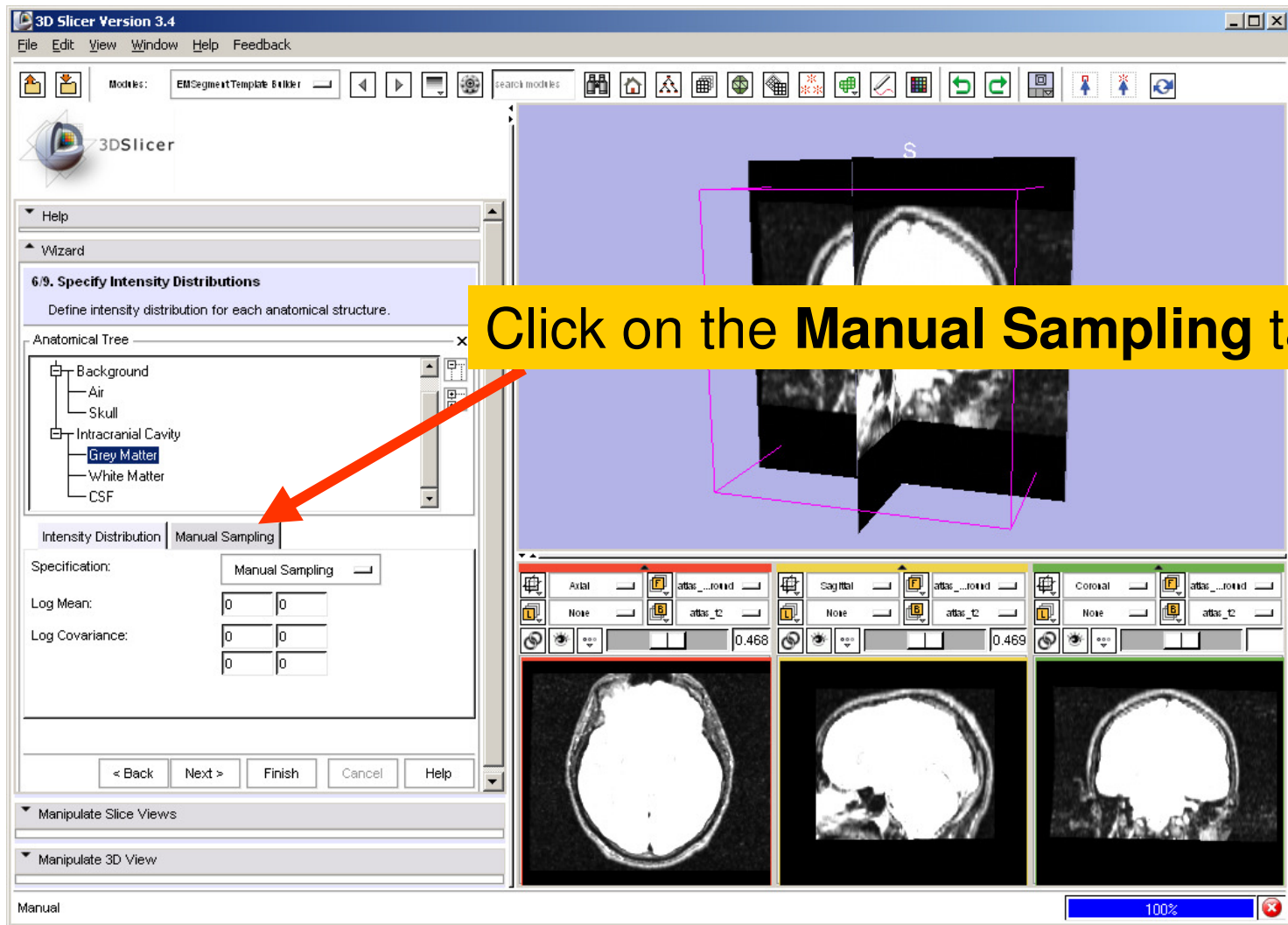
Intensity Distribution

Gaussian Intensity Distribution



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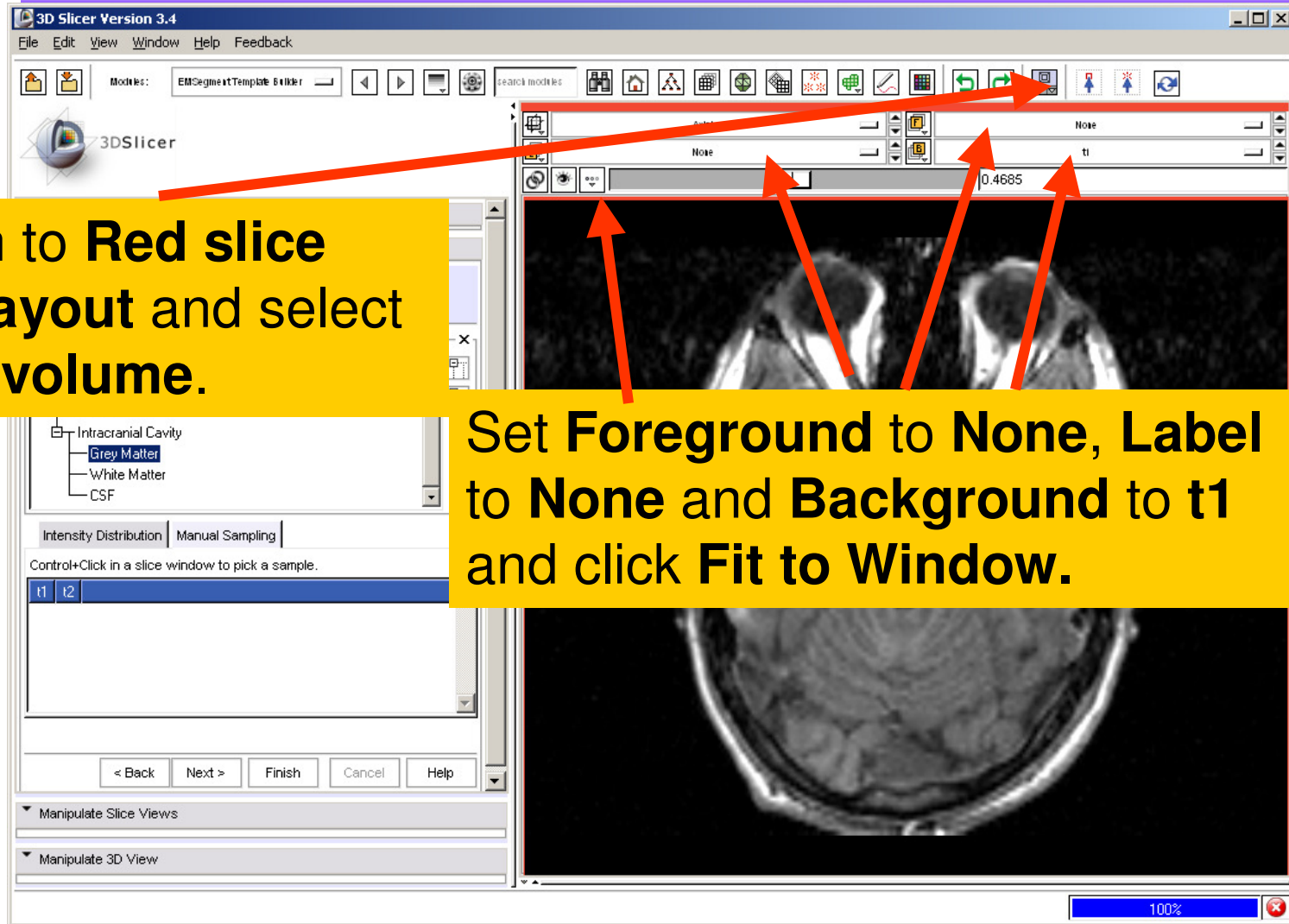
Gaussian Intensity Distribution



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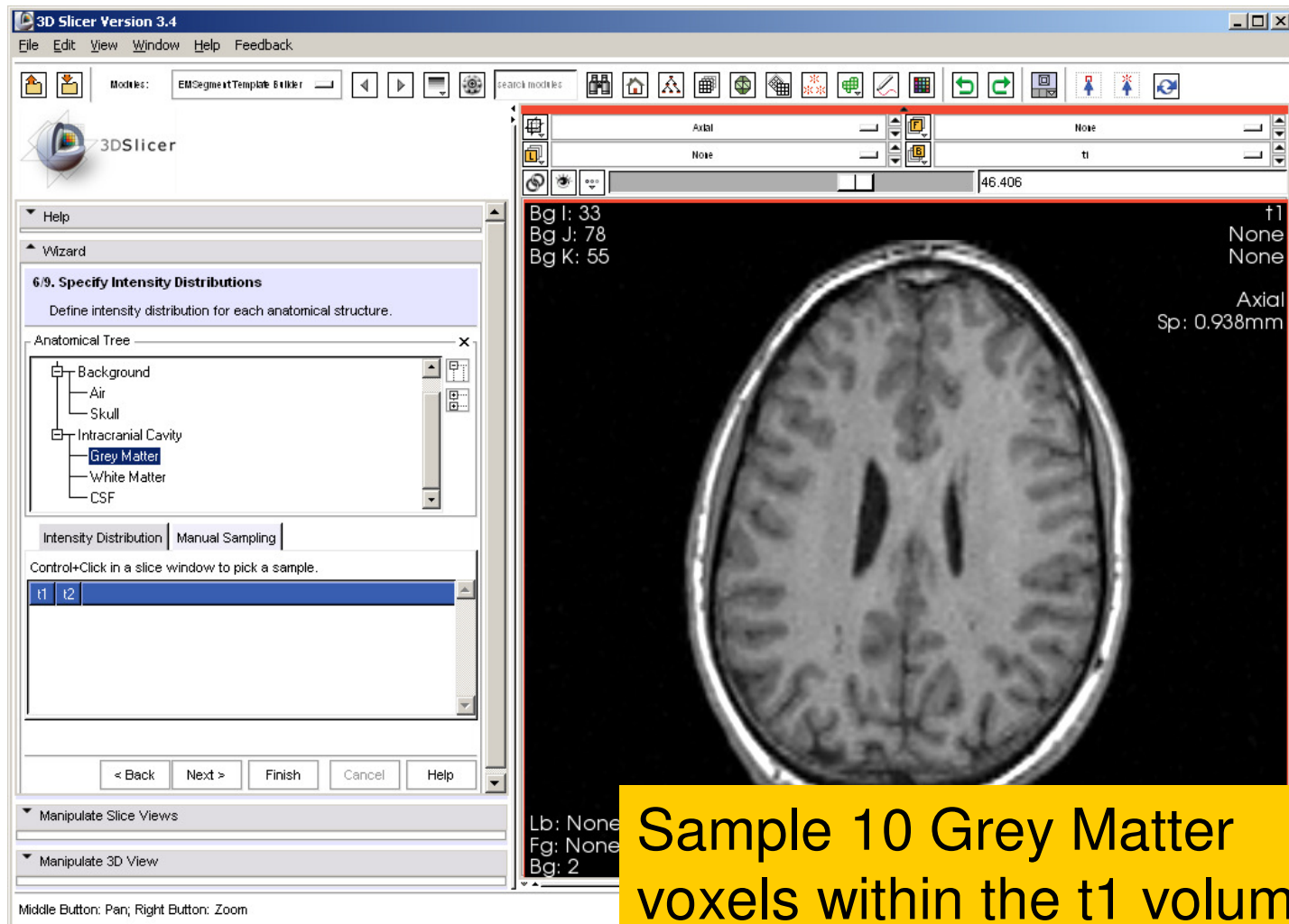
Gaussian Intensity Distribution

Switch to **Red slice only layout** and select the **t1** volume.



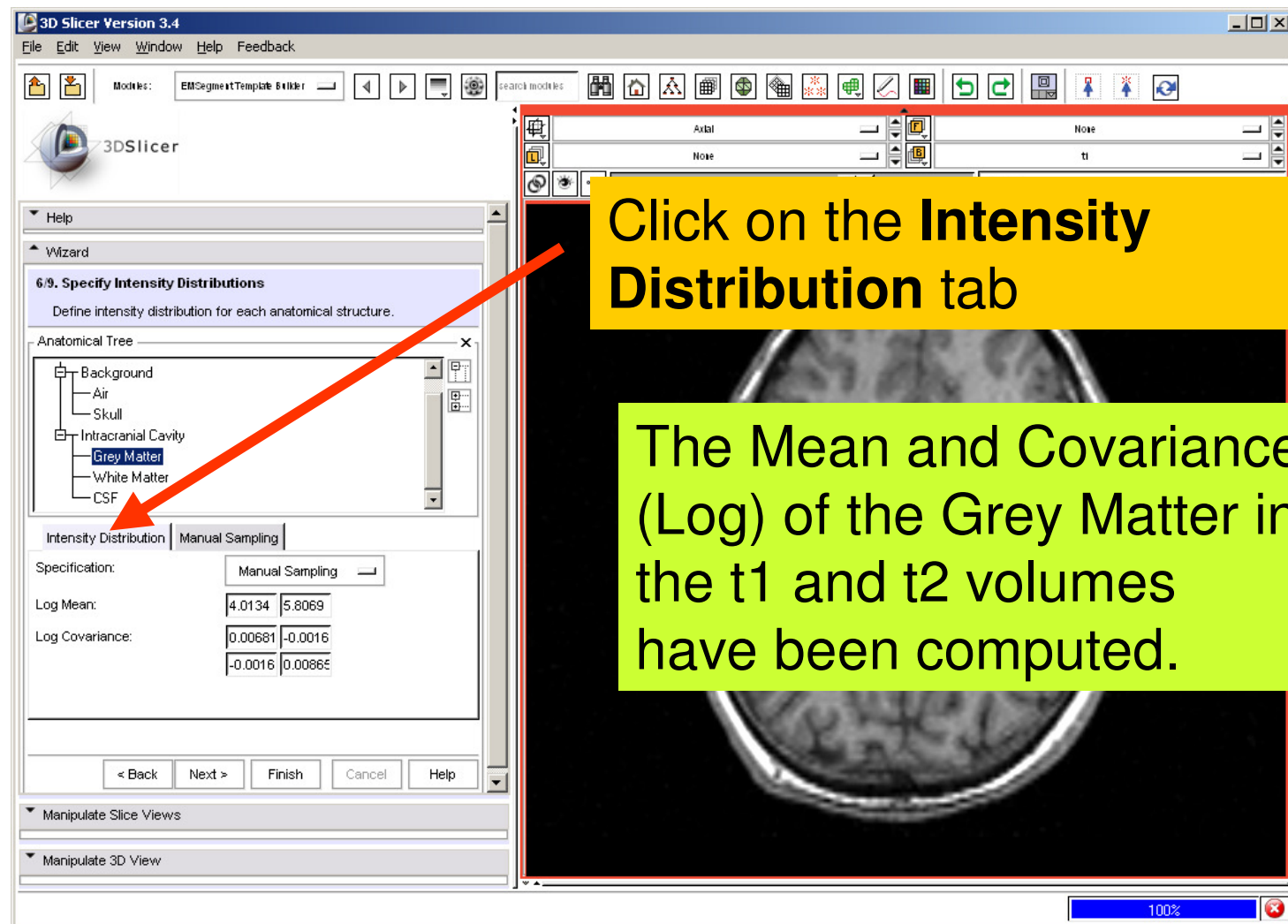
Set **Foreground to None**, **Label to None** and **Background to t1** and click **Fit to Window**.

Gaussian Intensity Distribution



**Sample 10 Grey Matter
voxels within the t1 volume
(Ctrl + Left Click)**

Gaussian Intensity Distribution



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Gaussian Intensity Distribution

3D Slicer Version 3.4

File Edit View Window Help Feedback

Modules: EMSegmentTemplateBuilder

6/9. Specify Intensity Distributions

Define intensity distribution for each anatomical structure.

Anatomical Tree

- Background
 - Air
 - Skull
- Intracranial Cavity
 - Grey Matter
 - White Matter**
 - CSF

Intensity Distribution | **Manual Sampling**

Control+Click in a slice window to pick a sample.

t1	t2
81.000000	219.000000
85.000000	228.000000
81.000000	229.000000
86.000000	193.000000
83.000000	197.000000

< Back Next > Finish Cancel Help

Manipulate Slice Views

Manipulate 3D View

100%

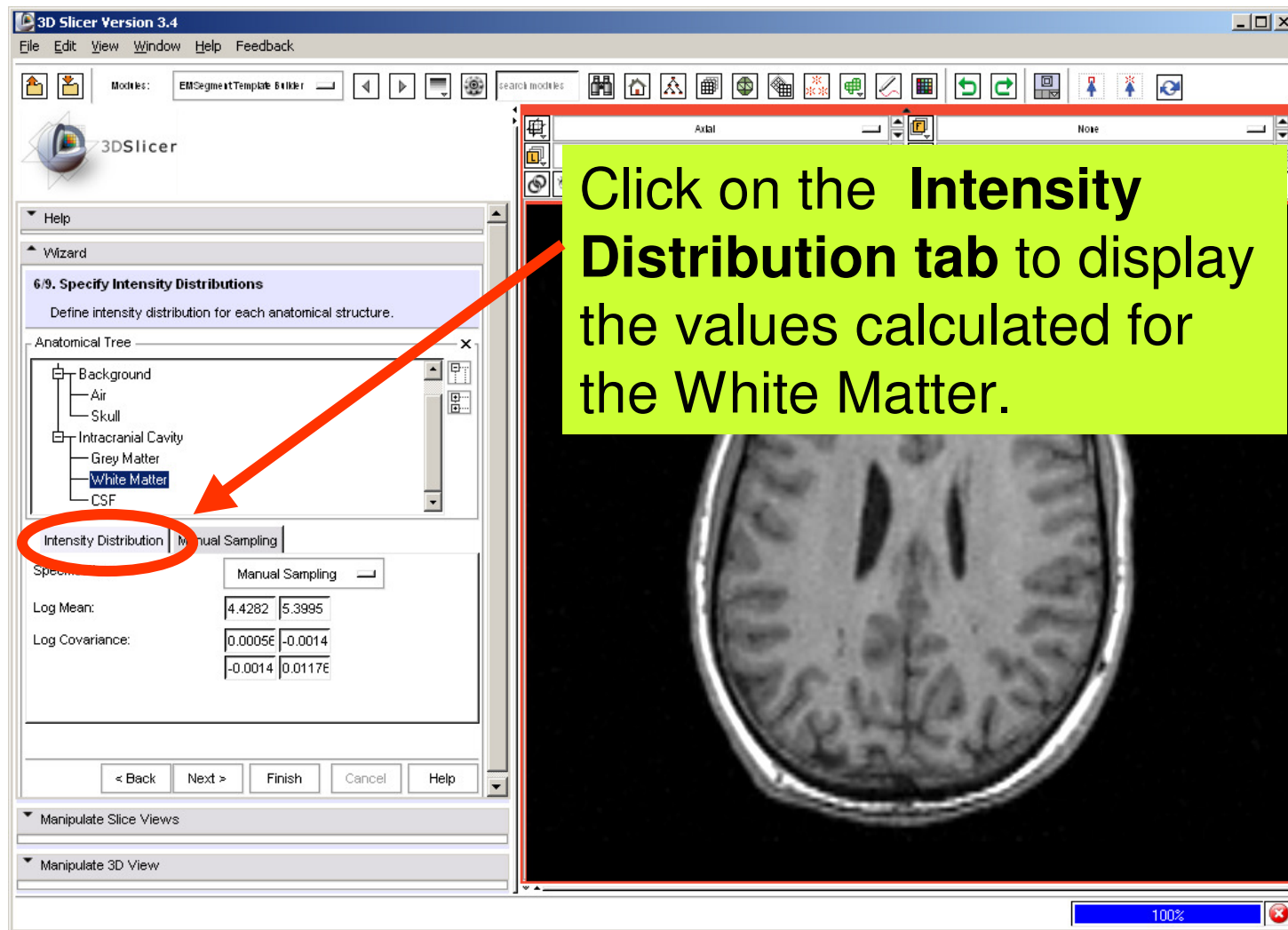
Select **White Matter** in the anatomical tree.

Left-click on the drop down menu and select **Manual Sampling**.

Click on the **Manual Sampling** tab and sample 10 White Matter voxels within the t1 volume

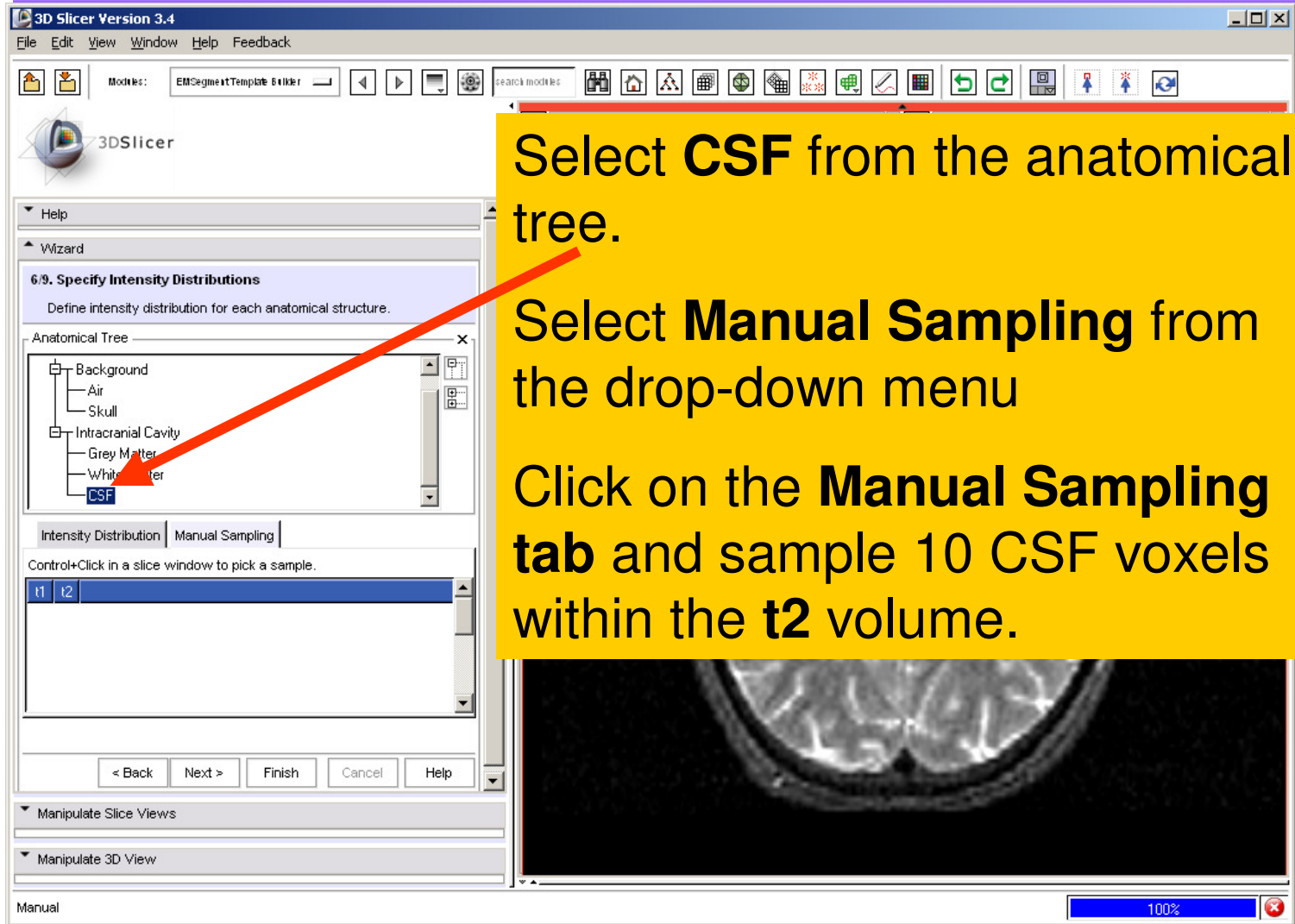
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Gaussian Intensity Distribution



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Gaussian Intensity Distribution



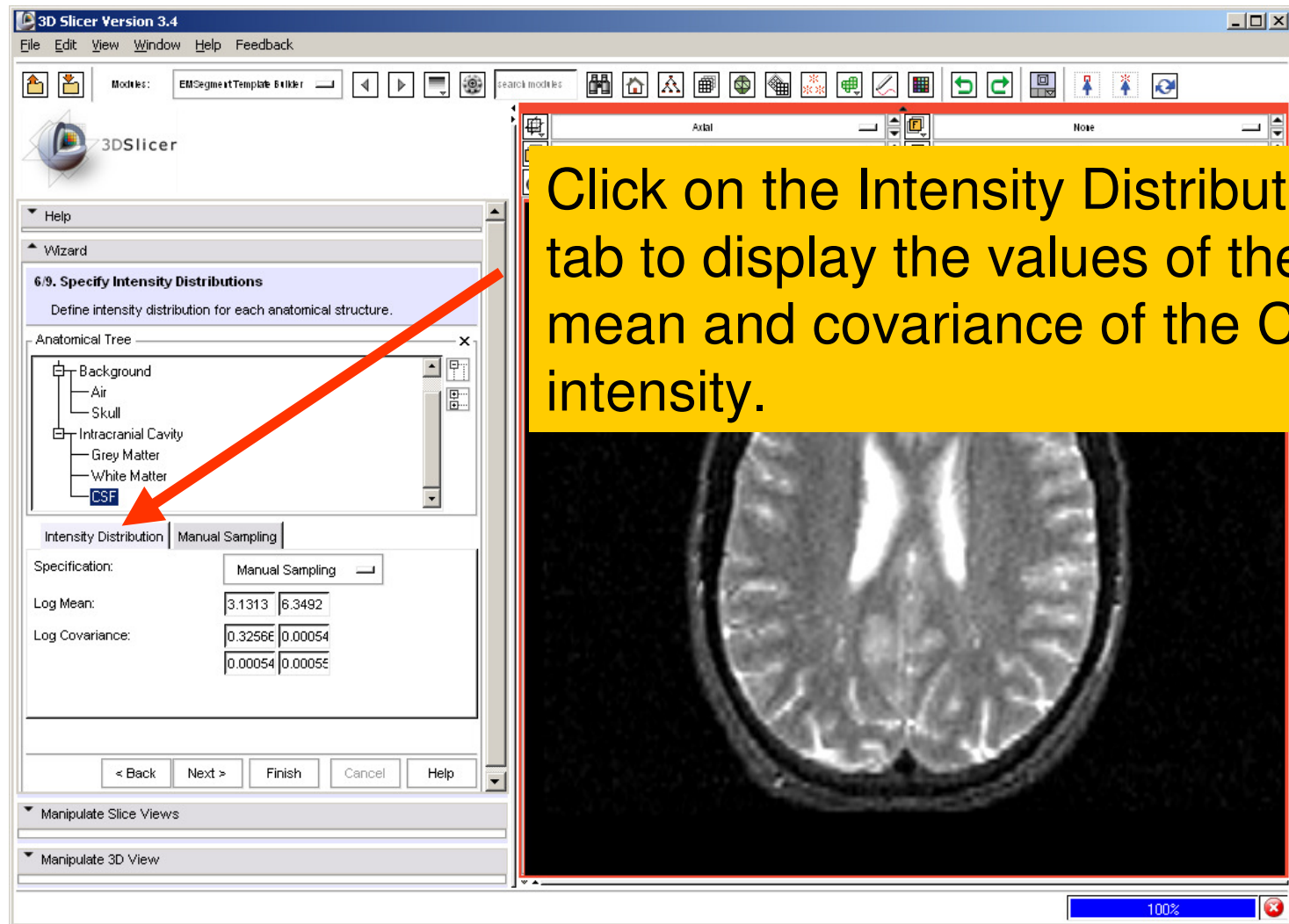
Select **CSF** from the anatomical tree.

Select **Manual Sampling** from the drop-down menu

Click on the **Manual Sampling tab** and sample 10 CSF voxels within the **t2** volume.

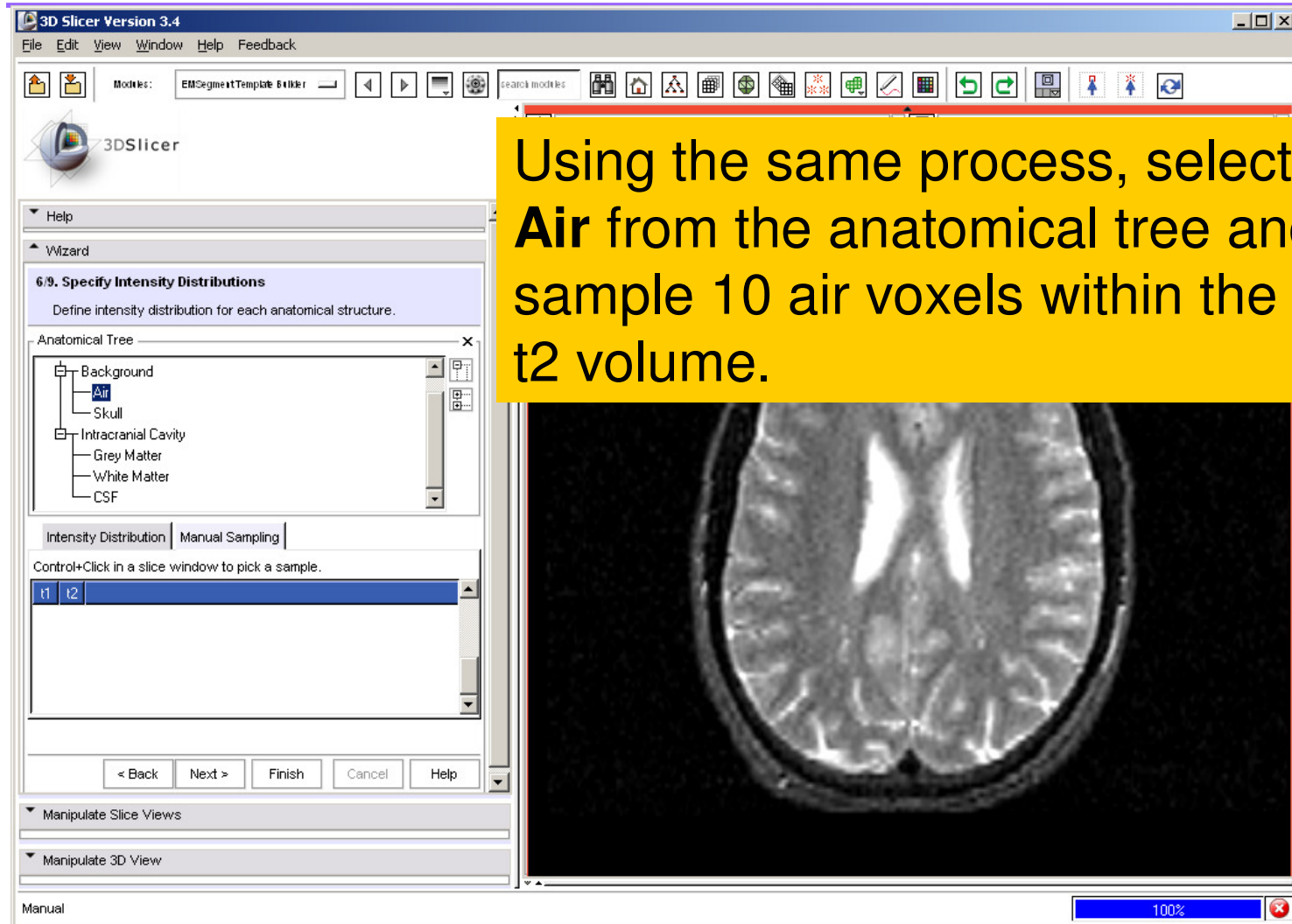
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Gaussian Intensity Distribution

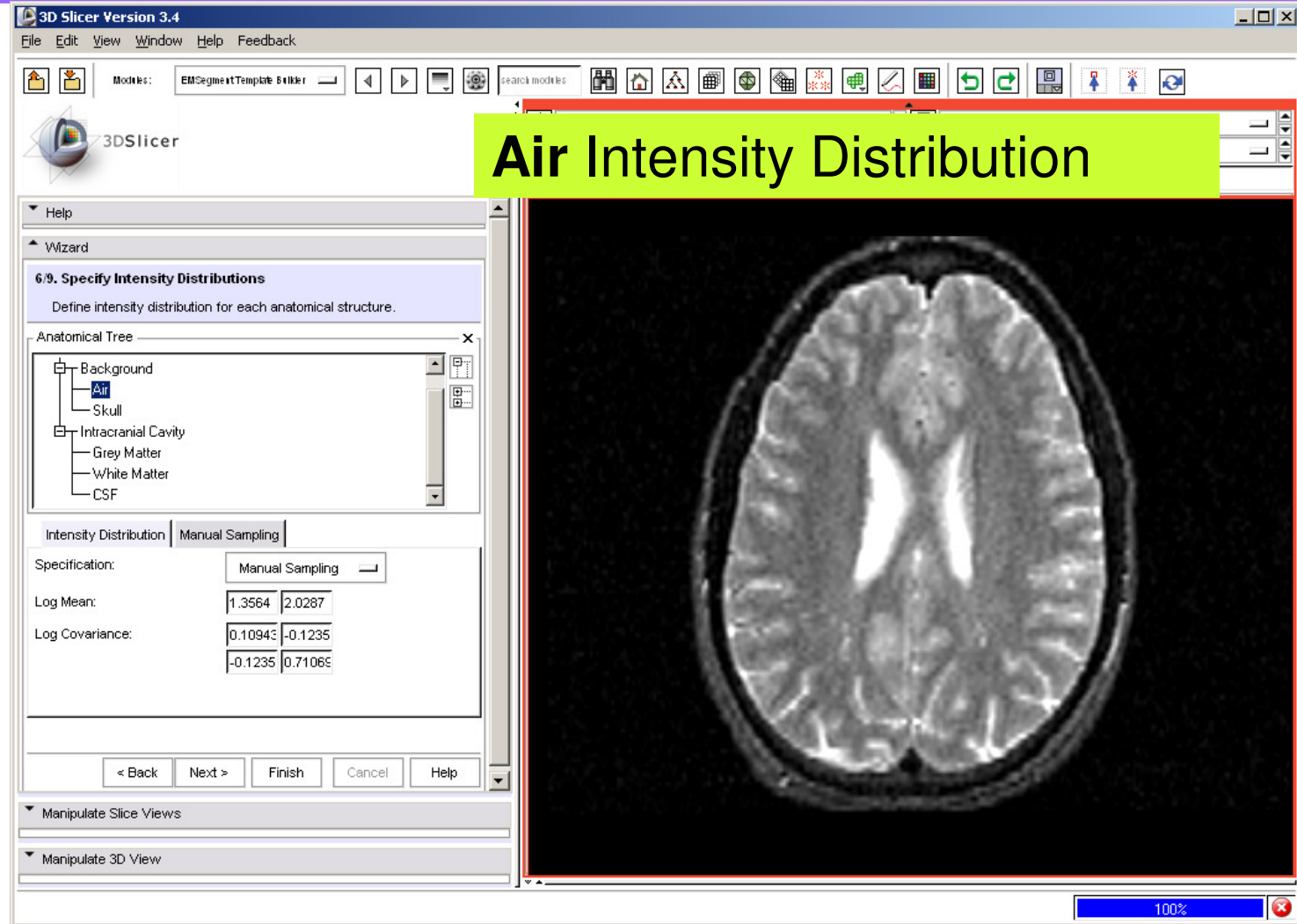


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Gaussian Intensity Distribution

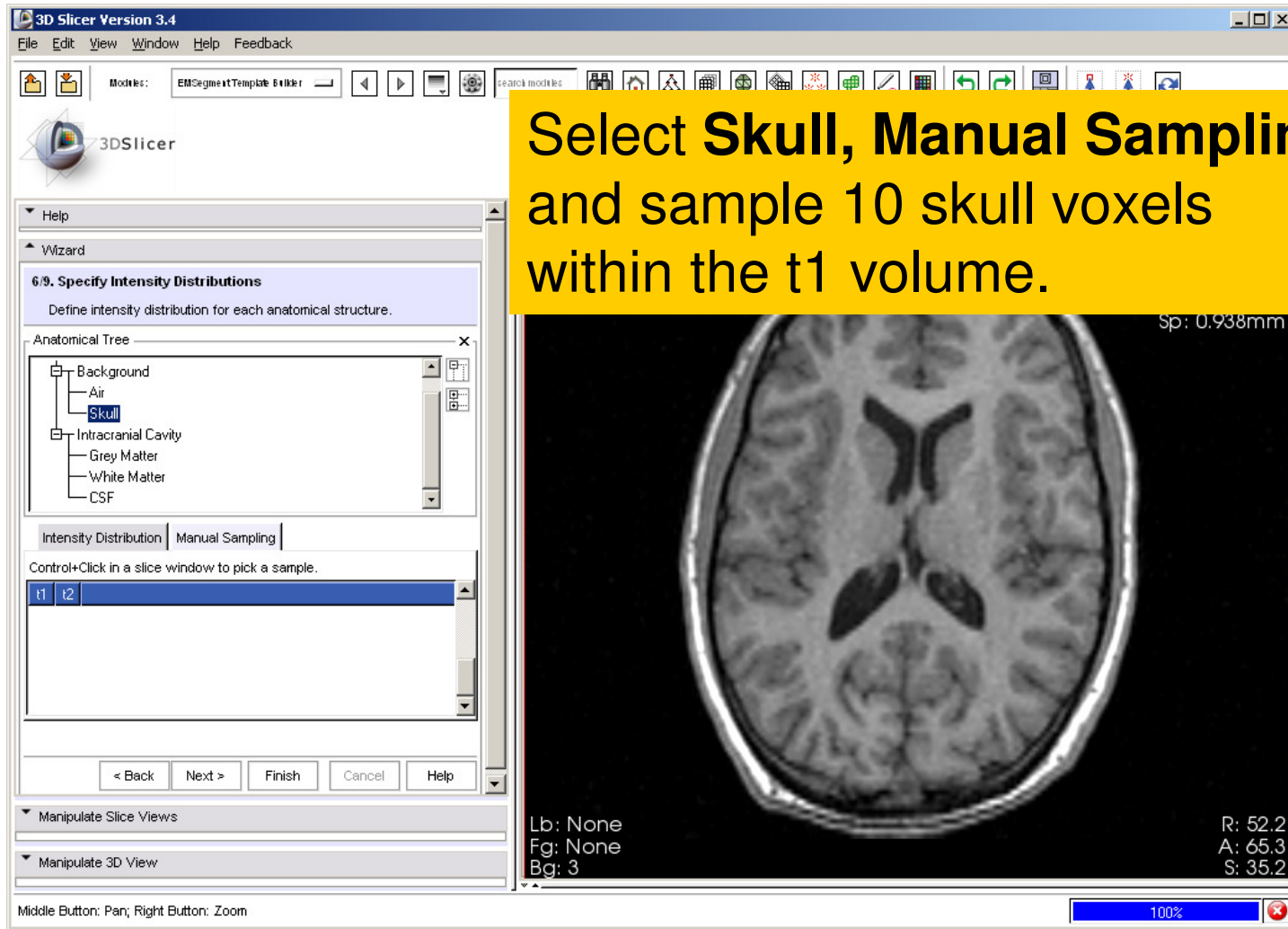


Gaussian Intensity Distribution



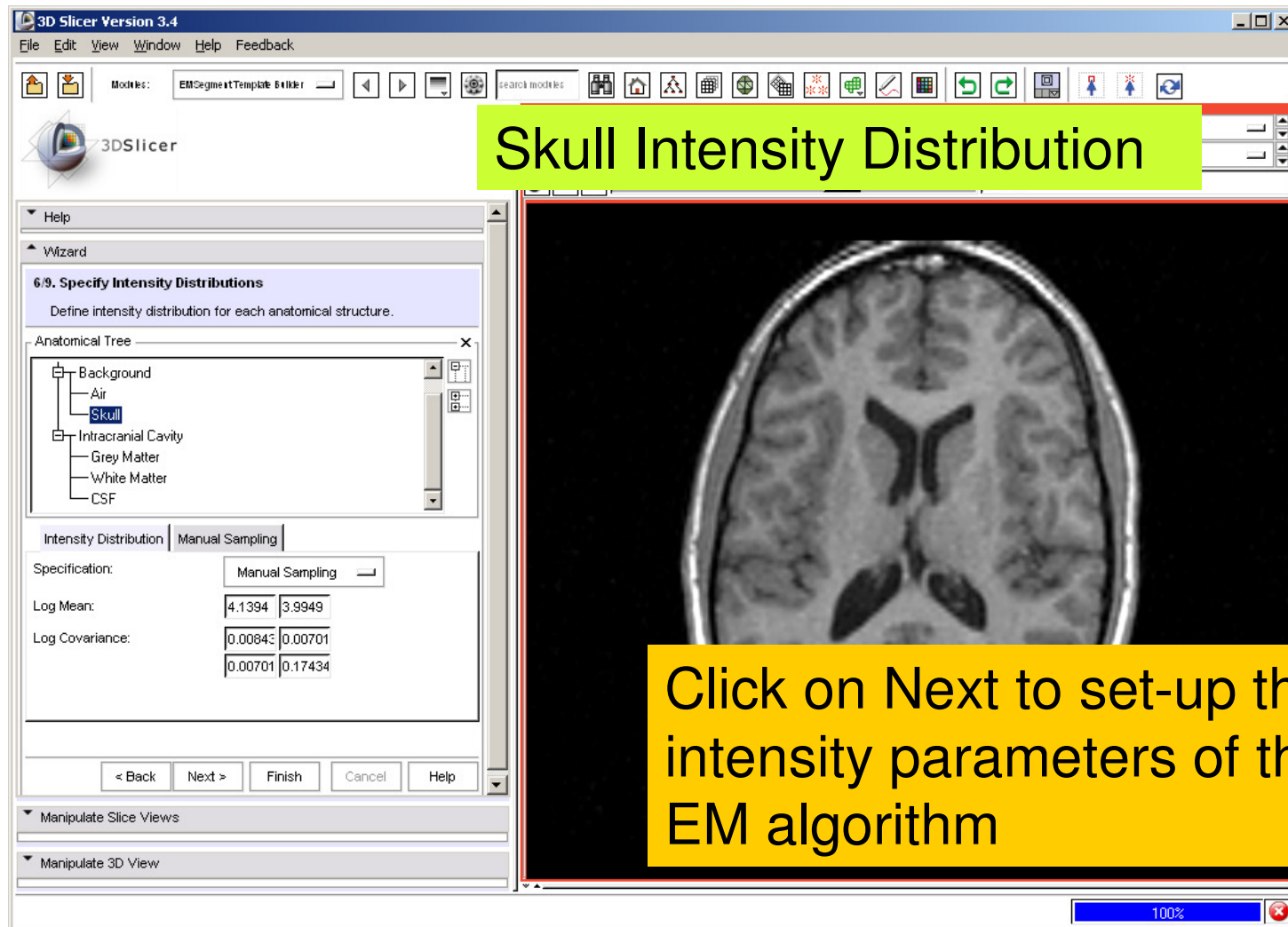
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Gaussian Intensity Distribution



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Gaussian Intensity Distribution



Skull Intensity Distribution

Click on Next to set-up the intensity parameters of the EM algorithm

6.9. Specify Intensity Distributions
Define intensity distribution for each anatomical structure.

Anatomical Tree

- Background
- Air
- Skull**
- Intracranial Cavity
- Grey Matter
- White Matter
- CSF

Intensity Distribution | Manual Sampling

Specification: Manual Sampling

Log Mean: 4.1394 3.9949

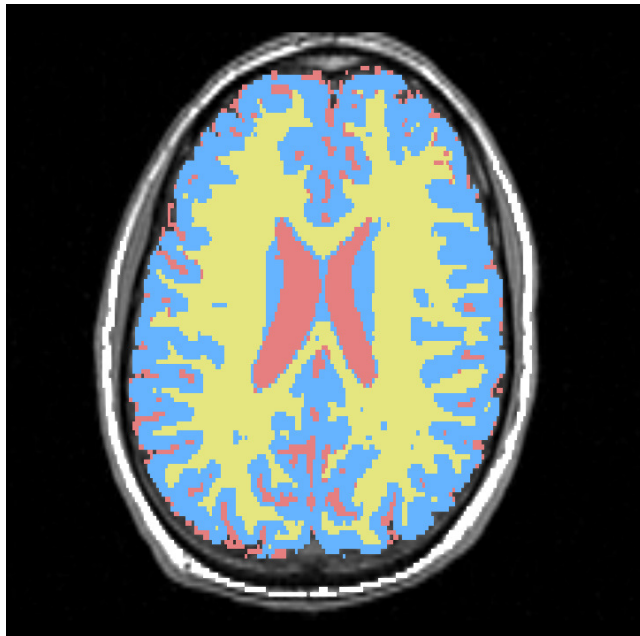
Log Covariance: 0.00843 0.00701
0.00701 0.17434

< Back Next > Finish Cancel Help

Manipulate Slice Views

Manipulate 3D View

100%

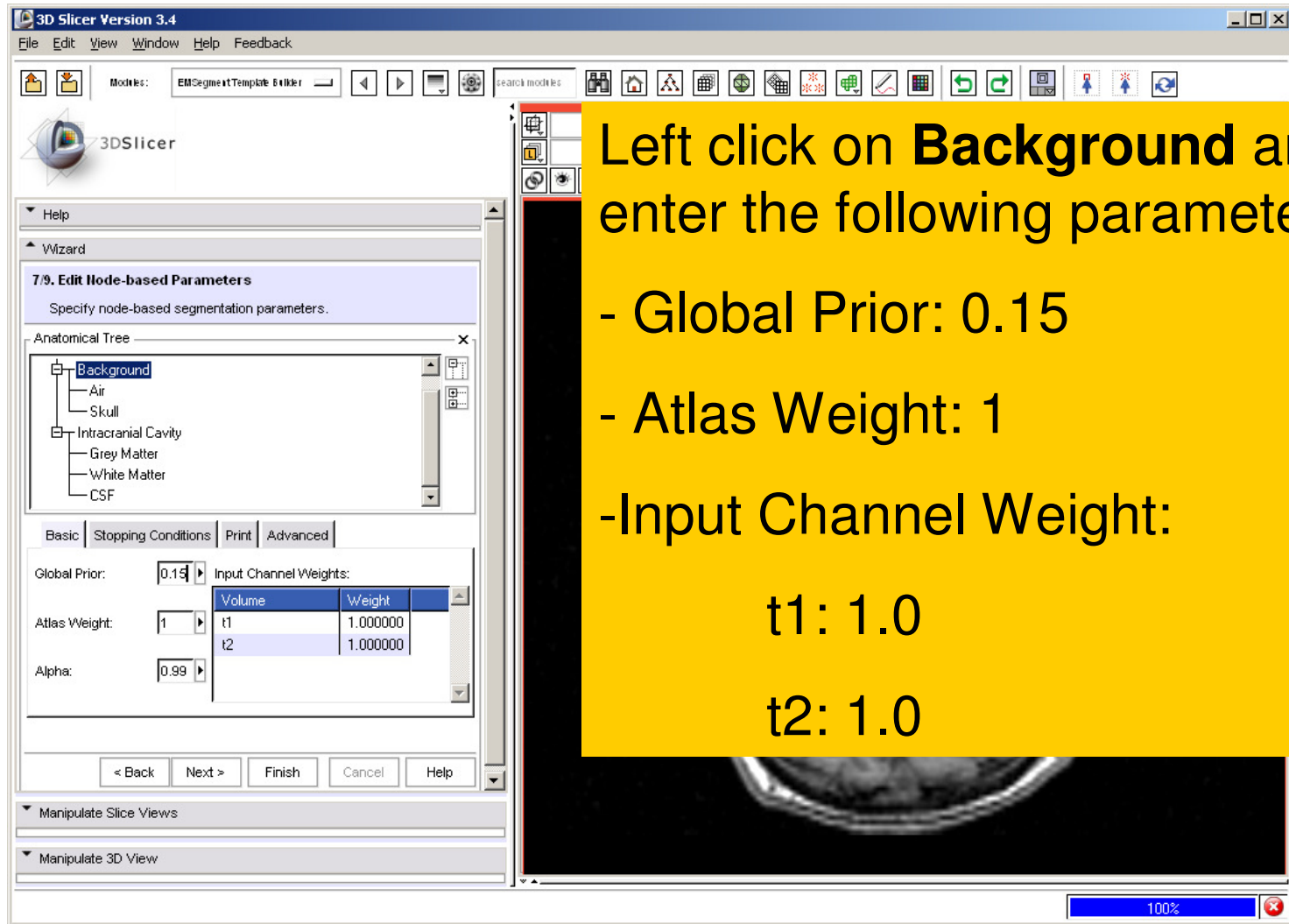


EM Input Parameters

EM Segmentation parameters

- Global Prior $p(L)$ → Global Prior Weight
- Gaussian Intensity Distribution $p(L|I)$ → Input Channel Weight
- Probabilistic Atlas $p(L|X)$ → Atlas Weight

Background Settings



Left click on **Background** and enter the following parameters:

- Global Prior: 0.15

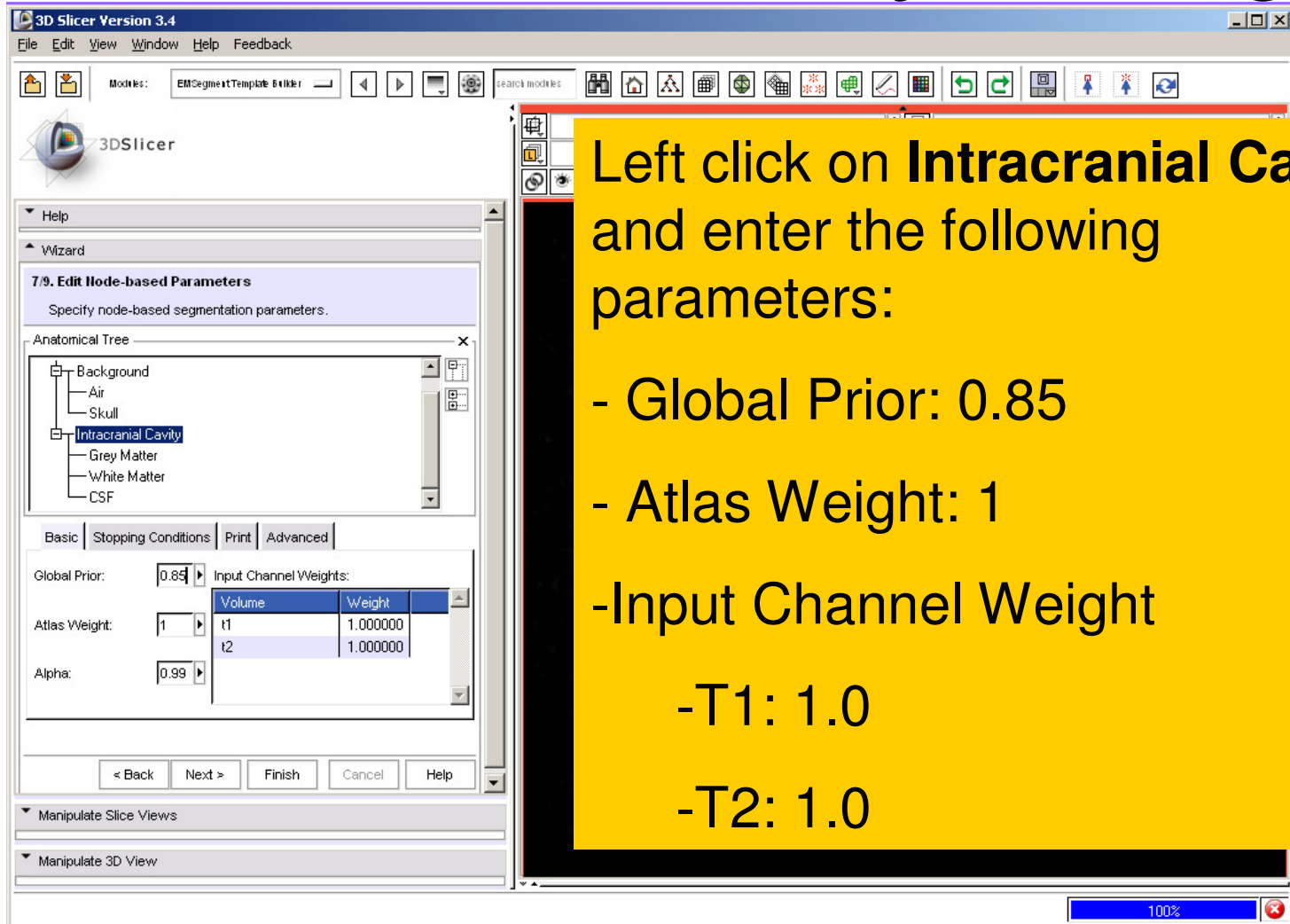
- Atlas Weight: 1

-Input Channel Weight:

t1: 1.0

t2: 1.0

Intracranial Cavity Settings



Left click on **Intracranial Cavity** and enter the following parameters:

- Global Prior: 0.85
- Atlas Weight: 1
- Input Channel Weight
 - T1: 1.0
 - T2: 1.0

Air and Skull settings

Air

Global Prior: $p(L) = 0.7$

Atlas Weight: $p(L|X) = 1$

Input Channel Weight: $t1, p(L|I) = 1.0$
 $t2, p(L|I) = 1.0$

Enter the following
parameters for Air and
Skull

Skull

Global Prior: $p(L) = 0.3$

Atlas Weight: $p(L|X) = 1$

Input Channel Weight: $t1, p(L|I) = 1.0$
 $t2, p(L|I) = 1.0$

Intracranial Cavity

GM

Global Prior: $p(L) = 0.45$

Atlas Weight: $p(L|X) = 0.01$

Input Channel Weight: $t1, p(L|I) = 1.0$
 $t2, p(L|I) = 0.1$

Enter the following parameters for GM, WM and CSF.

WM

Global Prior: $p(L) = 0.3$

Atlas Weight: $p(L|X) = 0.7$

Input Channel Weight: $t1, p(L|I) = 0.95$
 $t2, p(L|I) = 0.05$

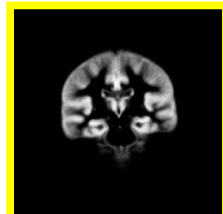
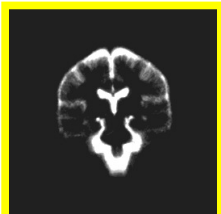
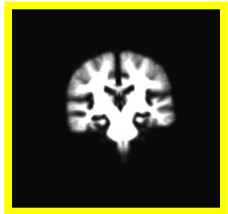
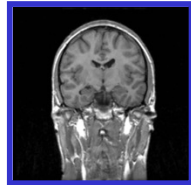
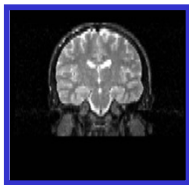
CSF

Global Prior: $p(L) = 0.25$

Atlas Weight: $p(L|X) = 0.01$

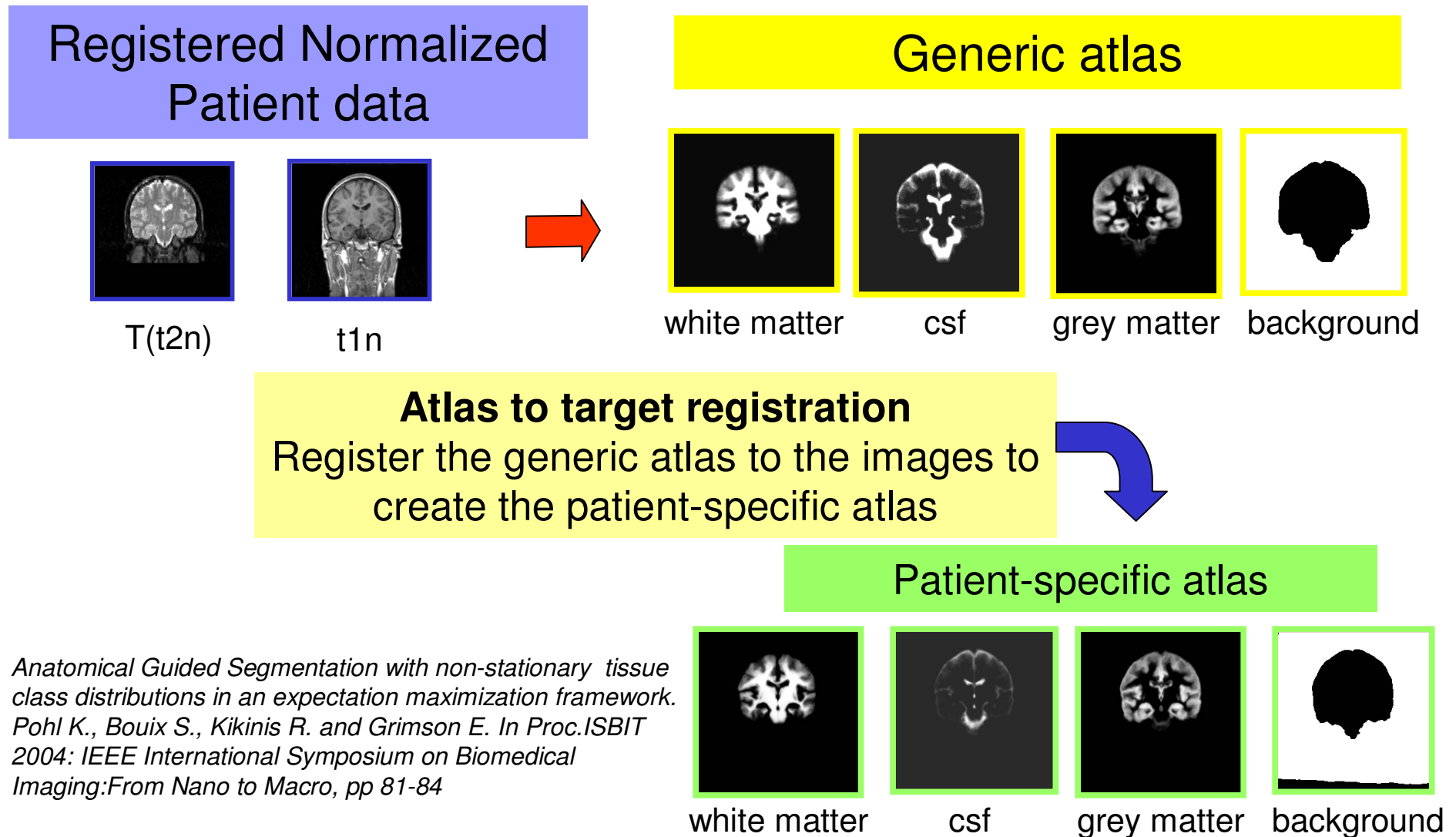
Input Channel Weight: $t1, p(L|I) = 0.1$
 $t2, p(L|I) = 1.0$

Click on **Next**.



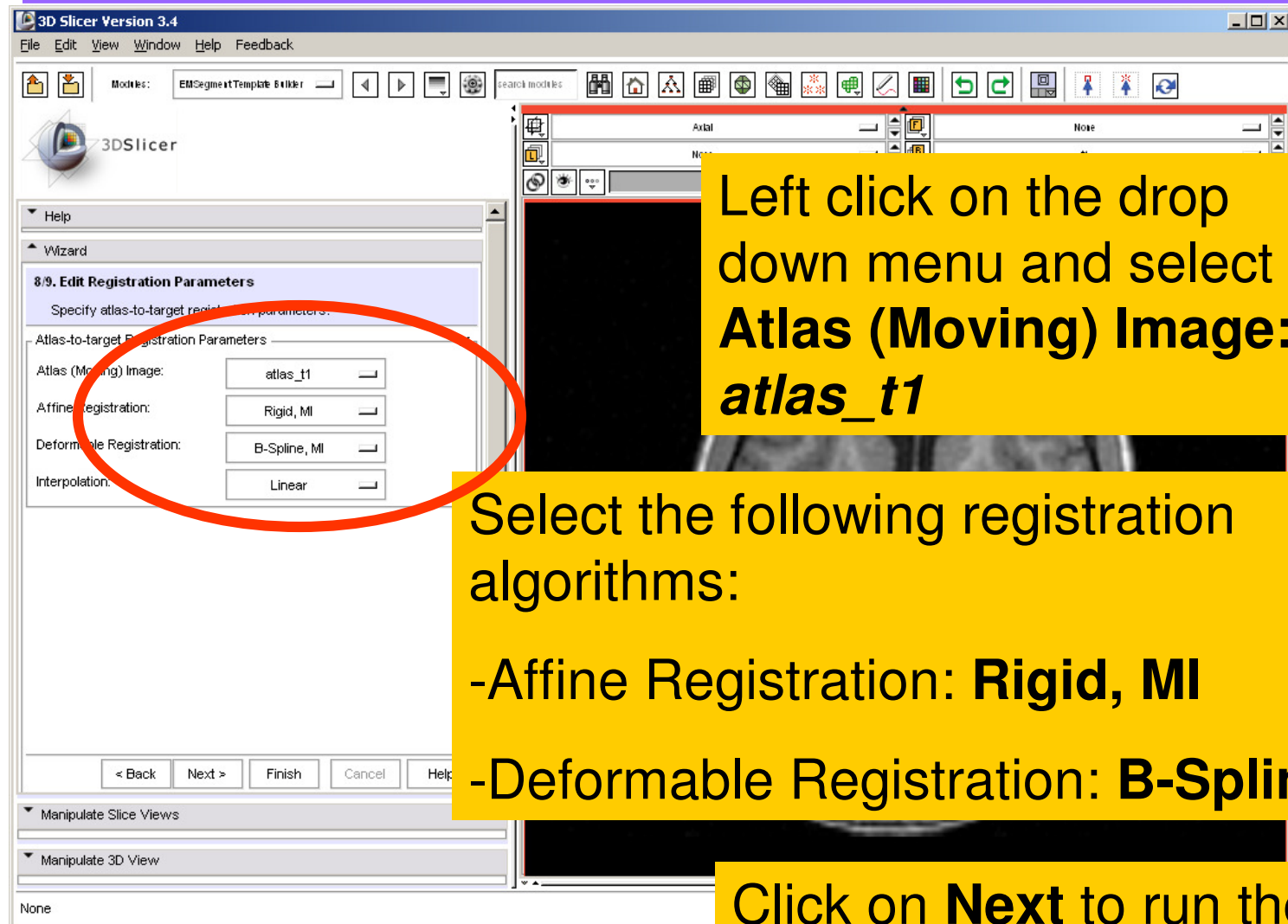
Atlas To Target Registration

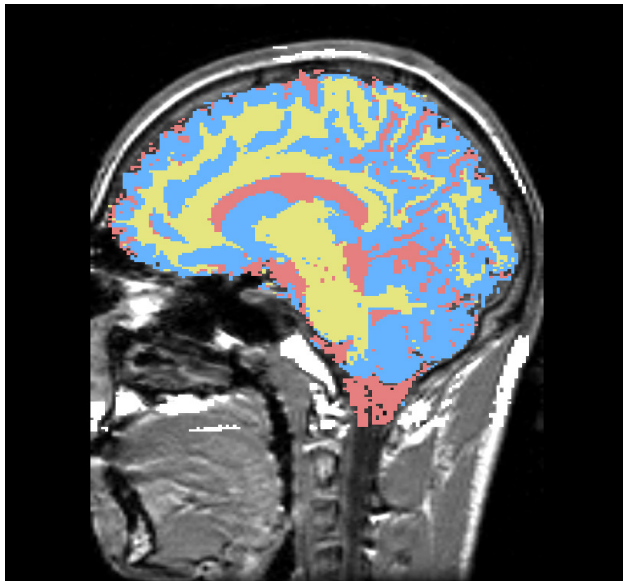
EM Pipeline: Patient-Specific Atlas Generation



Anatomical Guided Segmentation with non-stationary tissue class distributions in an expectation maximization framework. Pohl K., Bouix S., Kikinis R. and Grimson E. In Proc.ISBIT 2004: IEEE International Symposium on Biomedical Imaging:From Nano to Macro, pp 81-84

Atlas To Target Registration

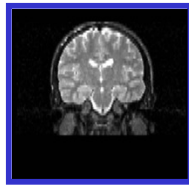




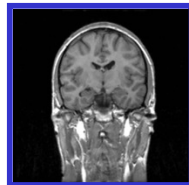
Segmentation

EM Pipeline: Segmentation

Normalized
Patient data



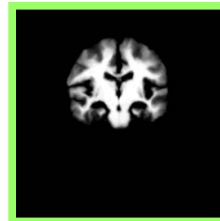
T(t2)
normalized



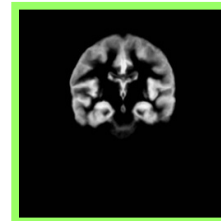
T1
normalized



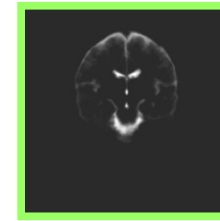
Patient-specific atlas



white matter



csf



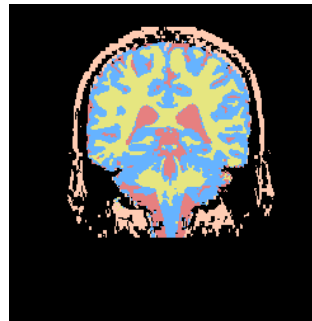
grey matter



background

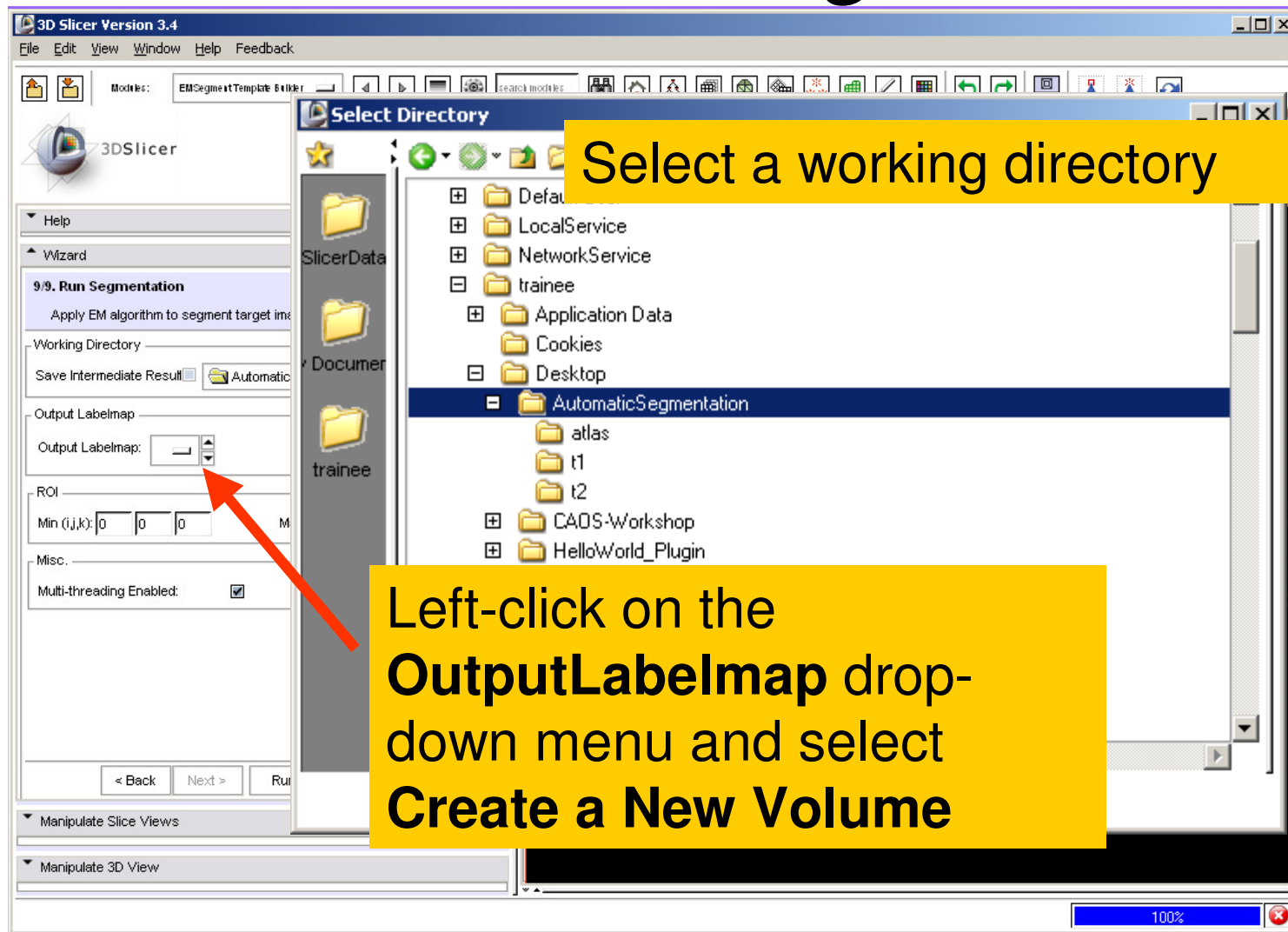


Anatomical Guided Segmentation with non-stationary tissue class distributions in an expectation maximization framework. Pohl K., Bouix S., Kikinis R. and Grimson E. In Proc.ISBIT 2004: IEEE International Symposium on Biomedical Imaging: From Nano to Macro, pp 81-84



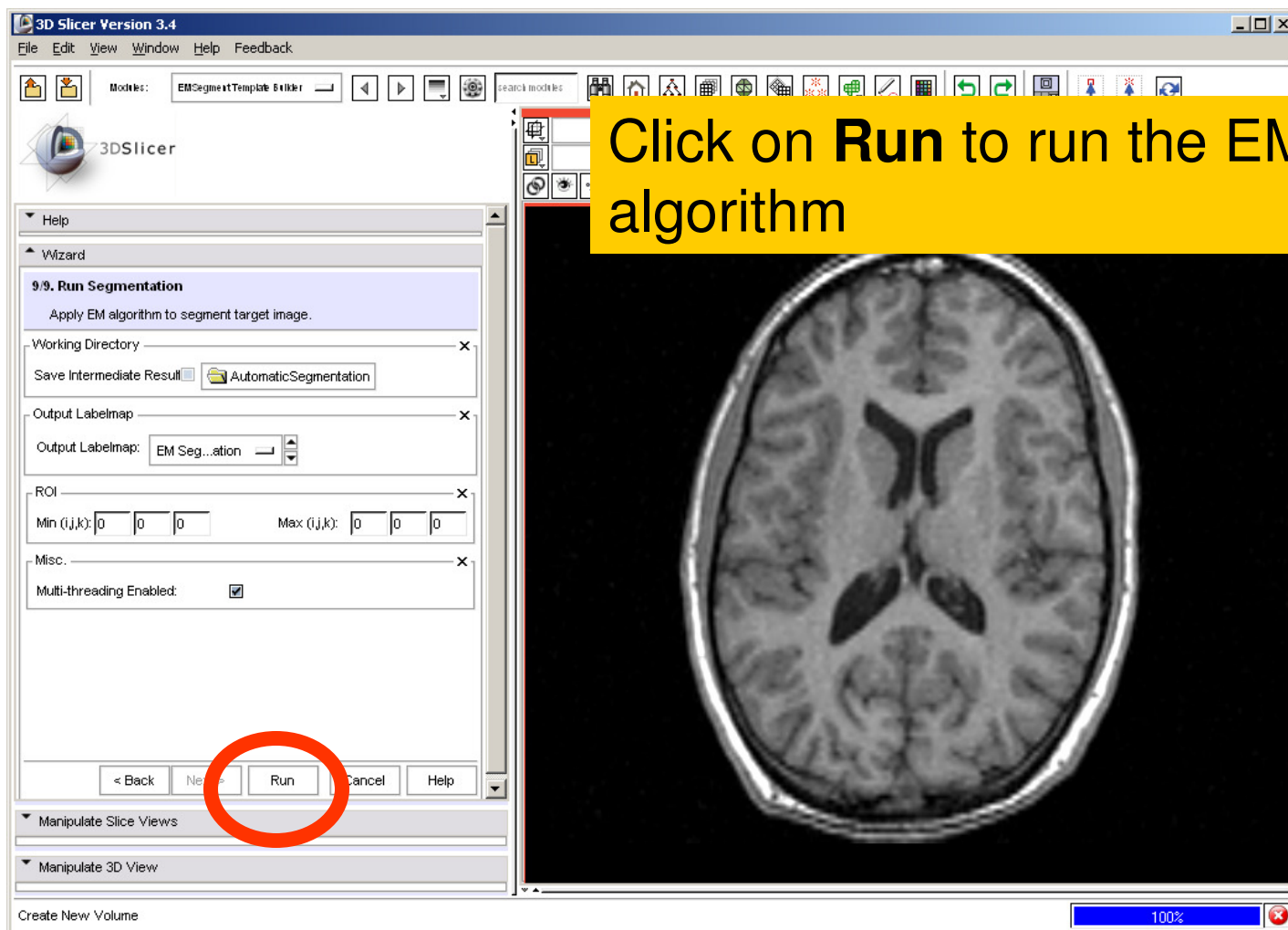
Segment using
the Expectation
Maximization
algorithm

Segmentation



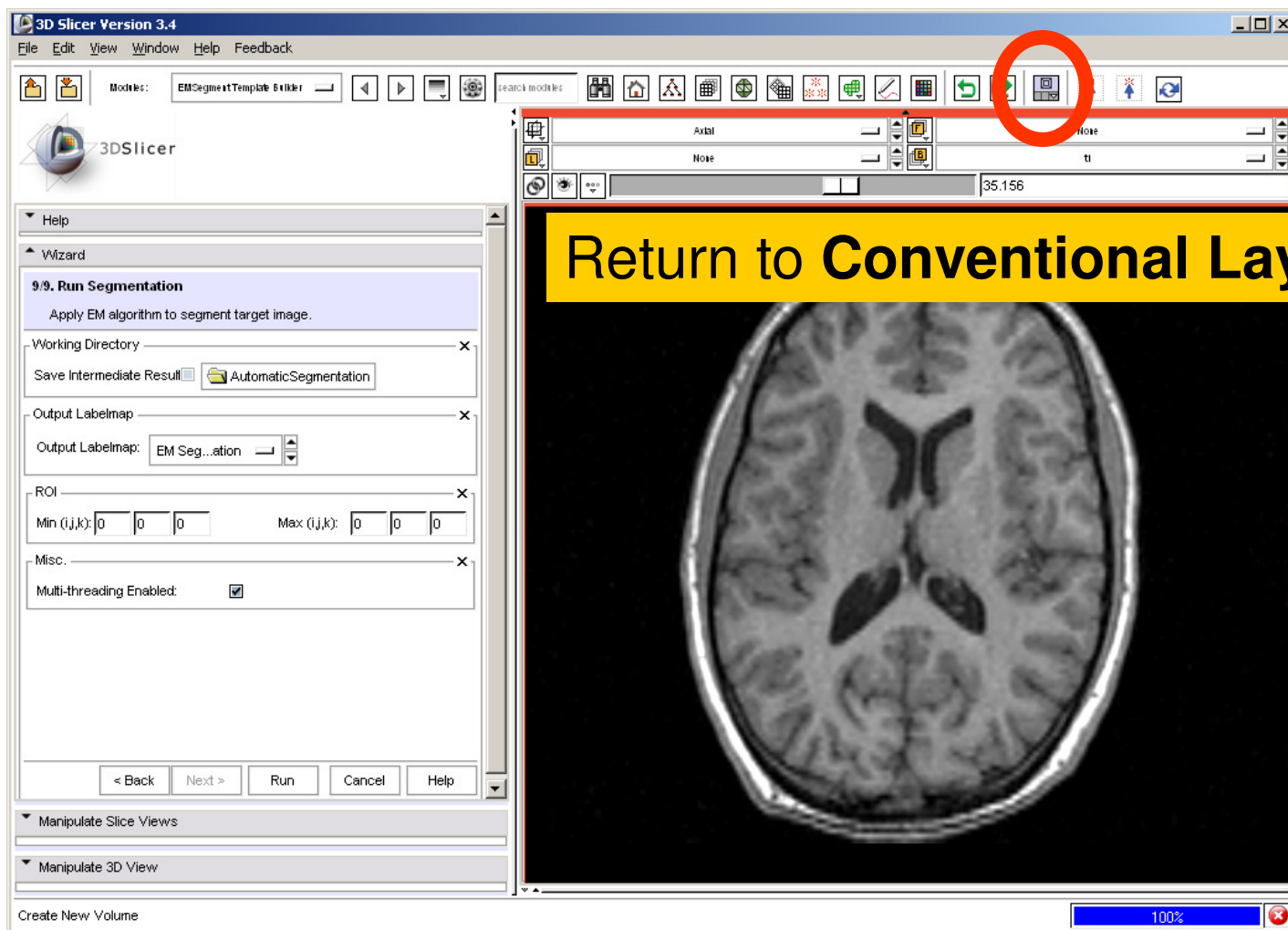
*Automatic Segmentation. Sonia Pujol, Ph.D., Harvard Medical School
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Segmentation



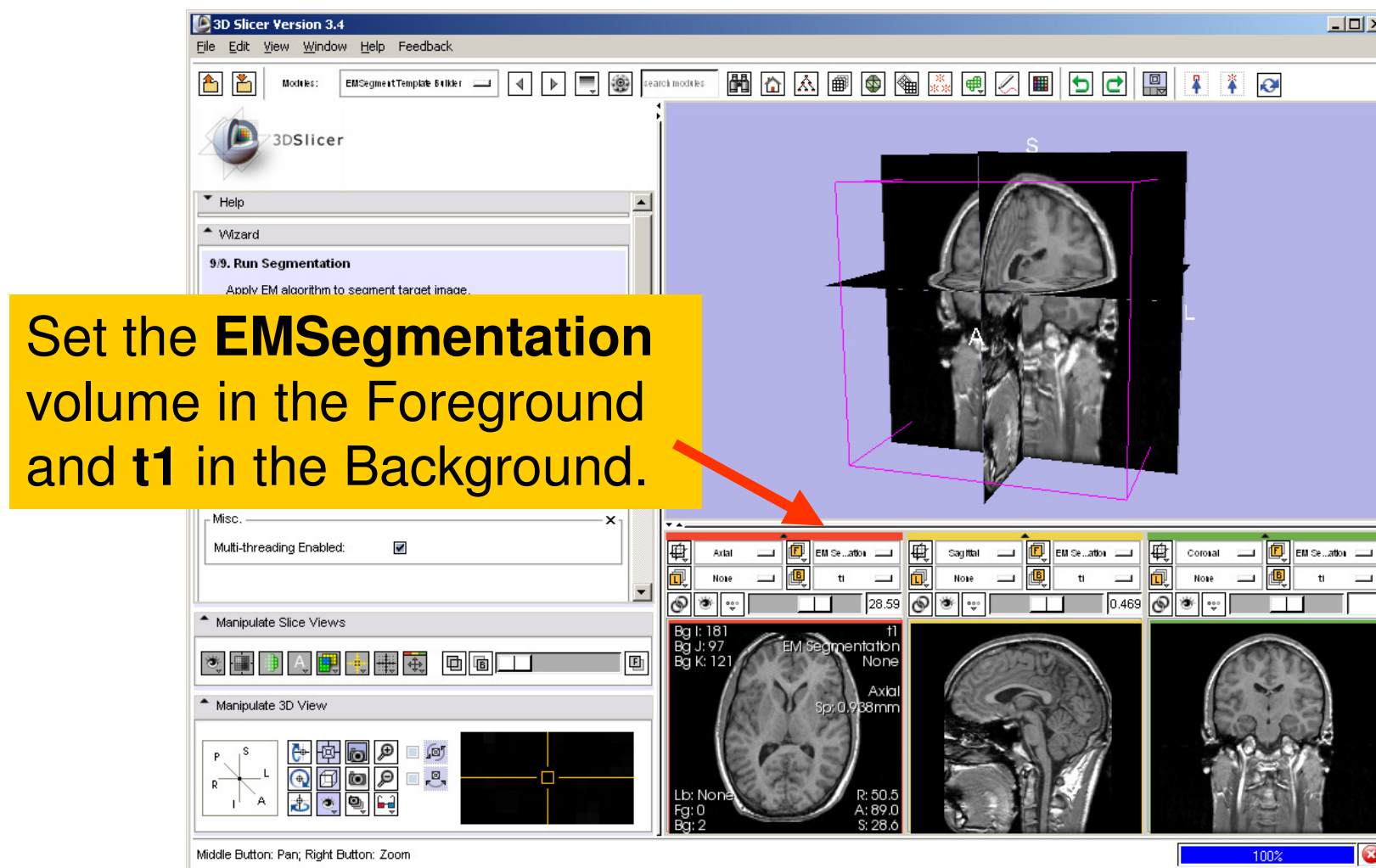
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Segmentation



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National Alliance for Medical Image Computing*

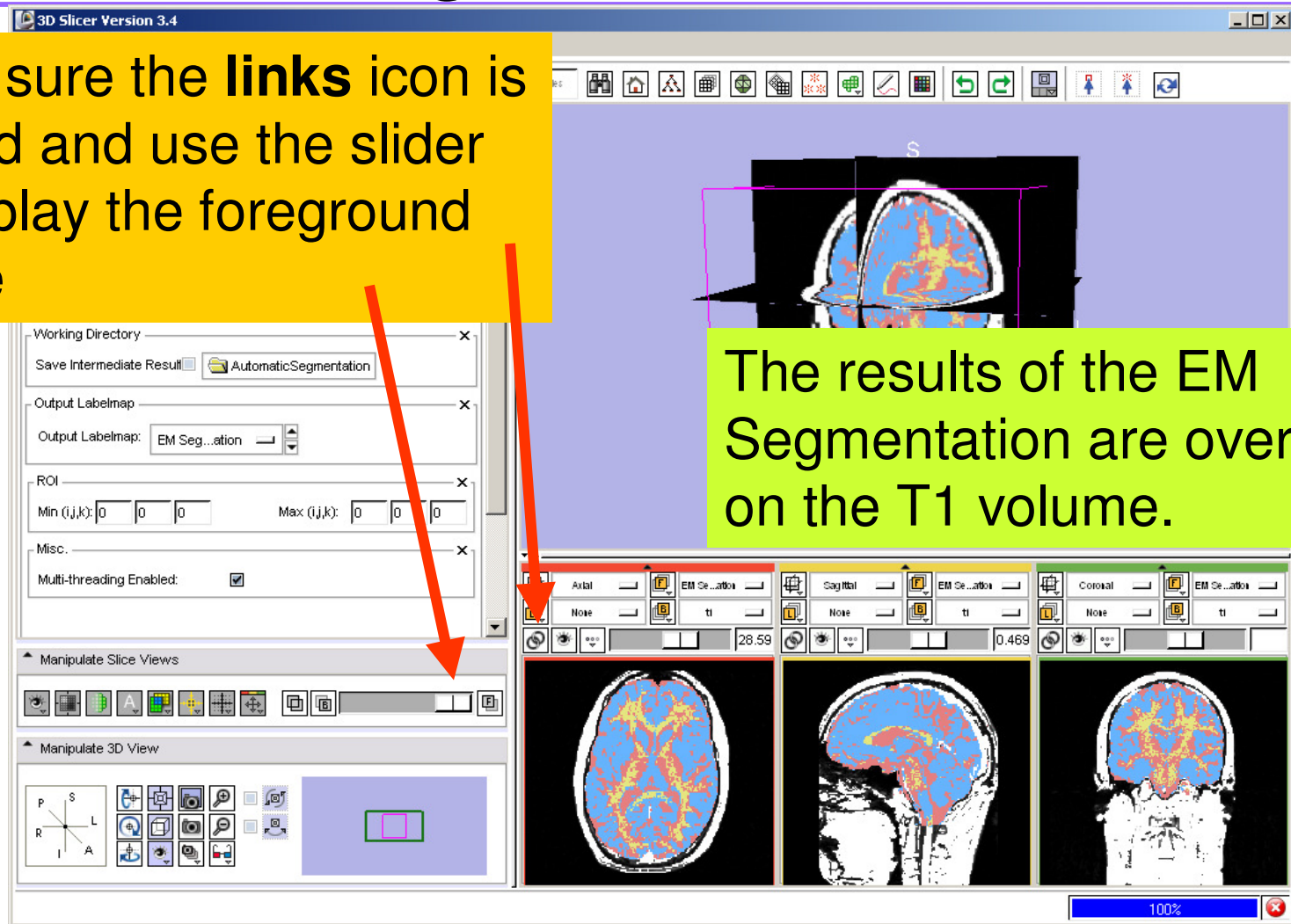
Segmentation Results



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Segmentation Results

Make sure the **links** icon is clicked and use the slider to display the foreground image



The results of the EM Segmentation are overlaid on the T1 volume.

Acknowledgments



National Alliance for Medical Image Computing

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Neuroimage Analysis Center

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