

# MONAI<sup>+</sup>Label

Andres Diaz-Pinto on behalf of the MONAI Label Team  
NVIDIA & King's College London  
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<https://github.com/Project-MONAI/MONAILabel>

# Outline:

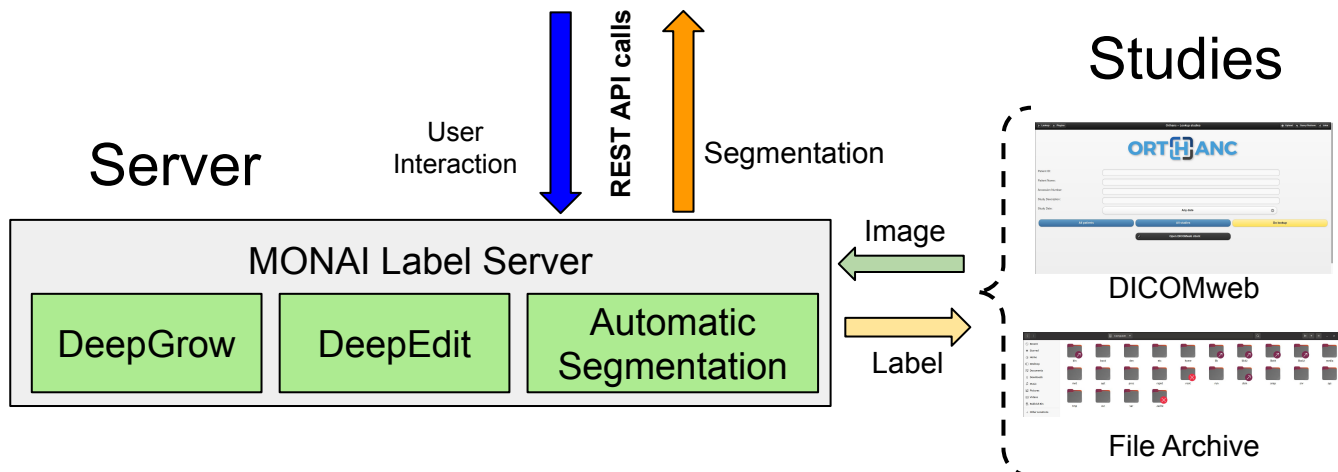
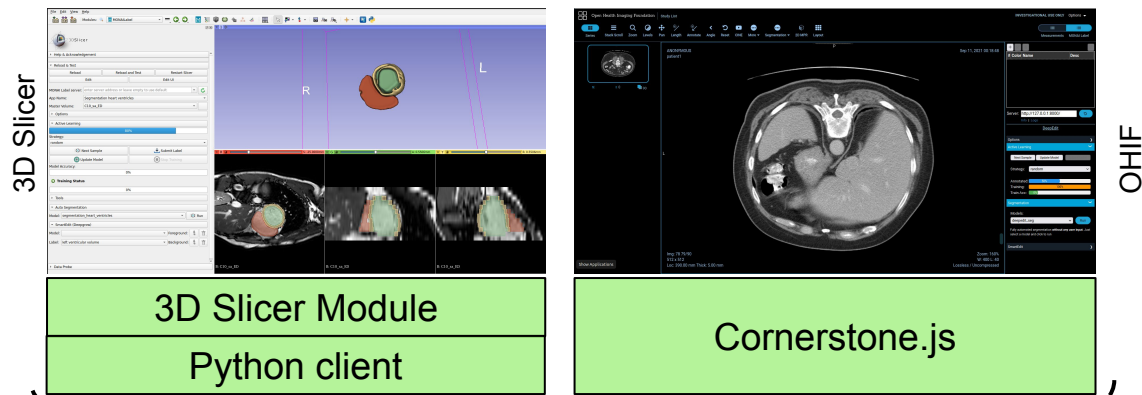
- What is MONAI Label?
- Why use MONAI Label?
- How to create a MONAI Label App?
  - Datastore in file archive
  - Heuristic Planner
  - Active Learning Strategies
- Scribbles in MONAILabel
- Create your customised Slicer Module
- OHIF + MONAI Label
- Demo

# What is MONAI Label?

- An intelligent **open source** image labeling and learning tool that enables users to create annotated datasets and build AI annotation models for clinical evaluation
- Framework for developing and deploying **MONAI Label Apps** to train and do inference **using** Deep Learning models
- MONAI Label is one of the first frameworks that introduces **Active learning strategies in a software annotation setting**
- It is all Python and can be installed with simple “**pip install monailabel**”
- Supported viewers:
  - **3D Slicer**
  - **Open Health Imaging Foundation (OHIF)**

# MONAI Label Infrastructure: server-client system

## Clients



# MONAI Label Infrastructure: REST API calls

AppService		
GET	/info/	Get App Info
GET	/download/{image}	Download Image
Infer		
POST	/infer/{model}	Run Inference for supported model
GET	/batch/infer	Get Status of Batch Inference Task
DELETE	/batch/infer	Stop Batch Inference Task
POST	/batch/infer/{model}	Run Batch Inference Task
Train		
GET	/train/	Get Status of Training Task
POST	/train/	Run Training Task
DELETE	/train/	Stop Training Task

# Why use MONAI Label?

- **Researcher Perspective:** MONAI Label allows researchers to
  - Create new annotation methods
  - Rapid App prototyping
  - Implement active learning techniques
  - Verify their effectiveness in real-world scenarios
  - Make incremental improvements
  - Readily deploy labeling apps to wider audiences
- **Clinician Perspective:** MONAI Label reduces the time and effort of annotating new datasets
  - Ready-to-use **3DSlicer**
  - Pre-built plugin for **OHIF Viewer**

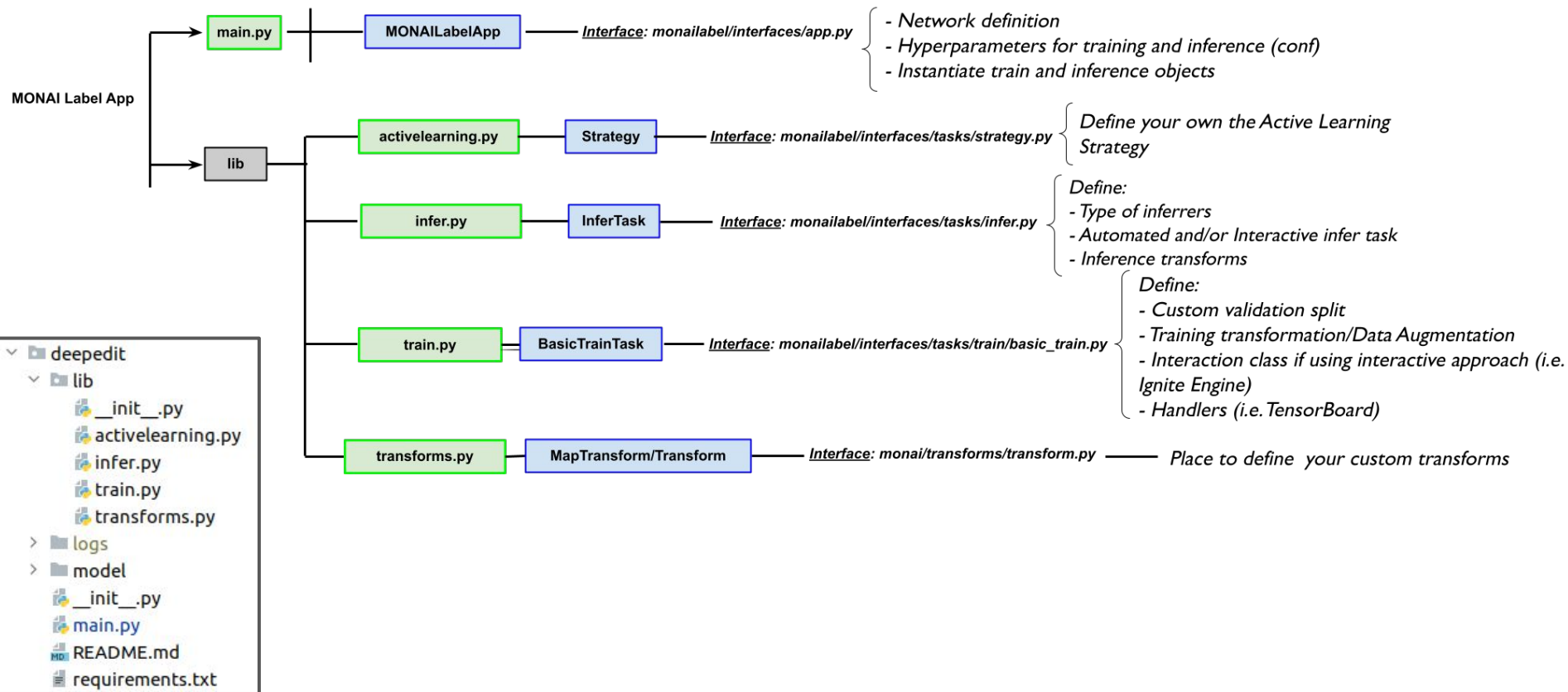
# How to create a MONAI Label App?

- Select the spatial/intensity transforms to preprocess images for training and inference
- Define the active learning technique use in the labeling app
- Define neural network architecture
- Preprocess scribbles, points, ROI, closed curve, or any input sent to the MONAI Label server through the REST API

Researchers can also use [sample apps](#) (i.e. DeepGrow, DeepEdit and UNet) to jumpstart the development of their own custom labeling apps

- For more details check out [our tutorial](#)

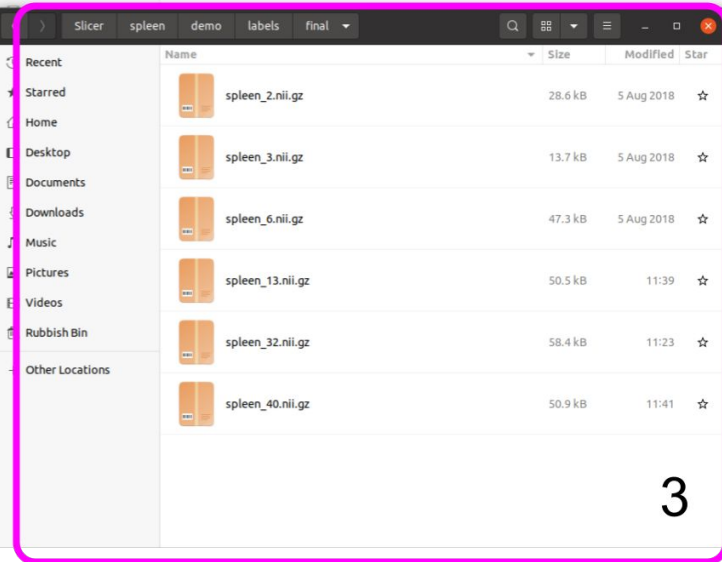
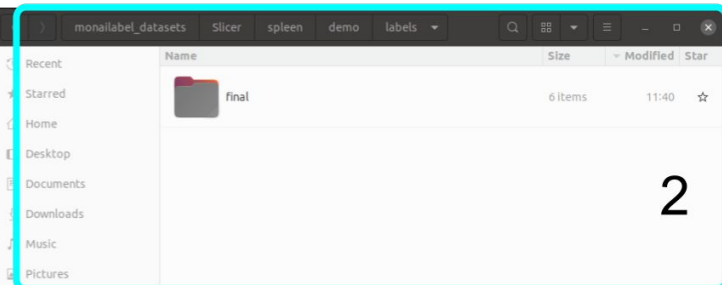
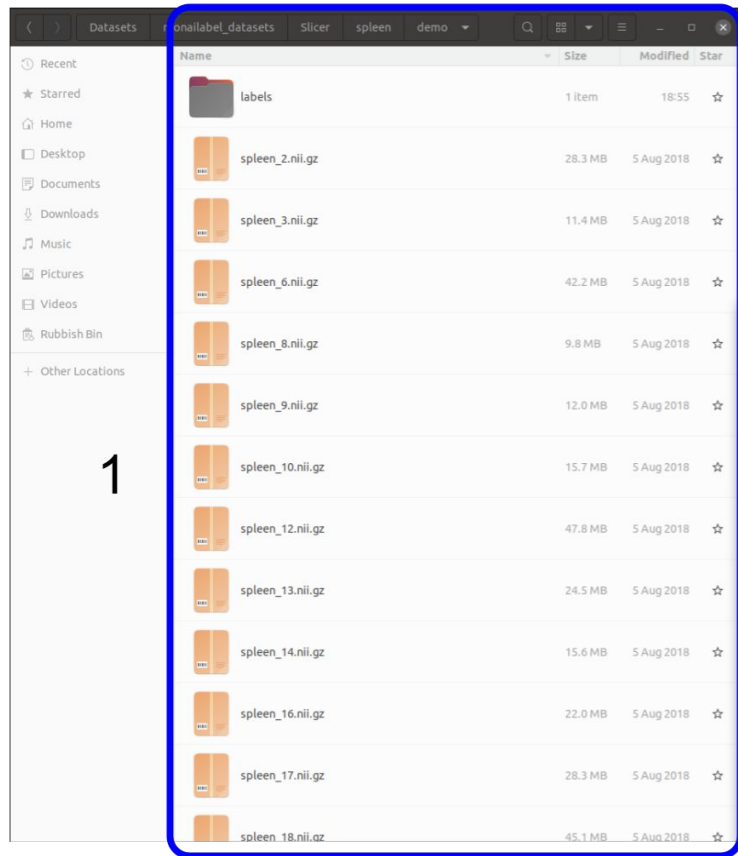
# MONAI Label App Structure





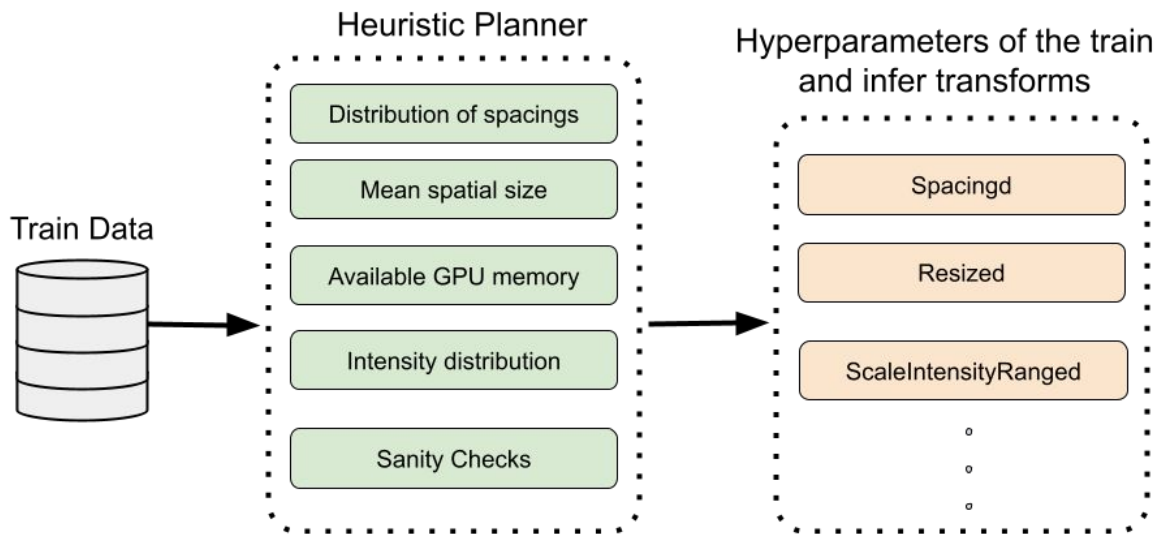
# MONAI Label App: Datastore in file archive

```
labels
├── final
│   ├── spleen_13.nii.gz
│   ├── spleen_2.nii.gz
│   ├── spleen_32.nii.gz
│   ├── spleen_3.nii.gz
│   ├── spleen_40.nii.gz
│   └── spleen_6.nii.gz
├── spleen_10.nii.gz
├── spleen_12.nii.gz
├── spleen_13.nii.gz
├── spleen_14.nii.gz
├── spleen_16.nii.gz
├── spleen_17.nii.gz
├── spleen_18.nii.gz
├── spleen_19.nii.gz
├── spleen_20.nii.gz
├── spleen_21.nii.gz
├── spleen_22.nii.gz
├── spleen_24.nii.gz
├── spleen_25.nii.gz
├── spleen_26.nii.gz
├── spleen_27.nii.gz
├── spleen_28.nii.gz
├── spleen_29.nii.gz
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├── spleen_33.nii.gz
├── spleen_38.nii.gz
├── spleen_3.nii.gz
├── spleen_40.nii.gz
├── spleen_41.nii.gz
├── spleen_44.nii.gz
├── spleen_45.nii.gz
├── spleen_46.nii.gz
├── spleen_47.nii.gz
├── spleen_6.nii.gz
├── spleen_8.nii.gz
└── spleen_9.nii.gz
```

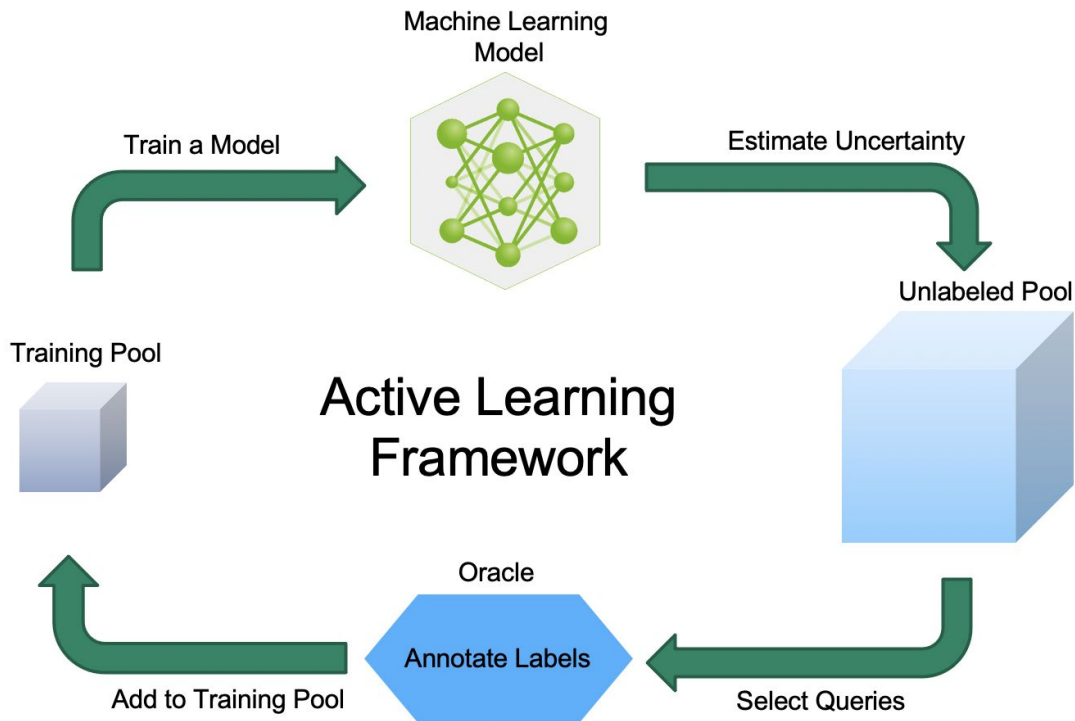


# MONAI Label App: *Heuristic planner*

- Considers available GPU memory to define image spatial size.
- Defines training transforms based on GPU memory, average spatial size and spacing of datastore.
- Performs sanity checks before starting training.
- Shows warning in case images are multimodality or multilabel.

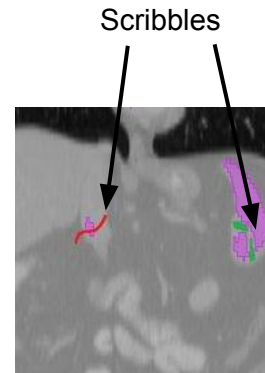


# MONAI Label App: *Active Learning Strategies*

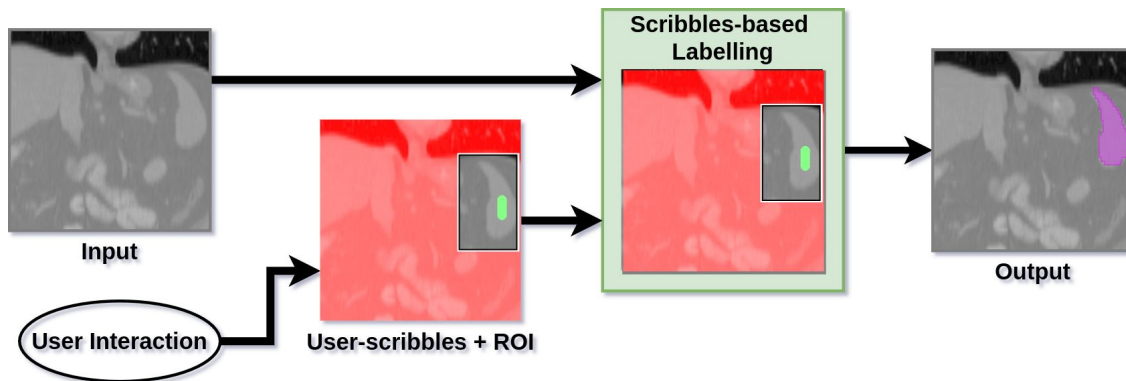


- **Active learning** is a semi-supervised machine learning approach where the algorithm can choose which data it wants to learn from
- Initially, random images are selected to be labeled  
However, random is not always the most efficient.
- After having a pretrained model, uncertainty of each image can be computed. Unlabeled samples that need more attention from the clinician will be selected
- Selection of harder samples or samples that need more attention

# Scribbles in MONAILabel



- Scribbles → free-hand line drawings for minimal interaction
- MONAILabel provides two scribbles-based modes:
  - **Scribbles-only**: uses scribbles to generate segmentation labels (demo) [1, 2]
  - **Scribbles-based refinement**: refines labels from a deep learning model [2]



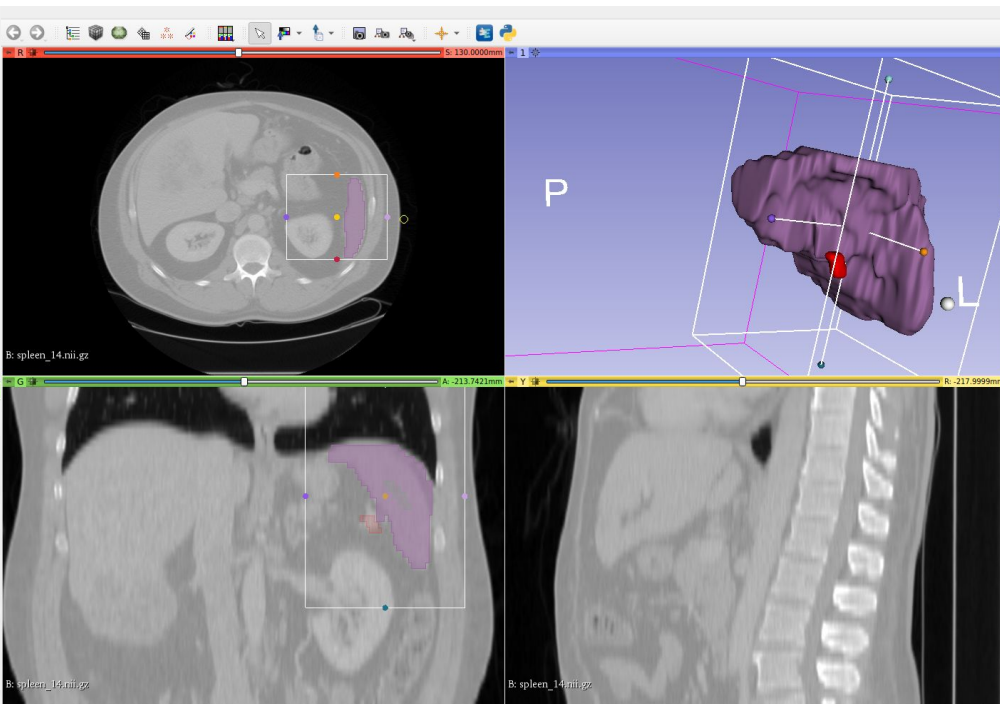
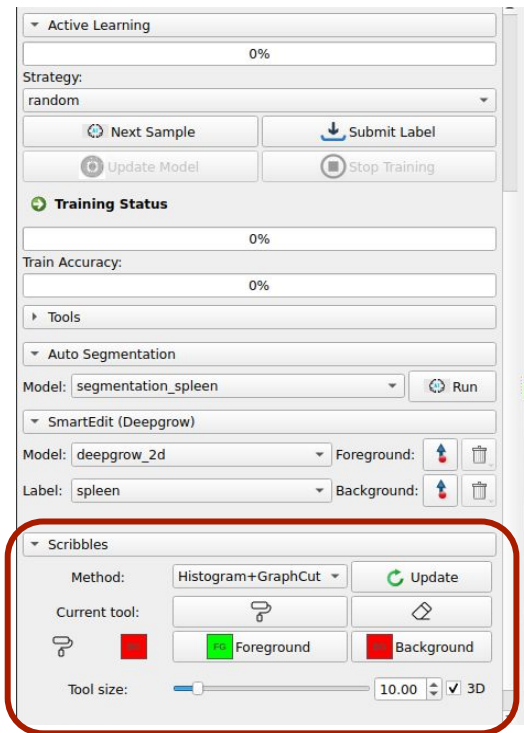
- 3D ROI Bounding box method to efficiently do background selection

[1] Criminisi, Antonio, et al. "Geos: Geodesic image segmentation." ECCV, 2008.

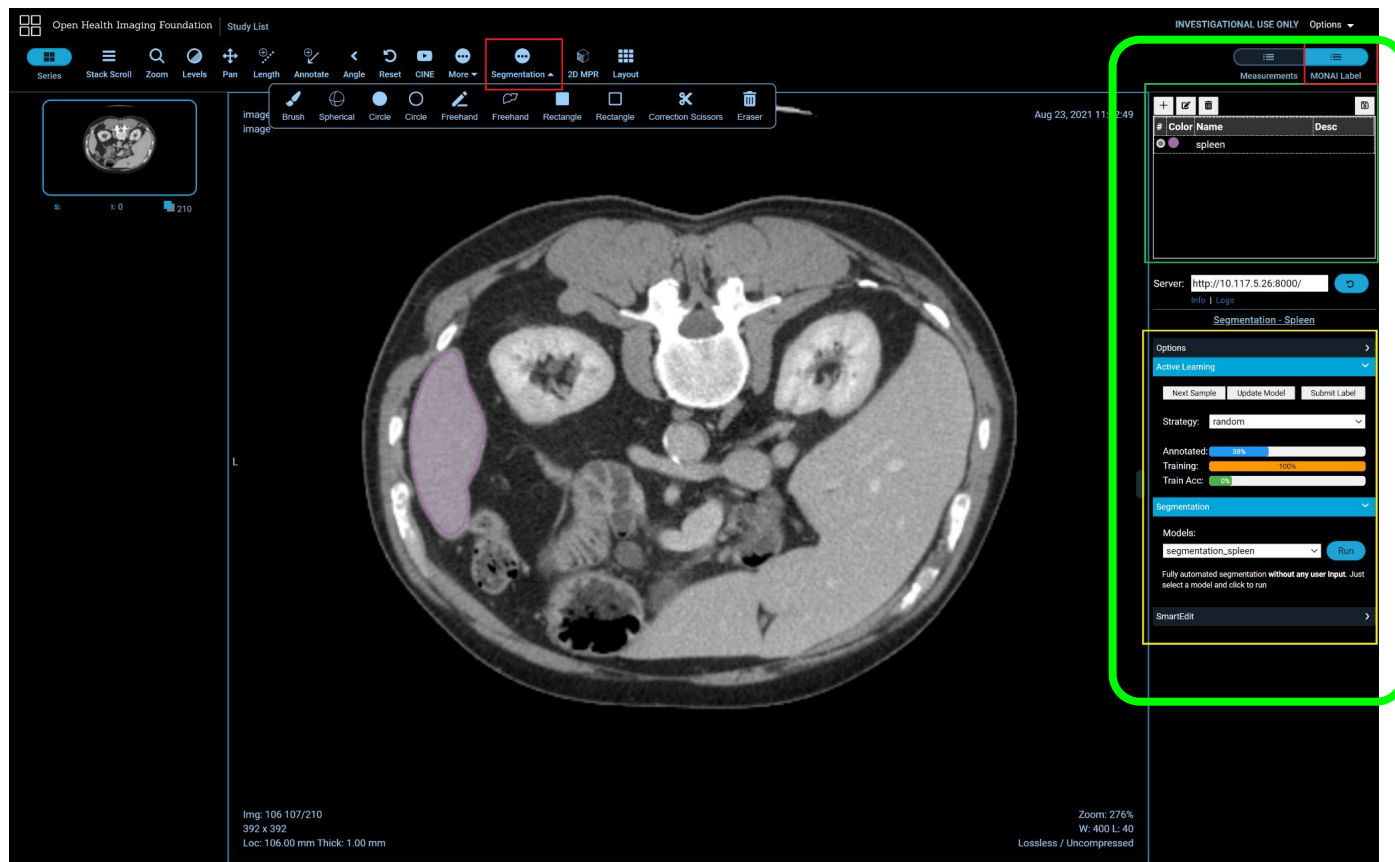
[2] Wang, Guotai, et al. "Interactive medical image segmentation using deep learning with image-specific fine tuning." IEEE TMI, 2018.

# Create your customised Slicer Module

*More dynamic extensions! Support different types of interactions such as closed curves.*



# MONAI Label + OHIF



***MONAI Label***

# Demo

- **Scenario 1: Cold start - No segmentations available**
  1. Start App: `monailabel start_server -a ./sample-apps/deepedit/ -s /home/adp20local/Documents/Datasets/monailabel_datasets/Slicer/spleen/demo_scenario1/`
  2. Segmentation based on scribbles
  3. Start training in the background
- **Scenario 2: Interactive segmentation using DeepEdit**
  1. Inference
  2. Active Learning Strategy
- **Scenario 3: Use OHIF and DICOMweb server**
  1. Start App using Orthanc DICOMweb server: `monailabel start_server -a ./sample-apps/deepedit/ -s http://127.0.0.1:8042/dicom-web`
  2. Show images loaded in Orthanc
  3. Use DeepEdit in OHIF

# Conclusions and future work:

## **Conclusion:**

- MONAI Label is a open source project that facilitates annotations of 3D medical images.
- MONAI Label is one of the first frameworks that introduces Active learning strategies in a software annotation setting.

## **Future Work:**

- Multimodality support
- Multiple apps under single server
- Self-supervised learning or unsupervised learning algorithms to leverage unlabeled data for better performance.

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Thanks!