

Images

10.14.24

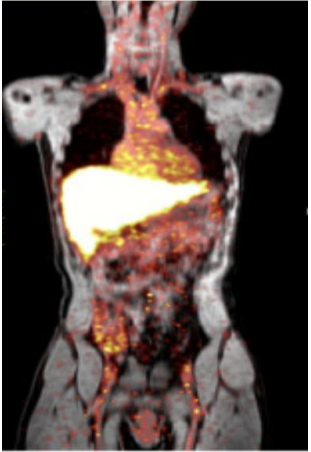
Learning Objectives

- Preprocessing
- Segmentation
- Transformations

What types of info can an “image” contain?

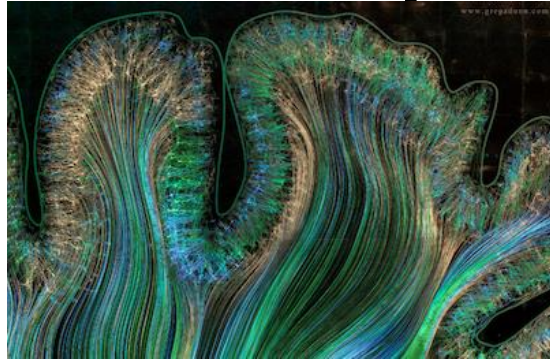
What types of info can an “image” contain?

Composition



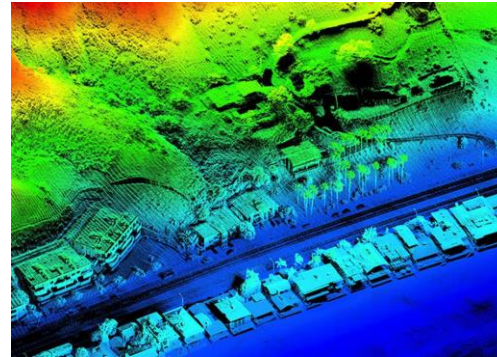
PET

Geometry



Diffusion-Tensor
Image

Depth



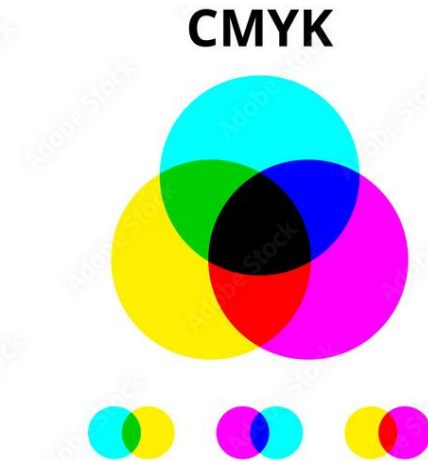
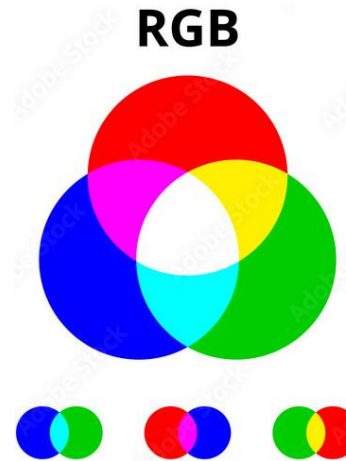
LIDAR

Temperature



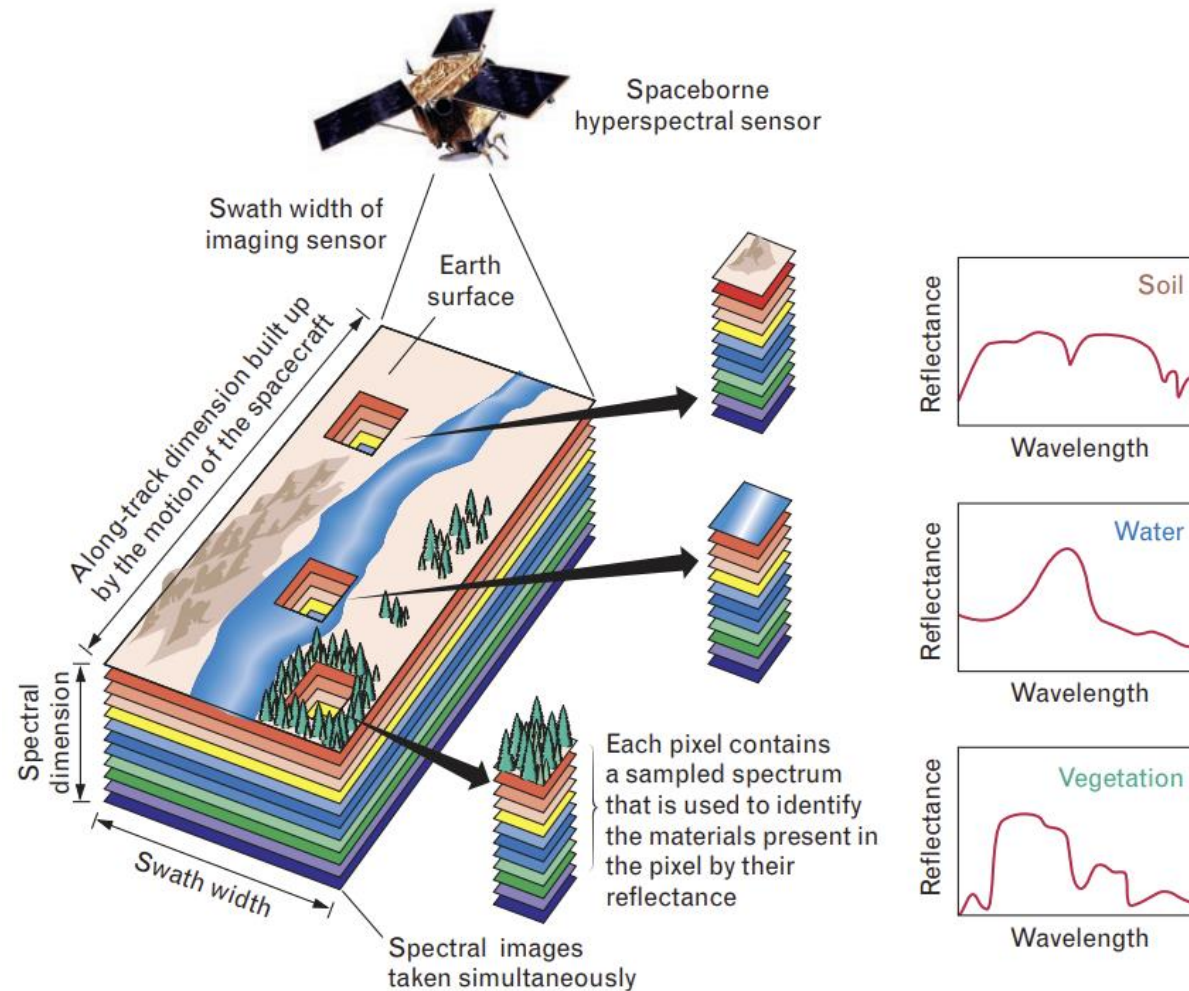
Infrared

“Color” in camera-imaging



“Color” in camera-imaging

Hyperspectral Imaging

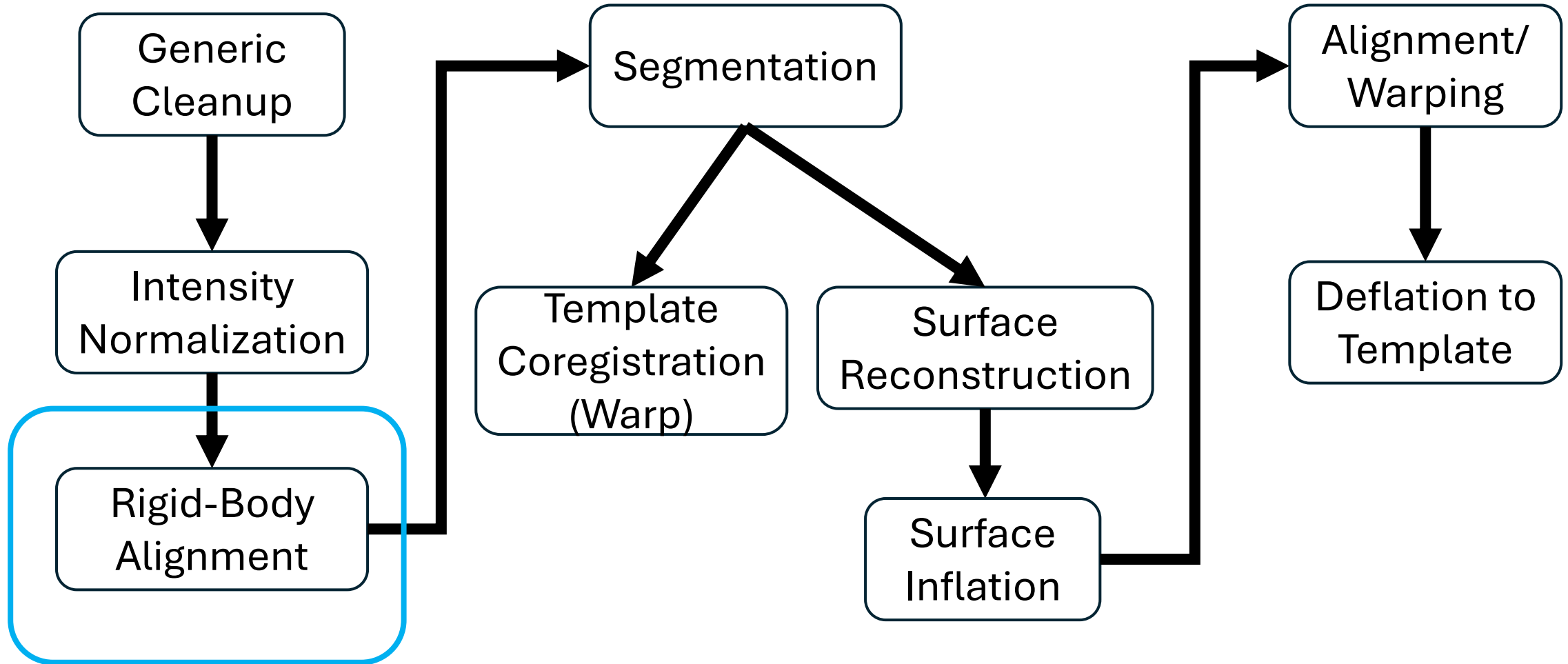


What are some challenges with image-data?

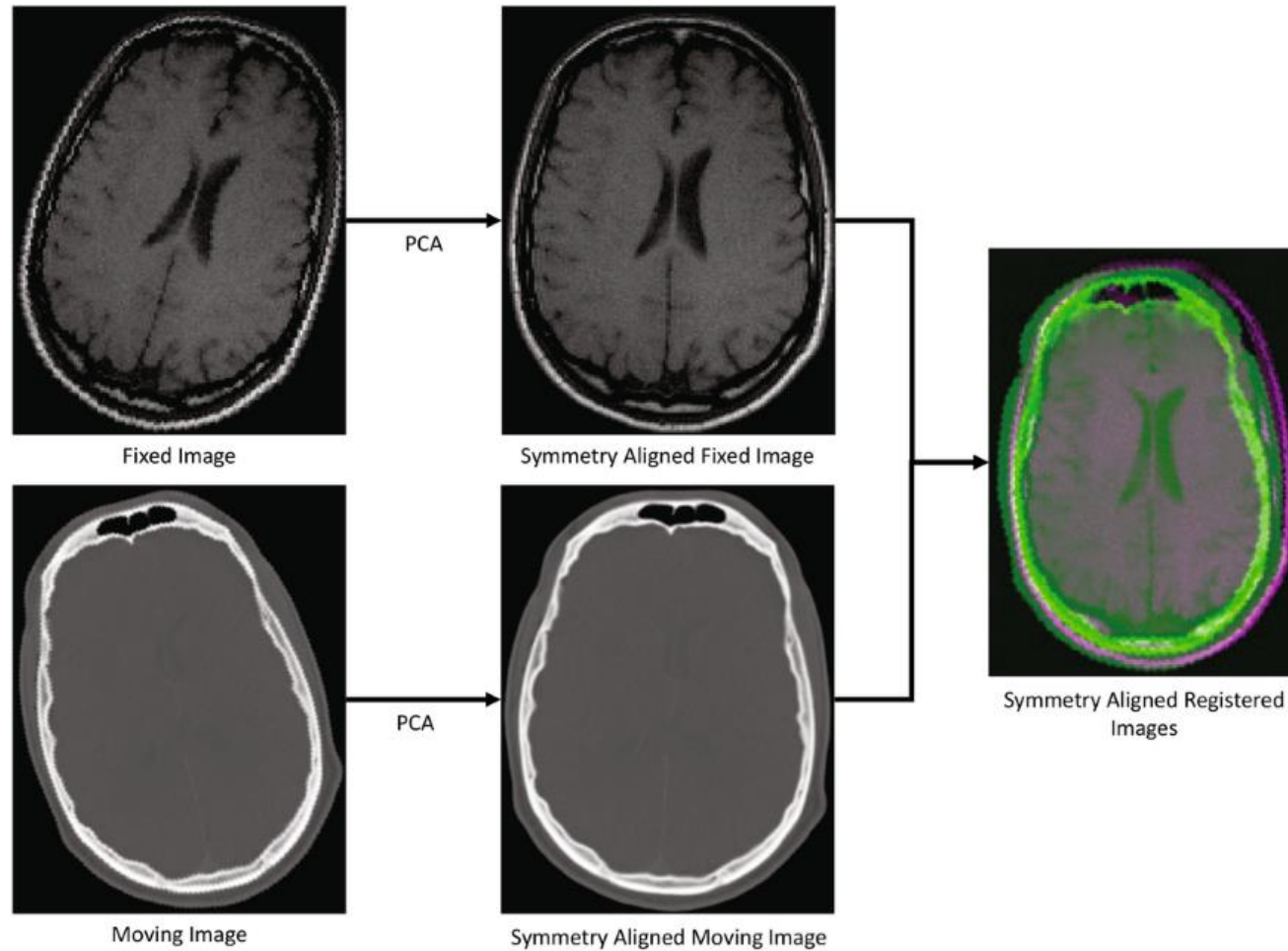


Some example images for “computer”

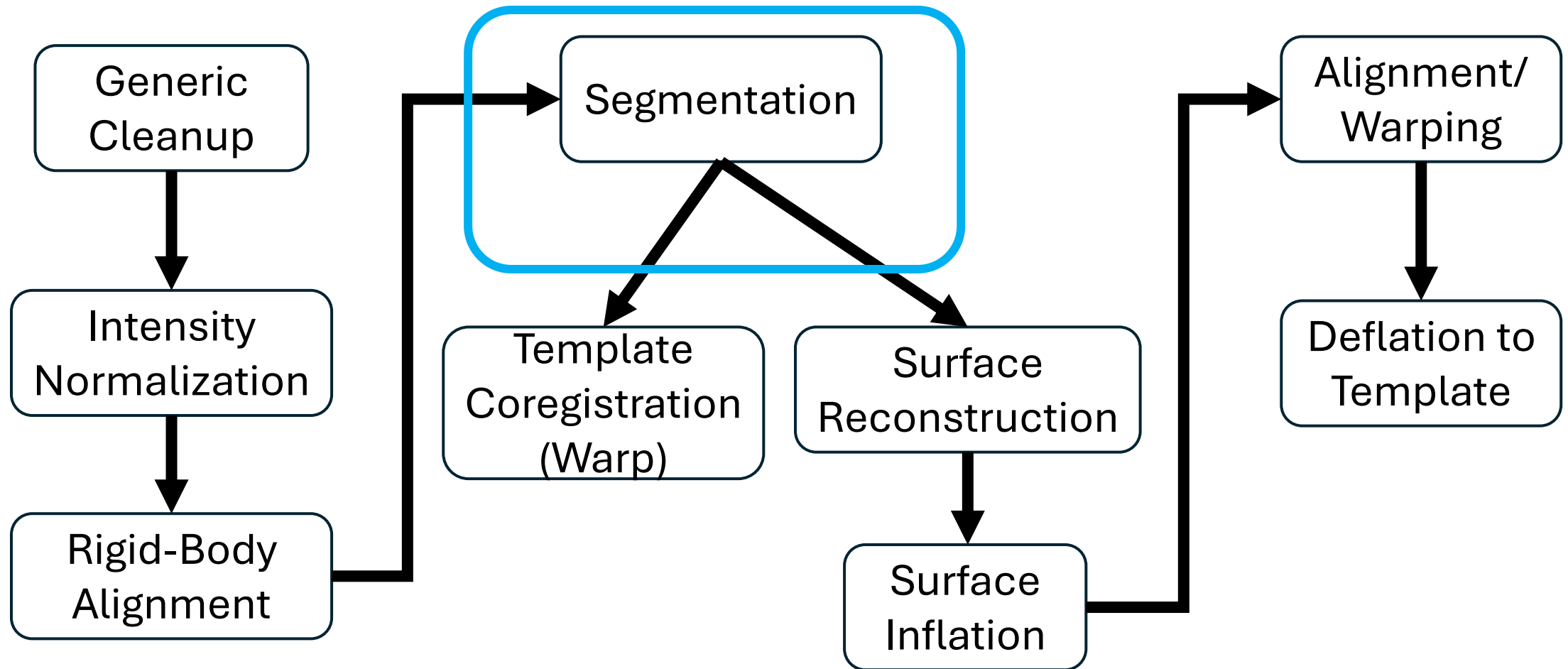
A typical fMRI processing pipeline



Registration



A typical fMRI processing pipeline

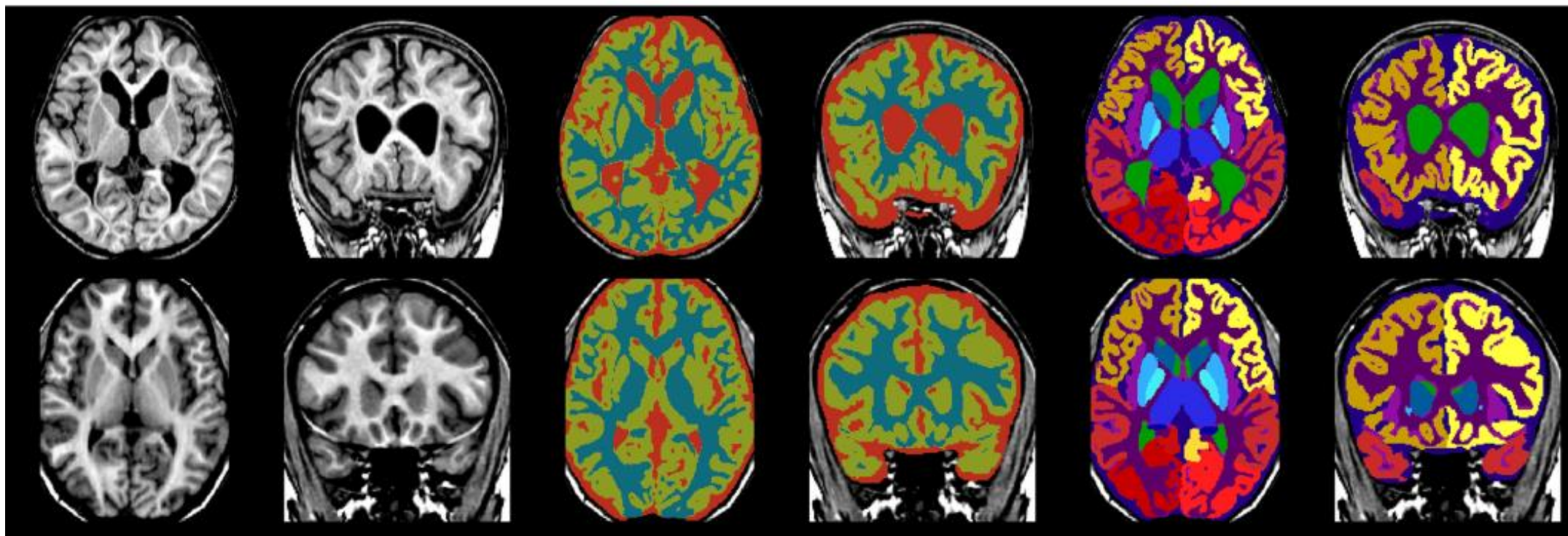


Segmentation

T1w image

Tissue segmentation

Structural segmentation



Imaging Modules in Python

- **opencv**

- Much faster. Designed for real-time application
- C++ and builtin CUDA integration (GPU)
- Color is BGR

- **scikit-image**

- Slower
- Easier access to basic processing functions
- Built-in generalization to ND images (e.g., volumes)
- Color is RGB

Scikit-Image Submodules

- Data: classic examples in image-processing

astronaut



coffee



camera



coins



```
coins = skimage.data.coins()
```

Scikit-Image Submodules

- **Feature**
- **Filters**
- **IO**
- **Morphology**
- **Restoration**
- **Segmentation**
- **Transform**

Processing: Filtering

Have I ever mentioned convolution?

Image Restoration

- Smoothing: Gaussian, box (movavg), median

```
from skimage import data,io,filters
from matplotlib import pyplot as plt
import numpy as np
from numpy import random as random

cam=data.camera();
## Poisson noise
camNoiseE=cam+random.exponential(10,np.shape(cam));
## Uniform noise
camNoiseU=cam+random.random(np.shape(cam))*20;
```



Jupyter

Image Restoration

- Convolution model of blurring

$$X_{meas} = X_{true} * M_{blur} + \eta_{noise}$$

- Convolution Theorem:

$$\mathcal{F}[A * B] = \mathcal{F}[A] \cdot \mathcal{F}[B]$$

$$X_{true} \approx \mathcal{F}^{-1} \left(\frac{\mathcal{F}[X_{meas}]}{\mathcal{F}[M_{blur}]} \right)$$

→ Gaussian η : Wiener Filter

→ Poisson η : Richardson-Lucy

Image Restoration



Input Image

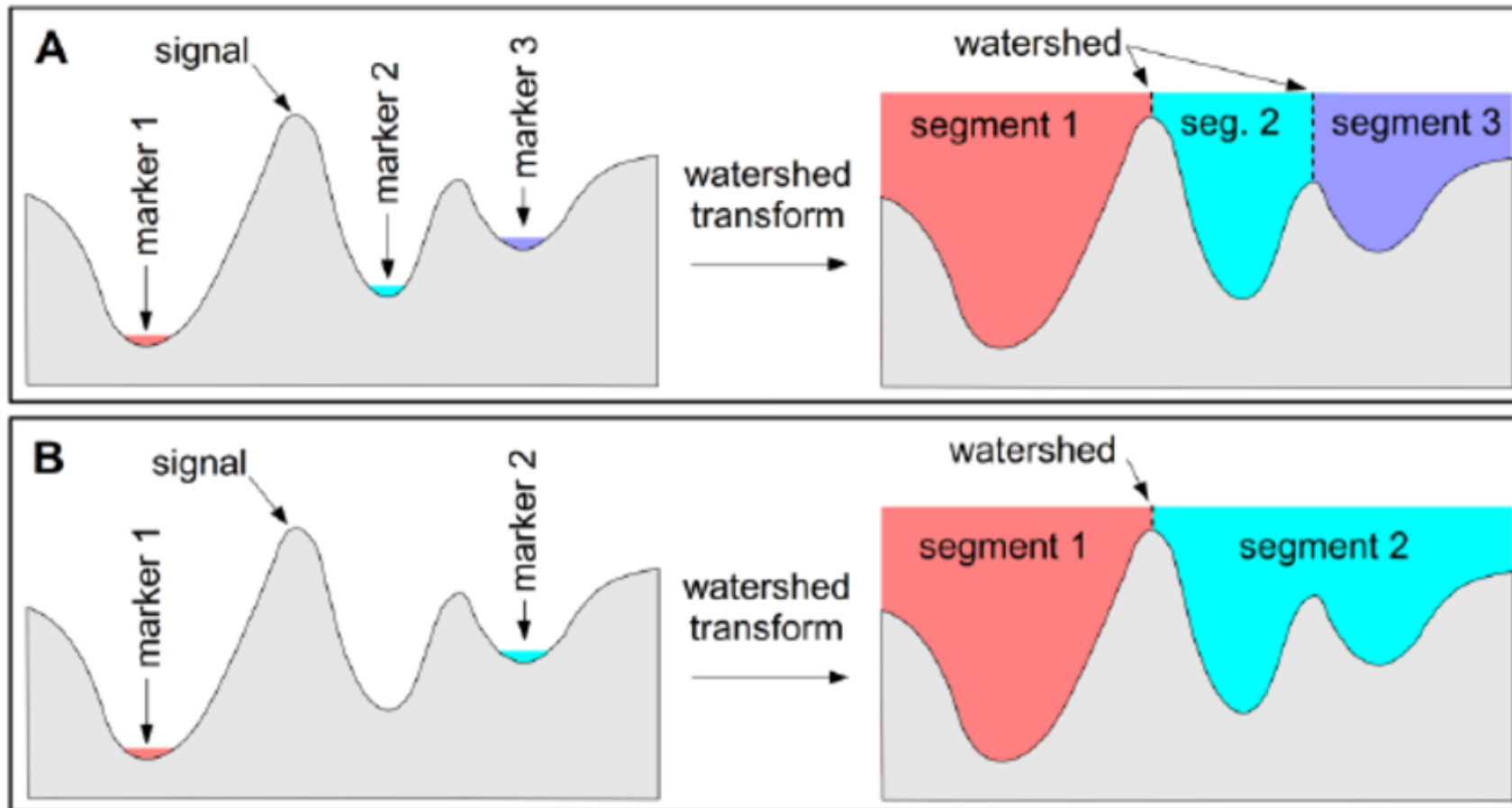


Simulate blur and Noise



Restored Image

Segmentation: Watershed Algorithm



Sobel Gradient Approximation

- Edges are where the change in intensity (gradient) is large
- Sobel kernels: gradient in one direction, smoother in the other:

$$G_x = \begin{bmatrix} 1 & 0 & -1 \\ 2 & 0 & -2 \\ 1 & 0 & -1 \end{bmatrix}, \quad G_y = G_x^T$$

- Full Sobel is gradient magnitude: $G_x^2 + G_y^2$

Coins Example

- Jupyter

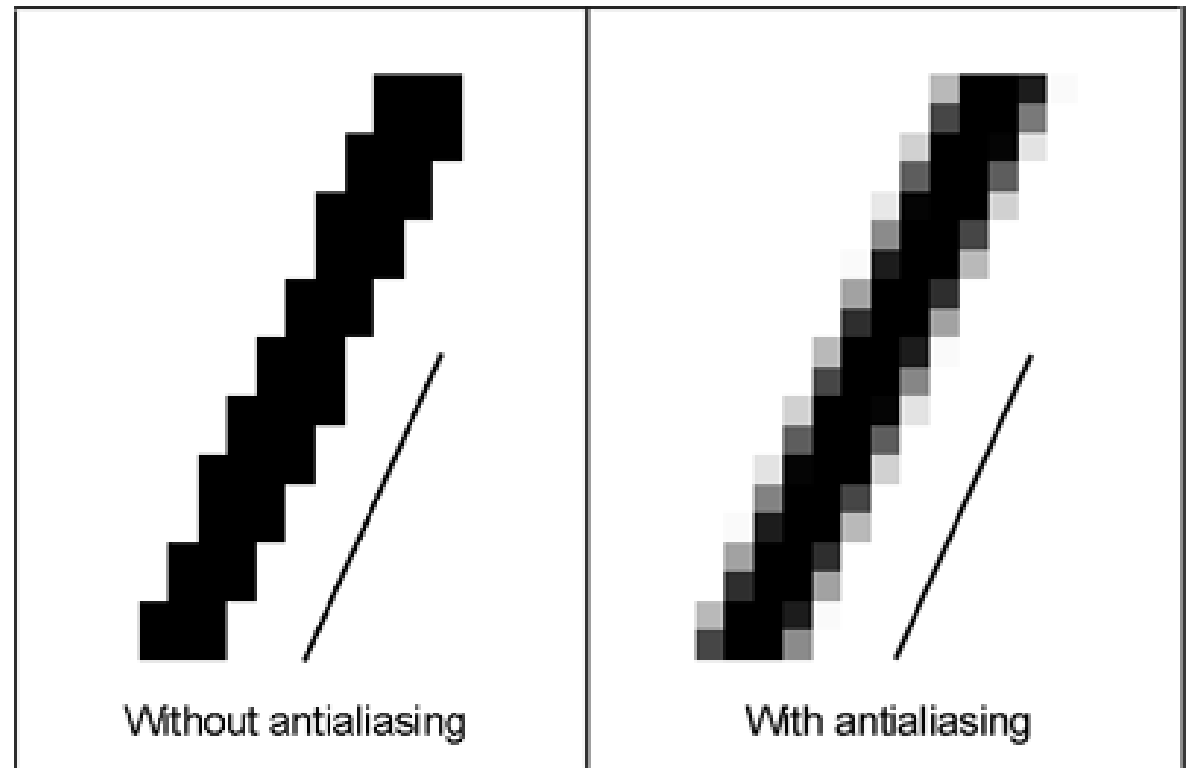


Practice: Edge detection

- Find values to separate the image into coins and background
- Apply the Sobel filter: `filters.sobel` and create an image of gradient intensity

Resizing

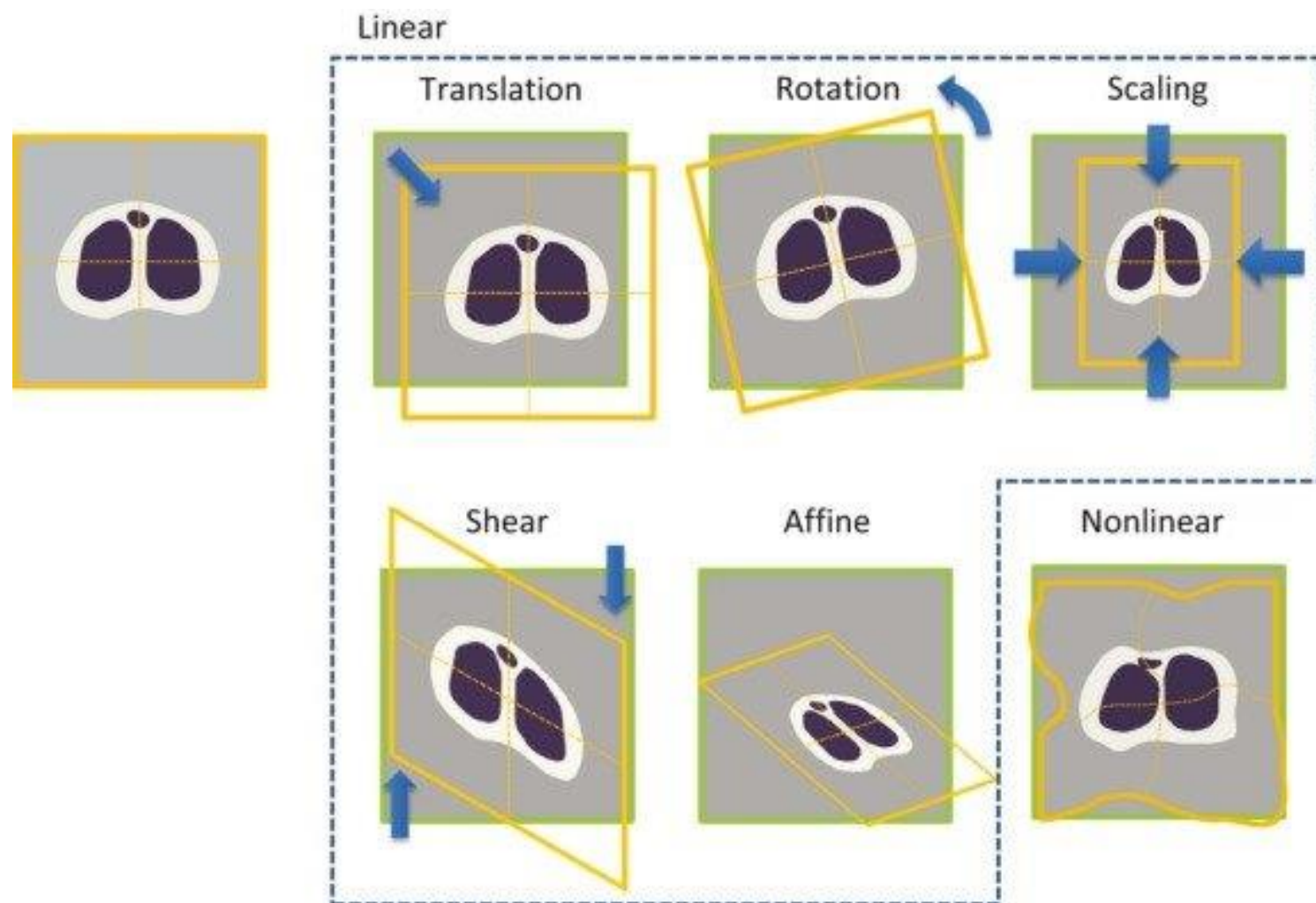
- rescale: Isotropic (same in each direction)
- resize: arbitrary dimensions
- antialiasing: usually Gaussian smoothing kernel



Transformation-Types

- Euclidean: Size-Preserving (rigid-body)
 - Translation, rotation
- Affine
 - Translation + Linear operator
 - Includes resizing and shearing transforms
- Nonlinear
 - Curvilinear warping
 - Custom coordinate-systems

Transformation Types



Learning Objectives

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