

PRESENTER:

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BACKGROUND: itk-elastic is a pythonic wrapping of the scientific image registration toolbox *elastix*.

Elastix

- ... is based on ITK
- ... offers modular design
- ... has two decades of development



Key characteristics

- Registration of 2D, 3D, 4D images
- Rigid / Affine / Non-rigid (B-splines)
- Multi-metric, Multi-resolution
- Setting masks/points to aid registration
- Detailed logging with verbosity levels
- Transform images, meshes and point sets

Integration with project MONAI

Combine itk-elastic with the pytorch-based medical deep learning library MONAI

Jupyter Notebook tutorials

To get started quickly, see [/examples](#)

Model zoo

Collection of published parameter files at <https://elastix.lumc.nl/modelzoo/>

Integration with Napari as plugin

<https://github.com/SuperElastix/elastix-napari>

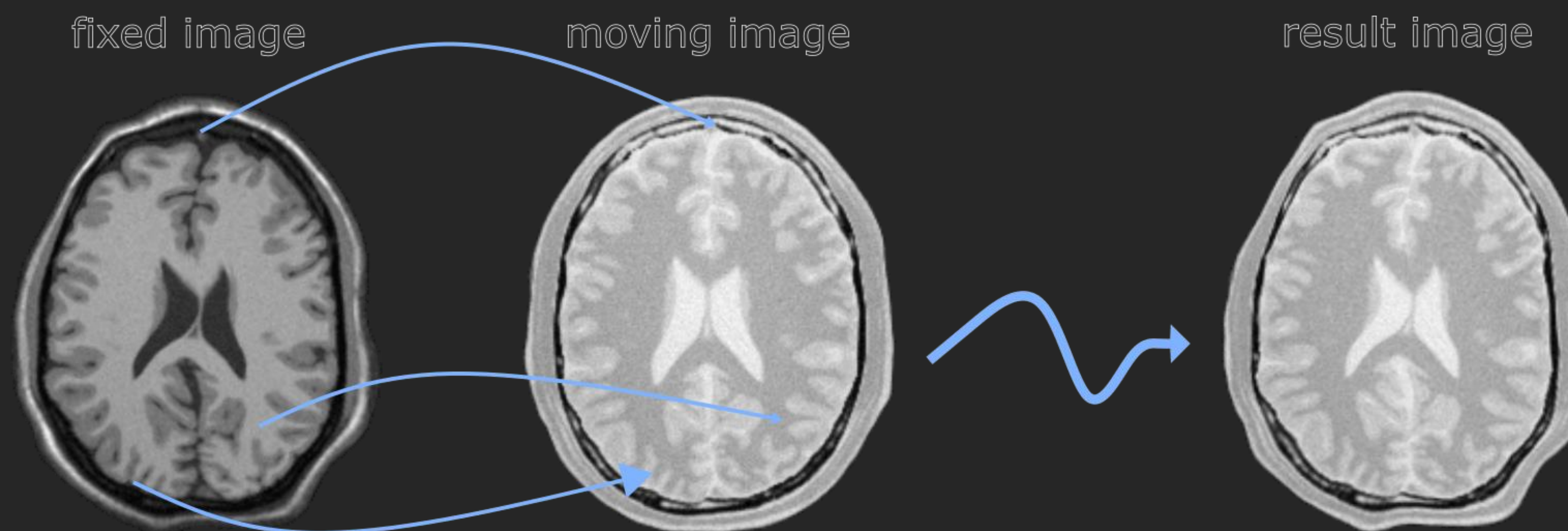
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itk-elastic: Medical image registration in Python *



GitHub repo



*** pip install itk-elastic today to use it for free**

<https://github.com/InsightSoftwareConsortium/ITKElastix>

CODE SNIPPET: Registration

```
import itk
from scipy.spatial.distance import dice

# Load the moving and the fixed image from disk
fixed_image = itk.imread('./data/fixed.mha', itk.F)
moving_image = itk.imread('./data/moving.mha',
itk.F)

# Configure a (default) parameter map with all the
# registration parameters
par_obj = itk.ParameterObject.New()
par_map = par_obj.GetDefaultParameterMap('bspline')
par_obj.AddParameterMap(par_map)

# Run the registration
# rtp: result transform parameter object
result_image, rtp = itk.elastix_registration_method(
    fixed_image, moving_image,
    parameter_object=par_obj)
```

CODE SNIPPET: Mask transformation & Dice calculation

```
(...continuing from the previous snippet)

# Load the corresponding masks
fixed_mask = itk.imread('./data/f_mask.mha', itk.UC)
moving_mask = itk.imread('./data/m_mask.mha', itk.UC)

# Transform the moving mask using the result from the
# registration
rtp.SetParameter(0, 'ResampleInterpolator',
    'FinalNearestNeighborInterpolator')
result_mask = itk.transformix_filter(moving_mask, rtp)

# Compute dice on masks
initial_dice = 1 - dice(fixed_mask[:].ravel(),
    moving_mask[:].ravel())
result_dice = 1 - dice(fixed_mask[:].ravel(),
    result_mask[:].ravel())
print(initial_dice, result_dice)
```

Parameter map/file example

```
(AutomaticParameterEstimation "true")
(CheckNumberOfSamples "true")
(DefaultPixelValue 0)
(FinalBSplineInterpolationOrder 3)
(FinalGridSpacingInPhysicalUnits 10)
(FixedImagePyramid "FixedGenericImagePyramid")
(GridSpacingSchedule 2.80322 1.9881 1.41 1)
(ImageSampler "RandomCoordinate")
(Interpolator "LinearInterpolator")
(MaximumNumberOfIterations 256)
(MaximumNumberOfSamplingAttempts 8)
(Metric "AdvancedMattesMutualInformation")
(MovingImagePyramid
    "MovingGenericImagePyramid")
(NewSamplesEveryIteration "true")
(NumberOfResolutions 4)
(NumberOfSamplesForExactGradient 4096)
(NumberOfSpatialSamples 2048)
(Optimizer
    "AdaptiveStochasticGradientDescent")
(Registration "MultiResolutionRegistration")
(Transform "BSplineTransform")
```