

MLIA Final Projects

- Please note that each paper has related code released online. Feel free to use it!
- All tasks will need to be performed on our provided images.
- Good luck and enjoy!

T3 (3 teams): this is for groups with THREE-member only!

*** Papers for tumor segmentation project:

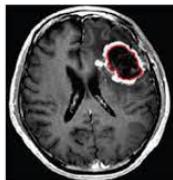


Fig. 1: An example of brain tumor segmentation.

[1] Swin UNETR: Swin Transformers for Semantic Segmentation of Brain Tumors in MRI Images

[<https://arxiv.org/abs/2201.01266>]

[2] Self Pre-training with Masked Autoencoders for Medical Image Classification and Segmentation

[<https://arxiv.org/abs/2203.05573>]

*** Papers for image classification project (healthy vs. MCI. vs. Alzheimer's disease):

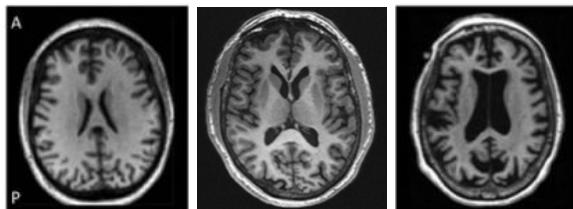


Fig. 2: An example of healthy (left) vs. MCI (middle) vs. disease (right).

[3] Supervised Contrastive Learning

[<https://arxiv.org/pdf/2004.11362v5.pdf>]

T4 (12 teams): this is for groups with FOUR-member only!

***** Papers for pairwise image registration project:**

Tasks: (a) Train and test the model below on brain MR images; (b) To understand the generalizability of the registration models on complex datasets. Train the model on datasets (MNIST) and test the performance on unseen datasets (Google Draw).

[1] Fourier-Net+: Leveraging Band-Limited Representation for Efficient 3D Medical Image Registration

[[https://arxiv.org/pdf/2307.02997](https://arxiv.org/pdf/2307.02997.pdf)]

[2] TransMorph: Transformer for unsupervised medical image registration

[<https://arxiv.org/abs/2111.10480>]

[3] DiffuseMorph: Unsupervised Deformable Image Registration Using Diffusion Model

[<https://arxiv.org/abs/2112.05149>]

[4] uniGradICON: A Foundation Model for Medical Image Registration

[[https://arxiv.org/pdf/2403.05780](https://arxiv.org/pdf/2403.05780.pdf)]

***** Papers for spatiotemporal image registration**

Dataset: A dataset with video sequences (2D x T) cardiac MR images (download from here: <https://www.creatis.insa-lyon.fr/Challenge/acdc/databases.html>). Use all images in the “Training Dataset” for training, and the ‘Testing Dataset’ for testing.

[5] TLRN: Temporal Latent Residual Networks For Large Deformation Image Registration

[<https://pmc.ncbi.nlm.nih.gov/articles/PMC11929566/>]

[6] MAE-TransRNet: An improved transformer-ConvNet architecture with masked autoencoder for cardiac MRI registration

[<https://pubmed.ncbi.nlm.nih.gov/36968818/>]

***** Papers for image regression project**

Dataset: A dataset with cardiac CMR video sequences (2D x T) that encode heart motions with provided ground truth labels.

Task: (a) First train a segmentation model [***U-Net: Convolutional Networks for Biomedical Image Segmentation***, [https://arxiv.org/pdf/1505.04597](https://arxiv.org/pdf/1505.04597.pdf)] to segment heart contours from all

images; **(b)** Implement a regression task using the predicted heart contours as inputs, selecting a network from the list below:

[7] ViViT: A Video Vision Transformer [<https://arxiv.org/abs/2103.15691>]

[8] Sequencer: Deep LSTM for Image Classification [<https://arxiv.org/pdf/2205.01972.pdf>]

*** Papers for image generation project

Dataset: Each team is required to download the dataset from <https://github.com/bearpaw/sysu-shape-dataset/tree/master> including all images of airplane, bicycle, boat, car, motorbike (e.g. <https://github.com/bearpaw/sysu-shape-dataset/tree/master/motorbike/images>).

Task: Each team will implement an image generation task using selected network from the list below:

[9] Denoising Diffusion Probabilistic Models
[<https://arxiv.org/abs/2006.11239>]

[10] Diffusion Models Beat GANs on Image Synthesis
[<https://arxiv.org/pdf/2105.05233.pdf>]

[11] High-Resolution Image Synthesis with Latent Diffusion Models
https://openaccess.thecvf.com/content/CVPR2022/papers/Rombach_High-Resolution_Image_Synthesis_With_Latent_Diffusion_Models_CVPR_2022_paper.pdf

[12] NVAE: A Deep Hierarchical Variational Autoencoder
[<https://arxiv.org/pdf/2007.03898.pdf>]