

Neural Networks for Detection of Subtle Epilepsy Lesions in MRI and Translation to Clinical Evaluation

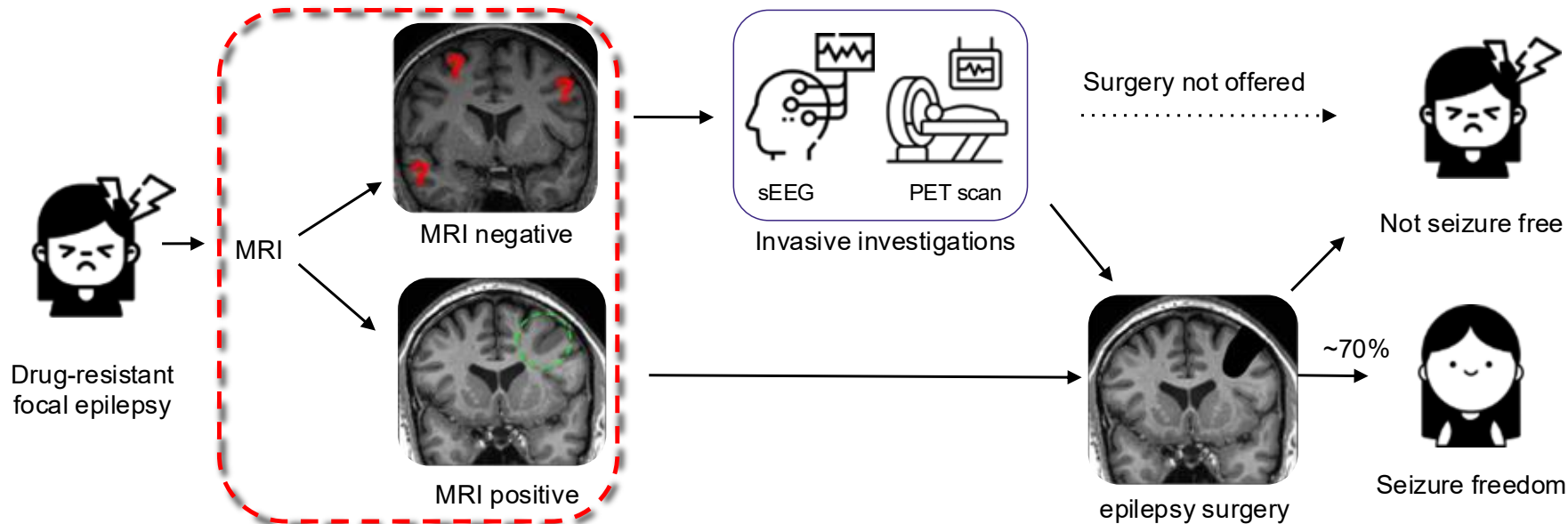


Dr Mathilde Ripart

KCL School of Biomedical Engineering & Imaging Sciences, UK

UCL Great Ormond Street Institute of Child Health, UK

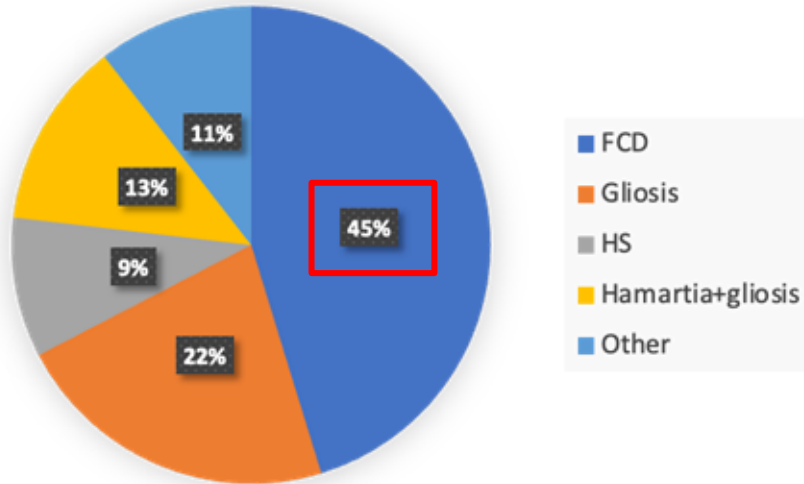
Goal: Improving the presurgical planning of patients with drug-resistant focal epilepsy



Can AI help us detect subtle lesions?

“MRI-negative” Epilepsy - “Missed” lesion

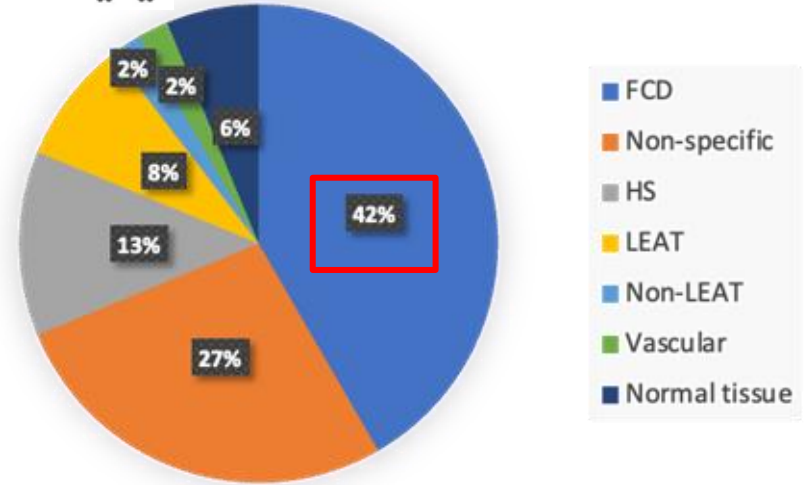
 Adults



95 MRI negative

Adapted from Wang et al., 2013 Mod Path

 Children



87 MRI negative

Eriksson et al., 2023 Epilepsia

Focal cortical dysplasia (FCD)

Malformation of cortical development

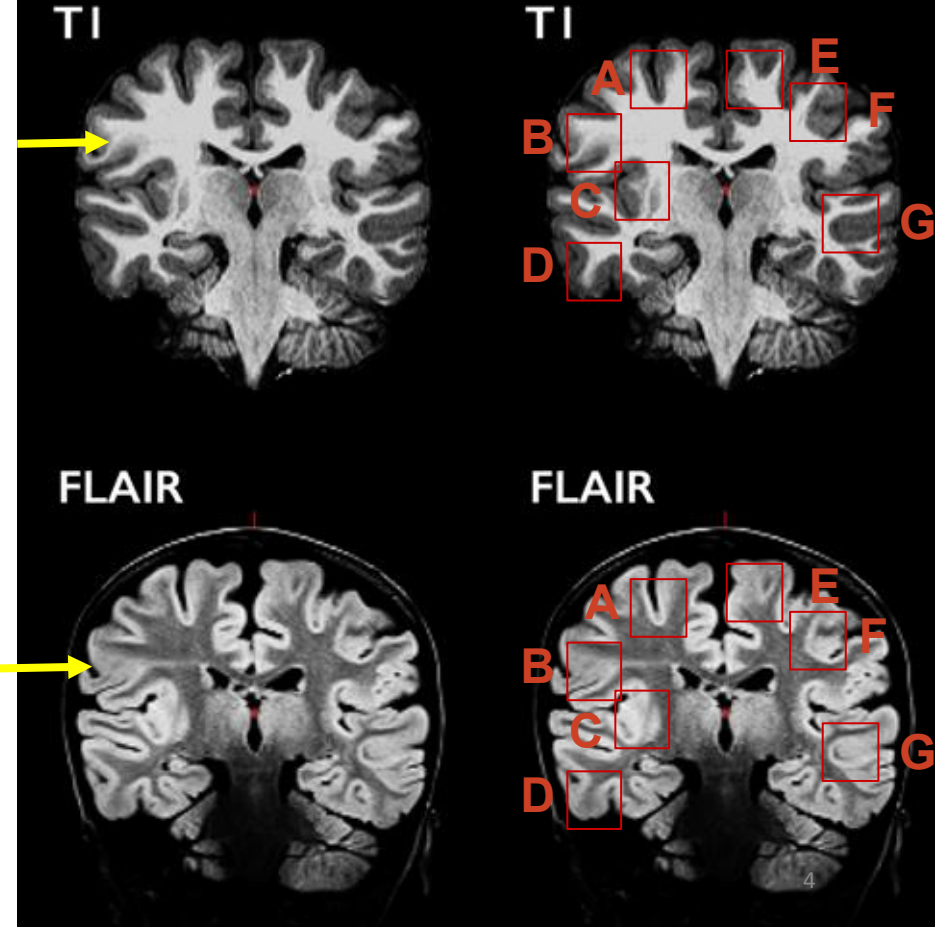
Subtle imaging features:

- Abnormal cortical thickness
- Blurring of the grey-white matter boundary
- FLAIR/T2-weighted intensity changes
- Abnormal folding pattern
- Hemispheric asymmetry

Where is the FCD?



Join at [menti.com](https://www.menti.com) | use code 8459 7868



A public benchmark for human performance in the detection of FCD

Rater group	Pinpointing (%)	Detection (%)		
		OVO, 0.0	Standard, 0.7	Adjusted, 0.22
Overall				
Layperson	29 [23, 34]	33 [27, 38]	2 [0, 4]	27 [22, 33]
Non-expert	48 [46, 51]	54 [52, 56]	3 [2, 4]	45 [42, 47]
Expert	68 [63, 72]	74 [70, 79]	9 [6, 12]	68 [63, 73]

Lennart Walger. (2025). Epilepsia Open. 10.1002/epi4.70028.

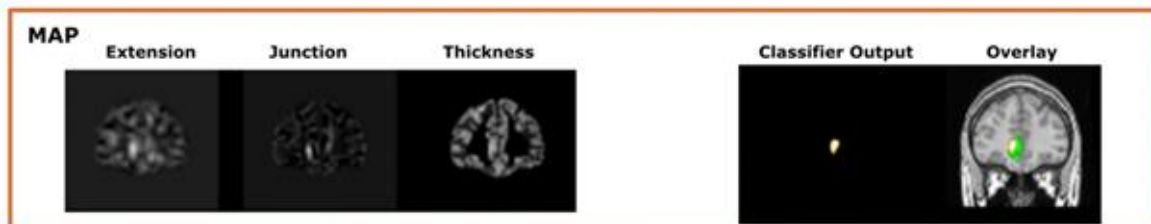
Key points of this talk

- AI can detect subtle focal cortical dysplasia (FCD) on MRI - MELD MLP
- Graph Neural network for a more confident detection of FCD - MELD Graph
- Translation into clinical practice
- Going further - a deep learning model for the detection of multiple epilepsy lesions

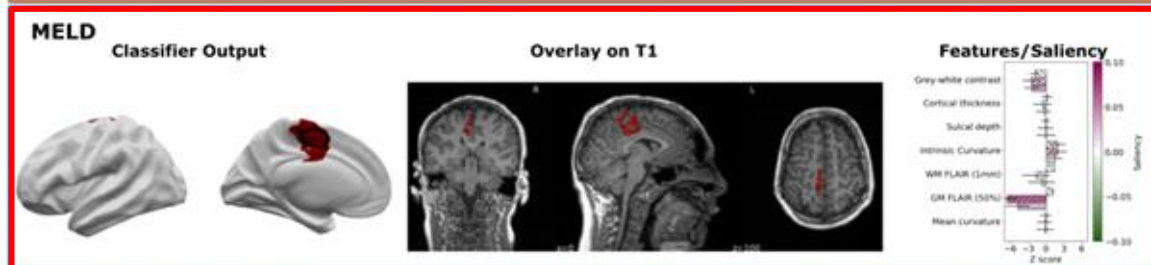
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AI tools with multi-centre validation



David et al., Epilepsia 2021



Spitzer, Ripart et al., BRAIN 2022



Gill et al., Neurology, 2021

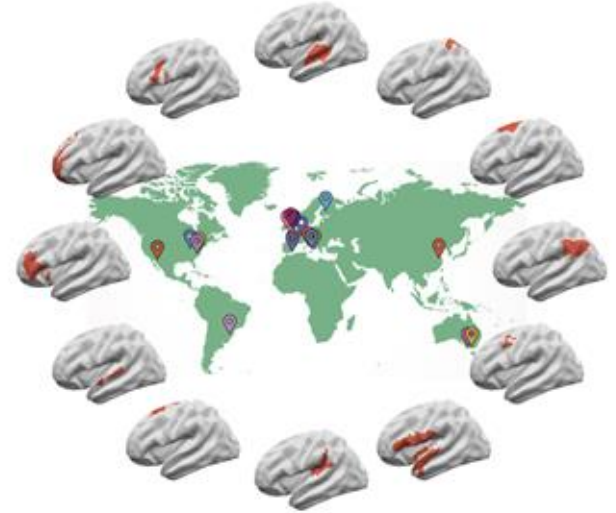


Dr Hannah Spitzer

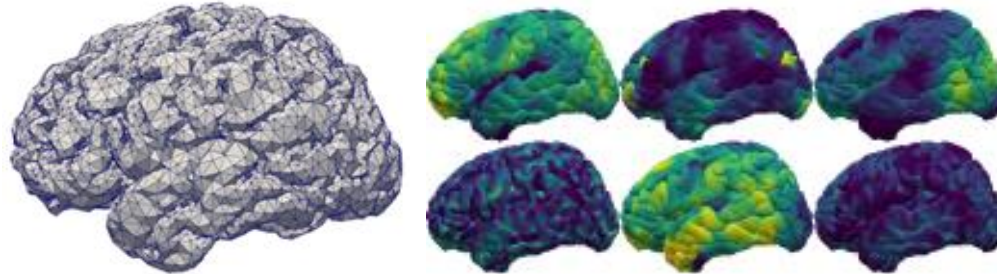
AI requires BIG DATA

MELD FCD dataset :

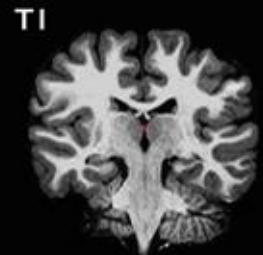
- 22 epilepsy surgery centres
- 1015 participants:
 - 618 patients with epilepsy due to FCD
 - 397 controls
- Paediatric & Adult
- 1.5T and 3T MRI



Surface based features



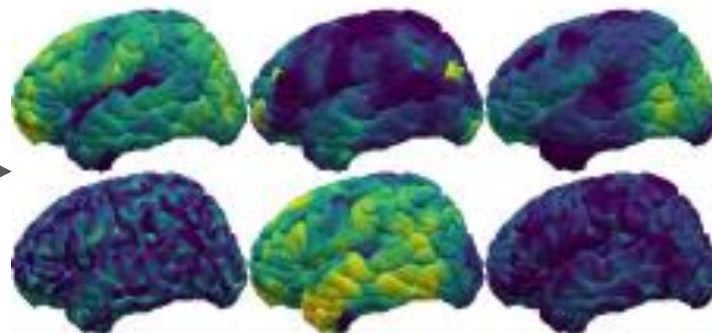
Surface-based features extraction



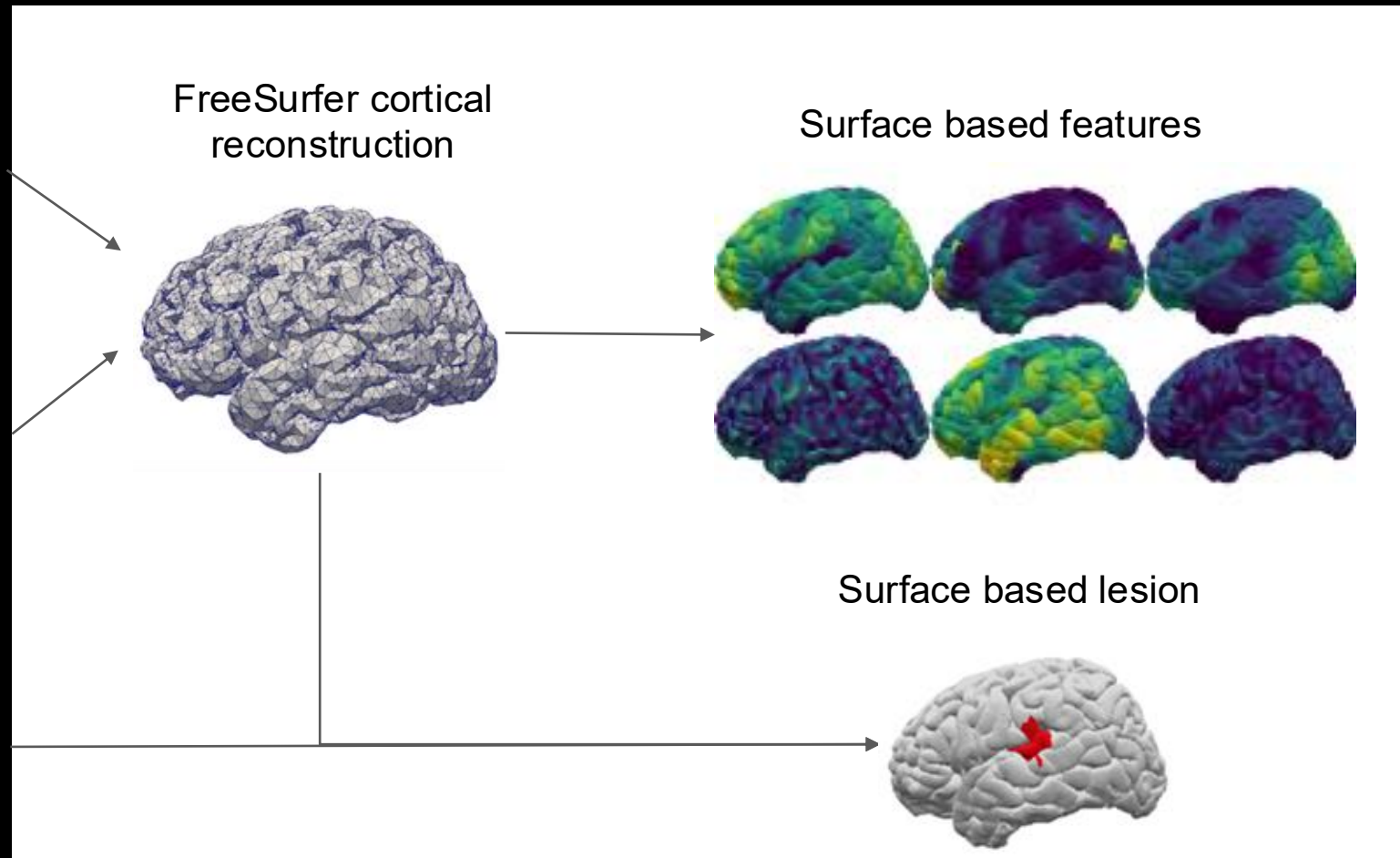
FreeSurfer cortical reconstruction



Surface based features



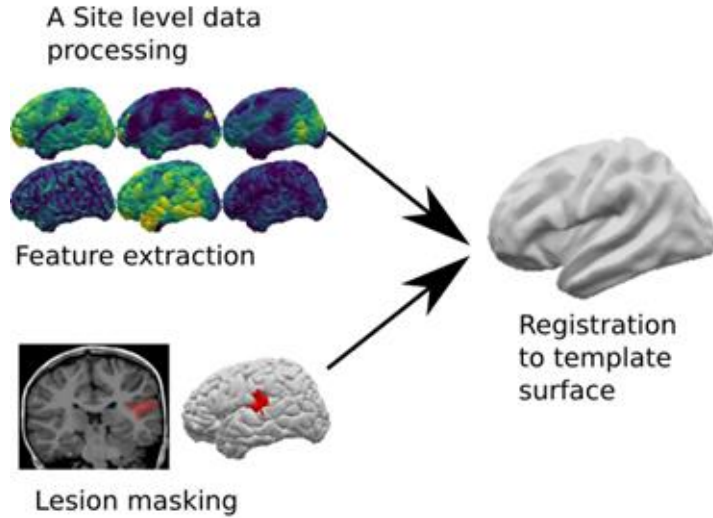
Surface based lesion



FCDs are non-uniformly distributed across the cortex

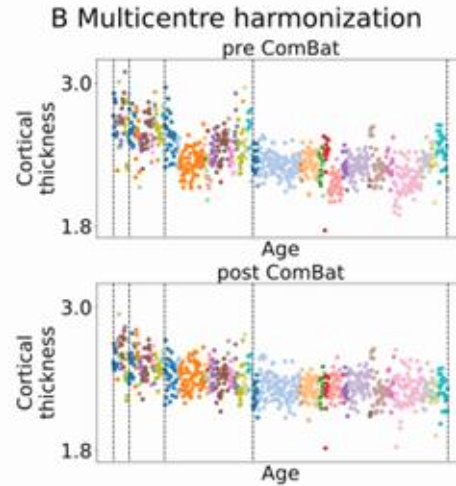
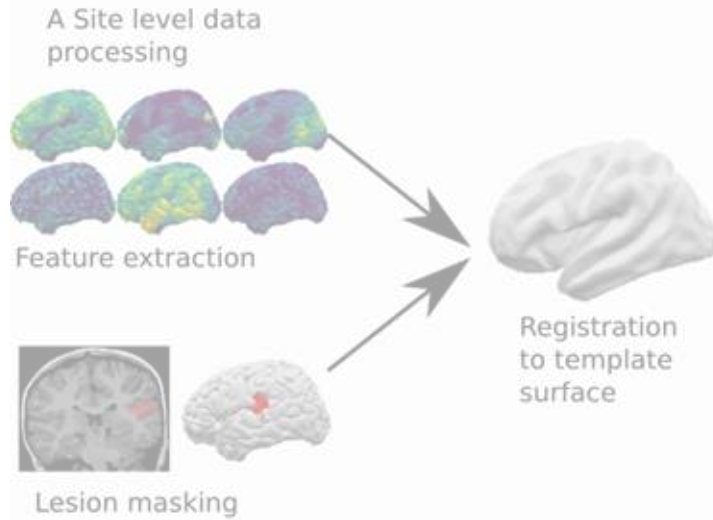


Automated lesion detection - Per Vertex “MLP”



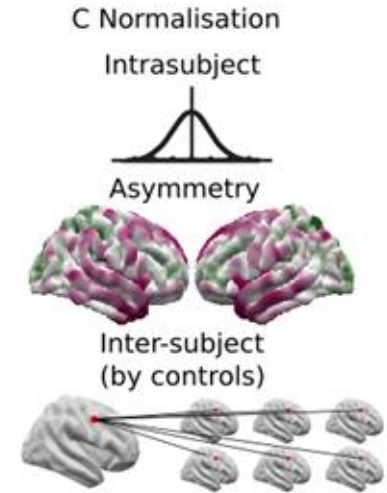
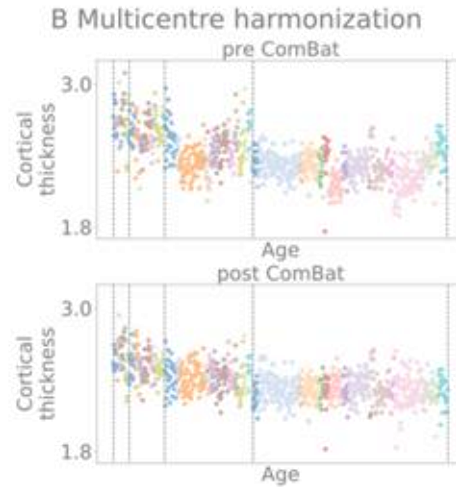
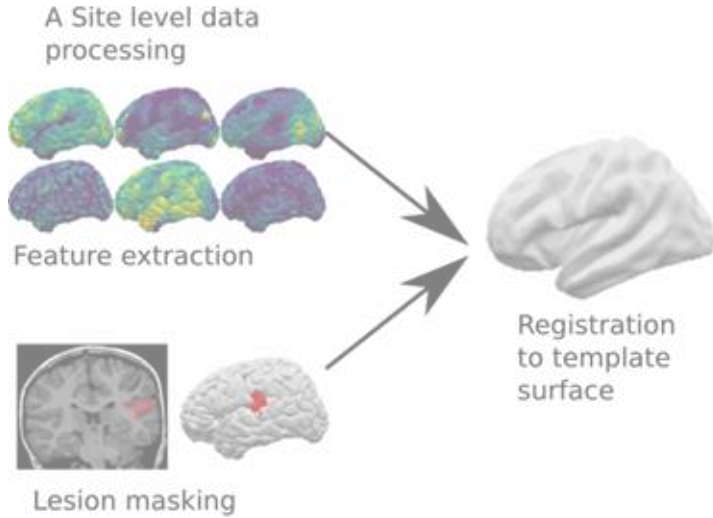
Dr Hannah Spitzer

Automated lesion detection - Per Vertex “MLP”



Dr Hannah Spitzer

Automated lesion detection - Per Vertex “MLP”



Dr Hannah Spitzer

Can you find the lesion: MRI

Patient 00:

- 18 year old male with focal epilepsy.
- Drug resistant epilepsy
- MRI

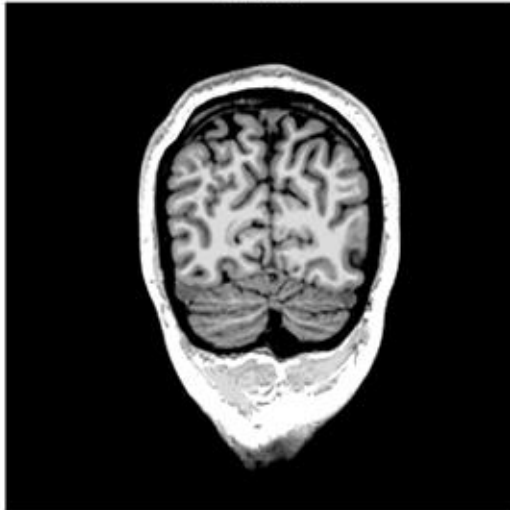
Q1. How confident are you about the lesion location?

1. Absolutely no idea - I can't see a lesion
2. Everything looks a bit abnormal
3. One or two areas look suspicious
4. I think I know where it is
5. I am very confident I know where the lesion is

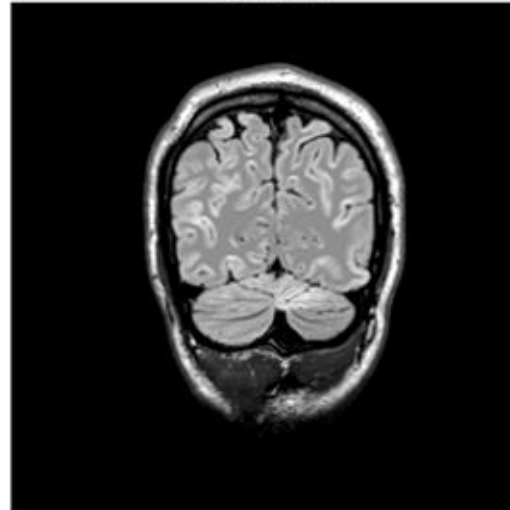
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T1 - Slice 58



FLAIR - Slice 58



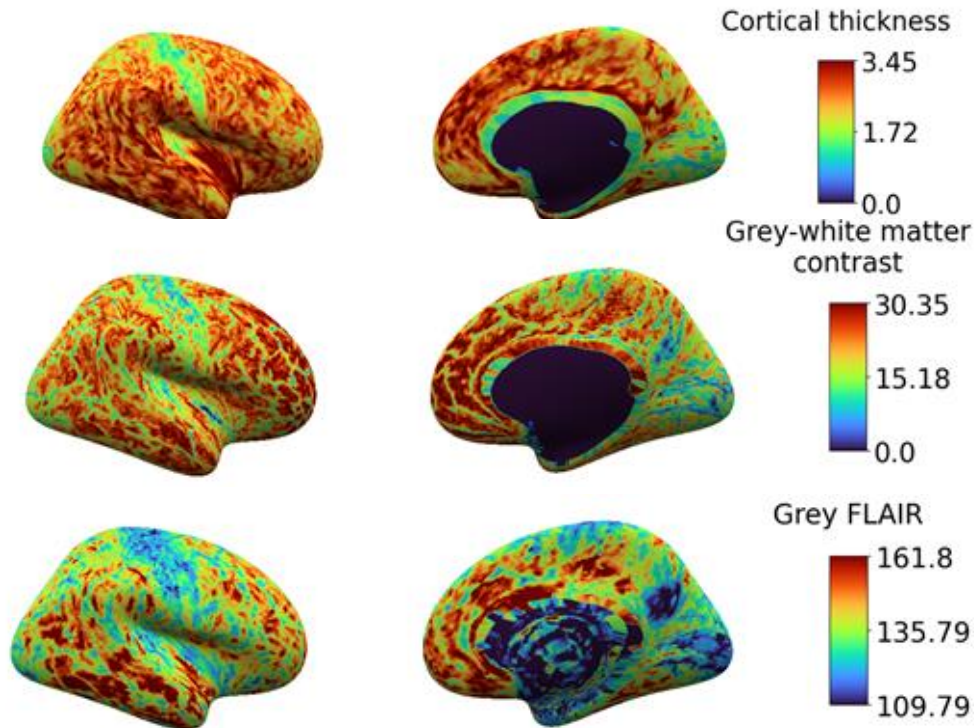
Can you find the lesion: Raw surface features

Q2. How confident are you about the lesion location?

1. Absolutely no idea
2. Everything looks a bit abnormal
3. One or two areas look suspicious
4. I think I know where it is
5. I am very confident

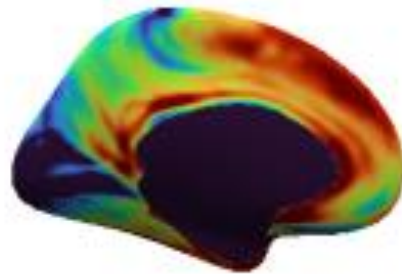
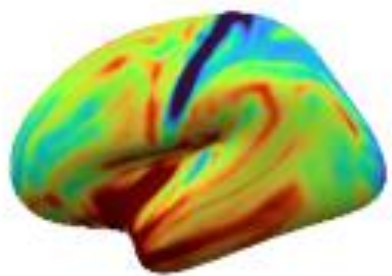


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Adjusting for normal patterns

Control mean cortical thickness



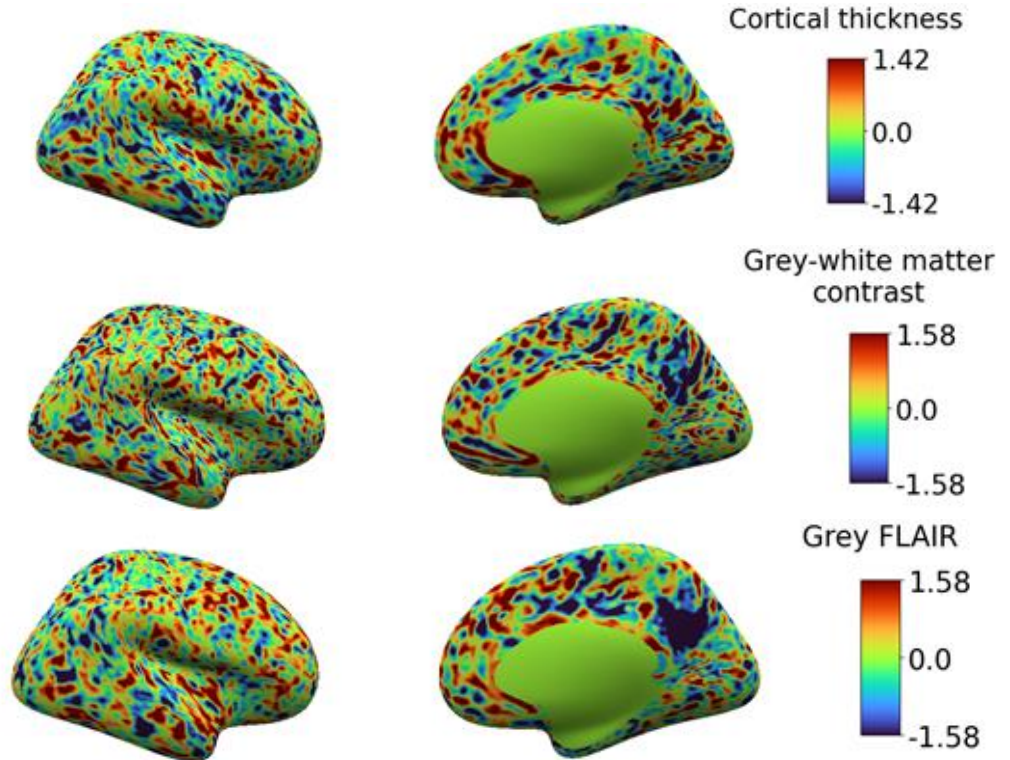
Can you find the lesion: Normalised surface features

Q3. How confident are you about the lesion location?

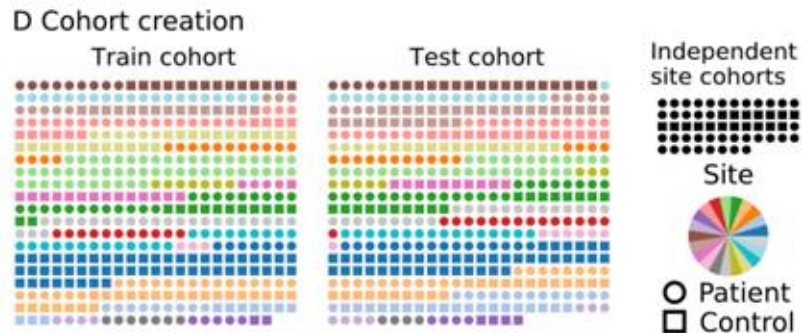
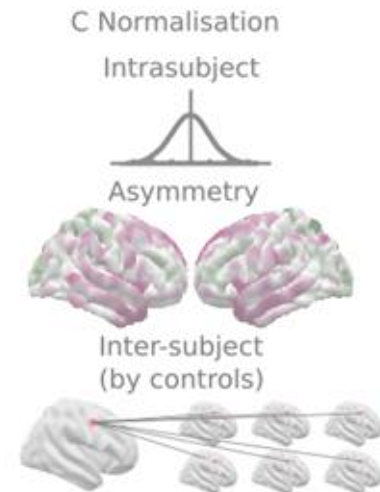
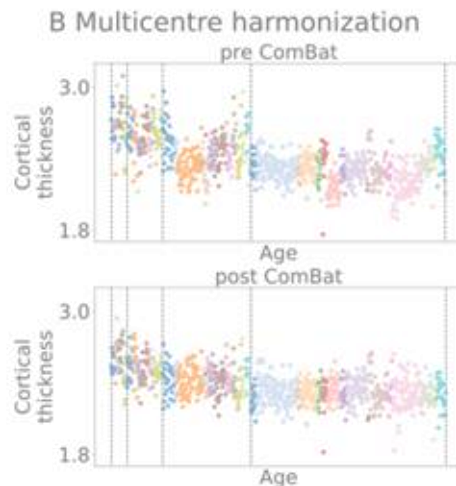
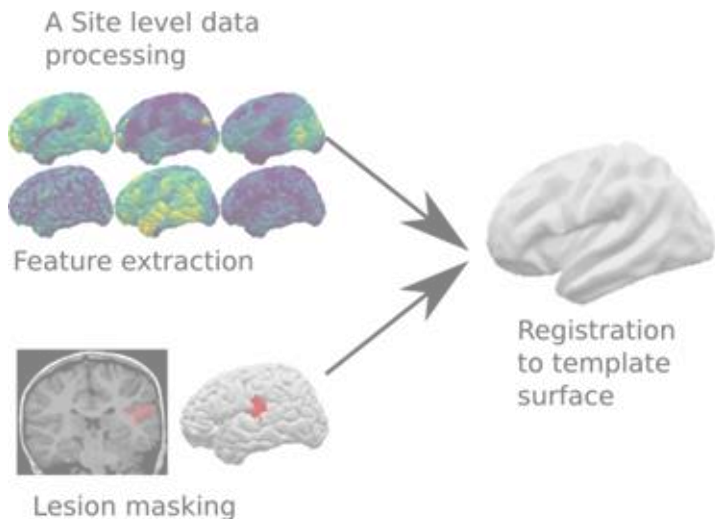
1. Absolutely no idea
2. Everything looks a bit abnormal
3. One or two areas look suspicious
4. I think I know where it is
5. I am very confident



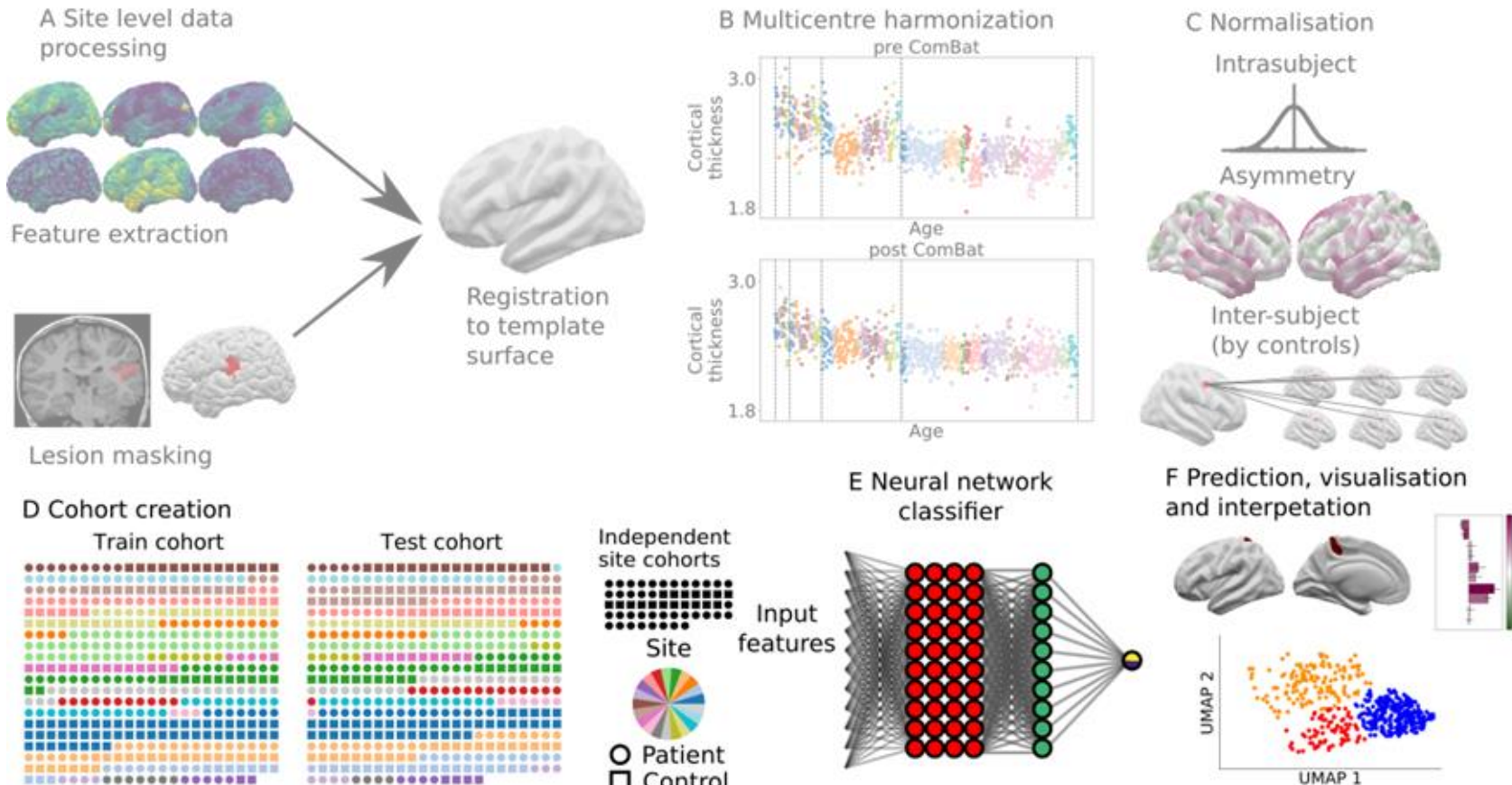
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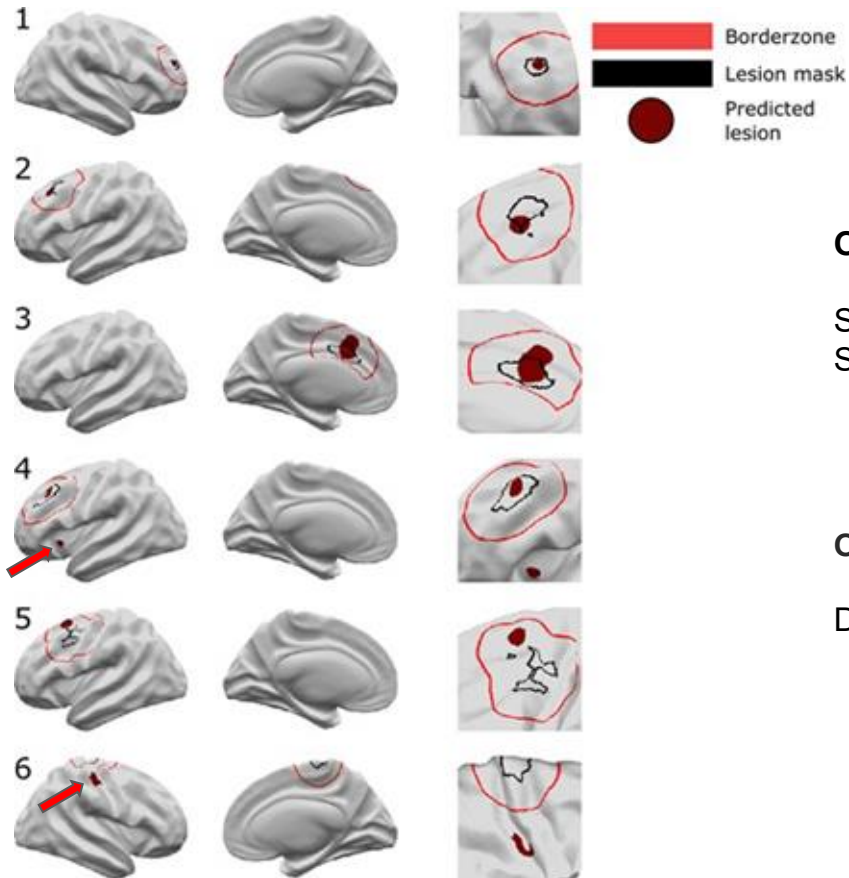
Automated lesion detection - Per Vertex "MLP"



Automated lesion detection - Per Vertex "MLP"



Automated lesion detection - Per Vertex "MLP"



On the test dataset

Sensitivity 67% (n=260)

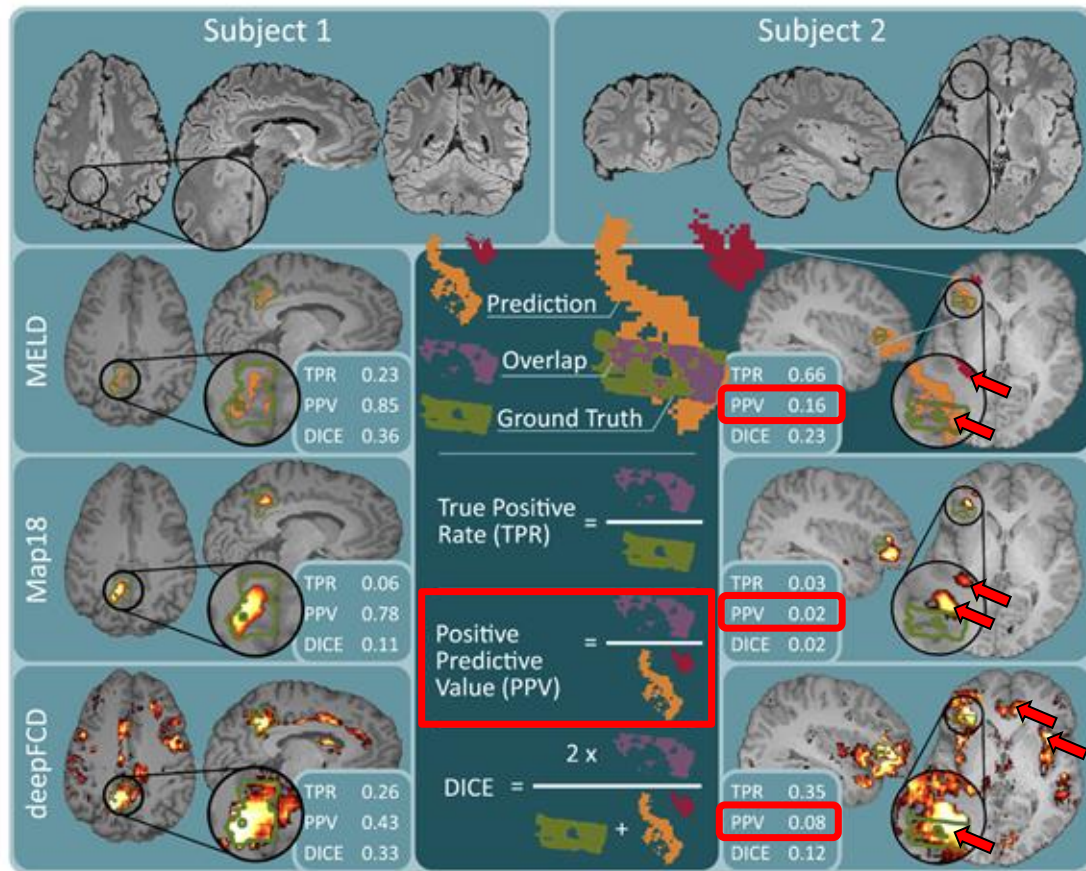
Specificity 54% (n=193)

**2 false positives
per patient on
average
(up to 12!)**

On the whole dataset

Detection in MRI negative: 63% (n=178)

False Positives are an outstanding challenge for all FCD detection algorithms



Spitzer, Ripart et al., *BRAIN* 2022

David et al., *Epilepsia* 2021

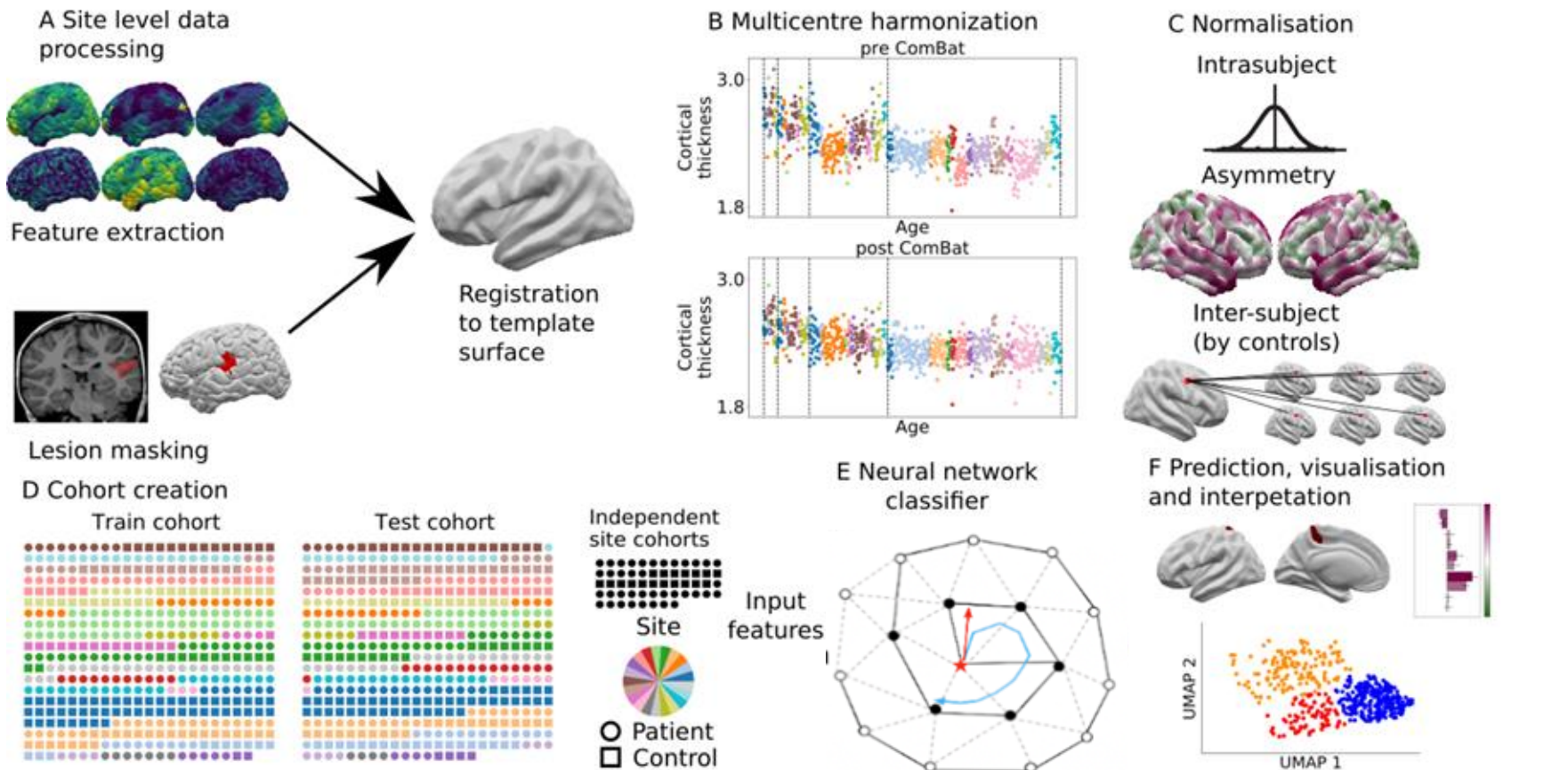
Gill et al., *Neurology* 2021

Walger et al., *Epilepsia* 2023

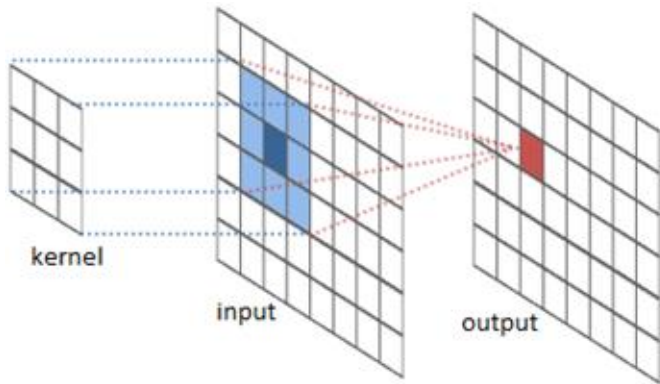
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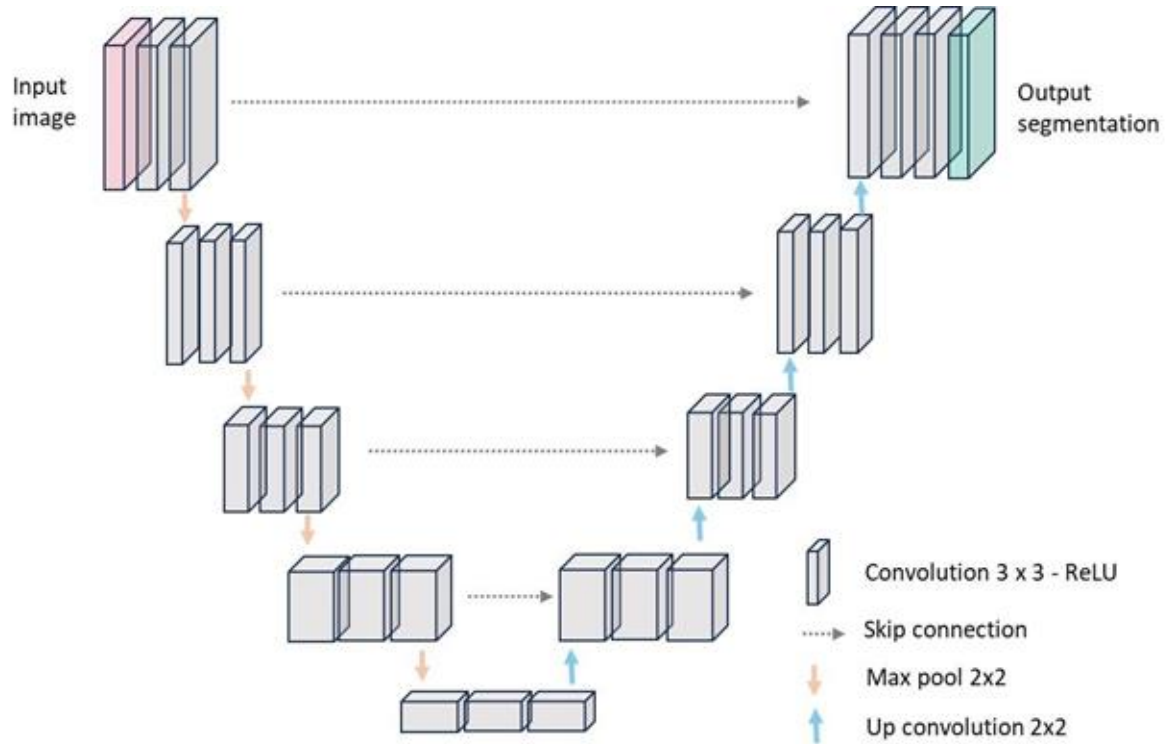
Automated lesion detection - MELD Graph



Parenthesis: Convolutional Neural Network for image segmentation

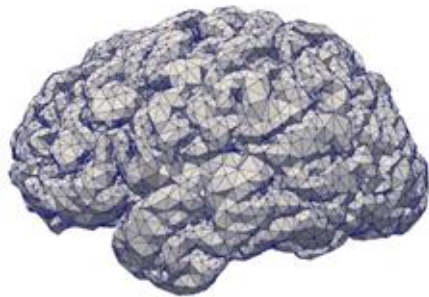


Convolution 3 x 3

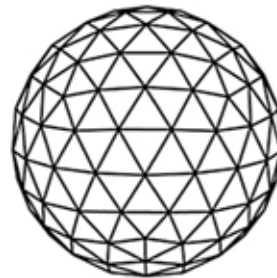


Example of the U-Net

A Graph Convolutional U-Net for surface segmentation

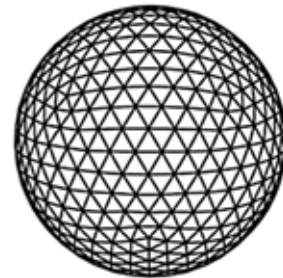


S_0



S_1

...



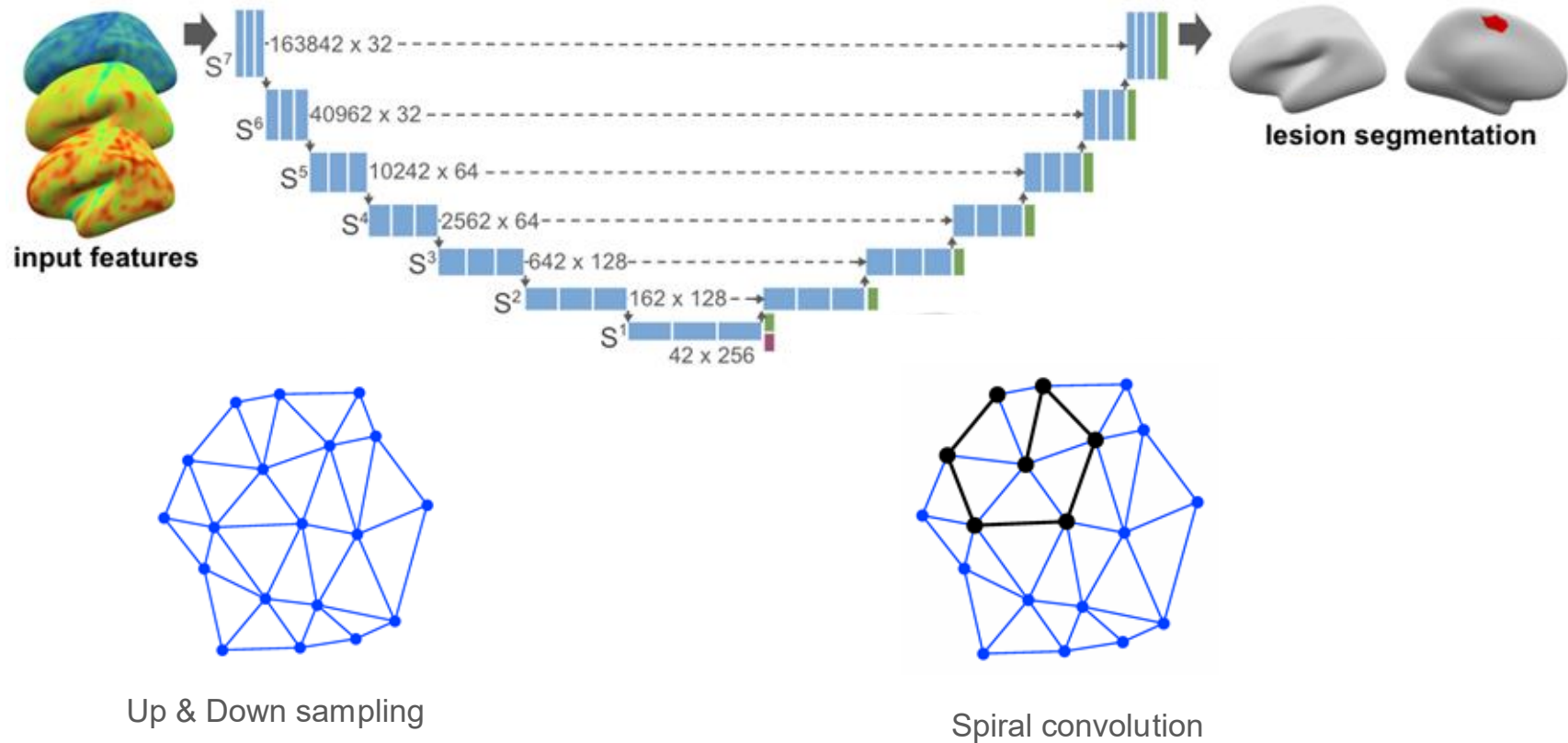
S_7

We created meshes at different resolutions based on icosahedral icospheres S_i

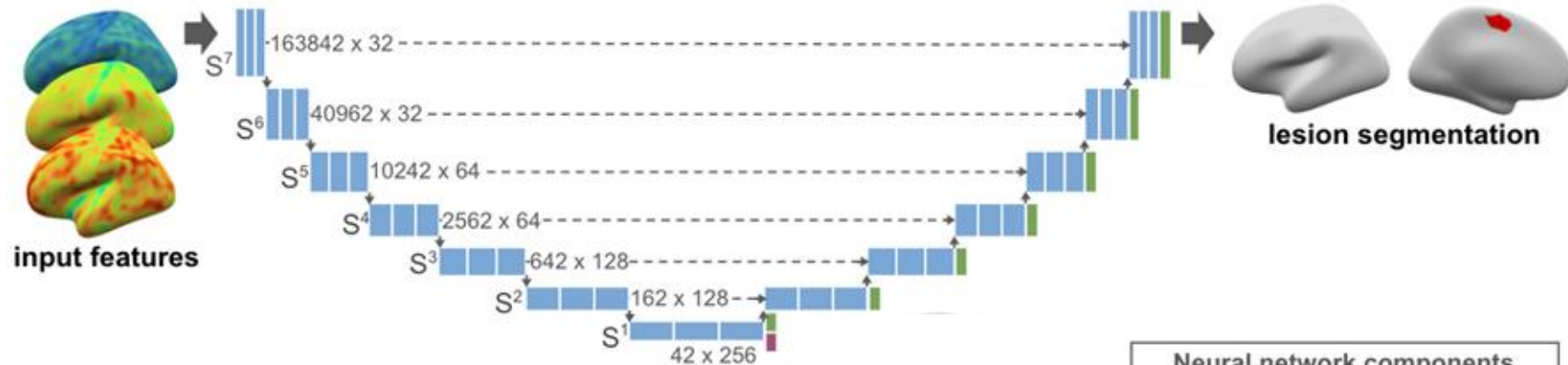
Advantage:

- Symmetrical
- Easy to resample to different resolutions

A Graph Convolutional U-Net for surface segmentation



A Graph Convolutional U-Net for surface segmentation

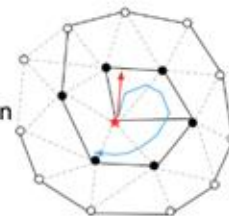


Neural network components

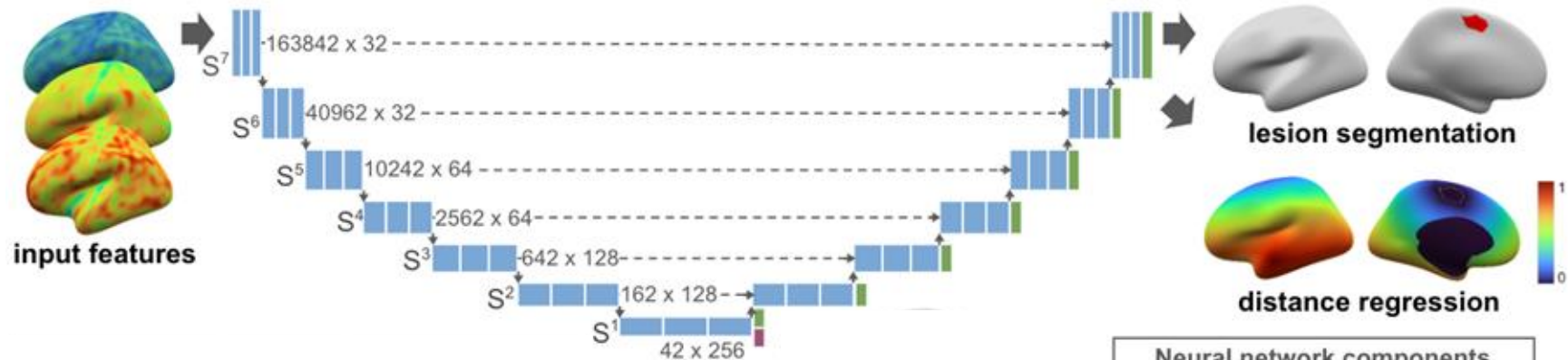
spiral convolution

softmax for segmentation & distance

fully connected layer + softmax for hemisphere classification



A Graph Convolutional U-Net for surface segmentation

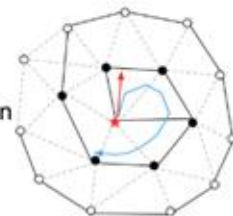


Neural network components

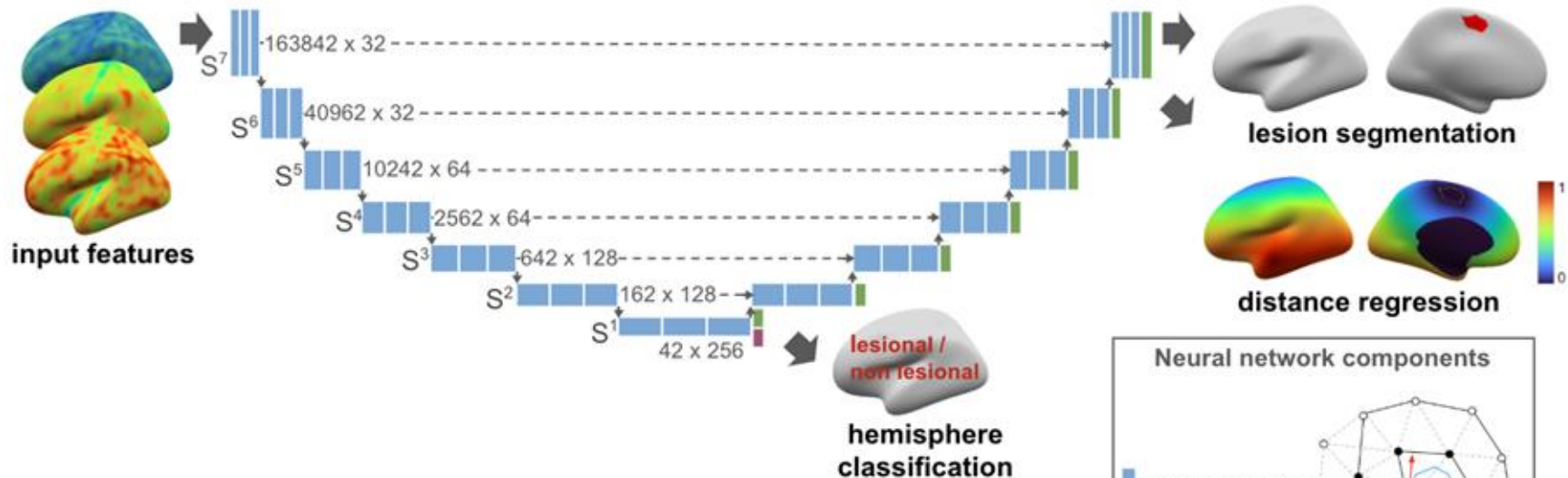
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A Graph Convolutional U-Net for surface segmentation



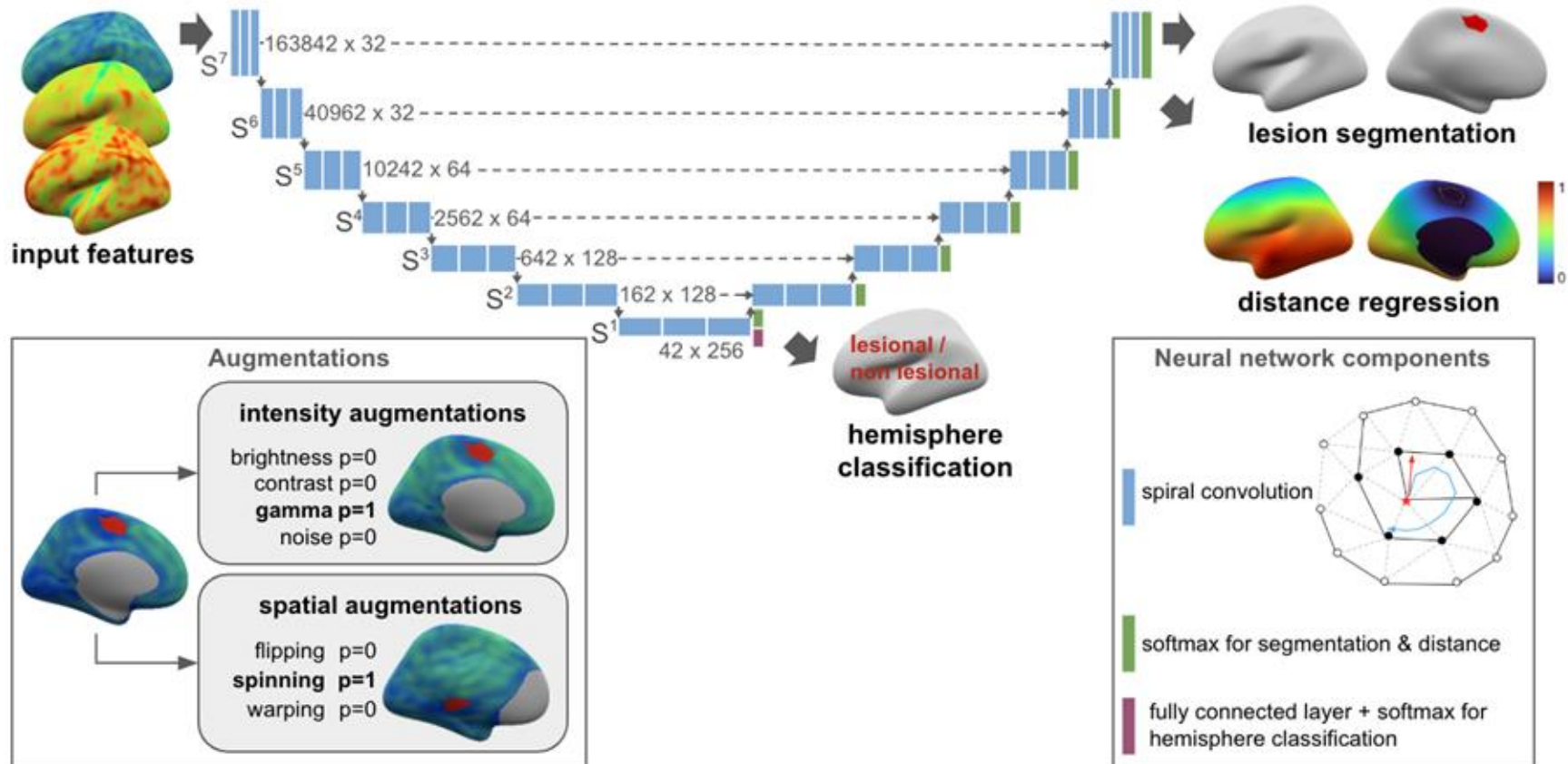
Neural network components

spiral convolution

softmax for segmentation & distance

fully connected layer + softmax for hemisphere classification

A Graph Convolutional U-Net for surface segmentation



A Graph Convolutional U-Net for surface segmentation

Table 1. Experiments

Experiment Name	Description
MLP [10]	vertex-wise multilayer perceptron
GC-nnU-Net	graph-based adaptation of nnU-Net
GC-nnU-Net+c	adding classification loss
GC-nnU-Net+d	adding distance loss
GC-nnU-Net+dc	adding distance loss and classification loss

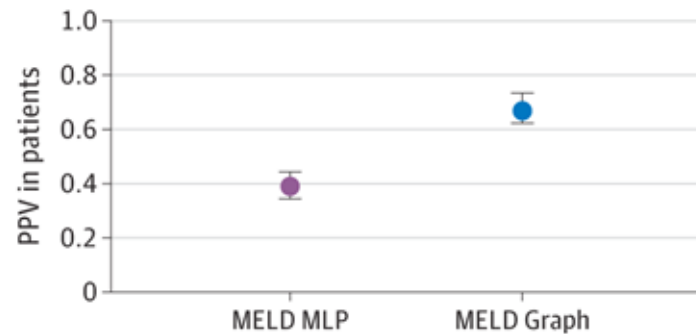
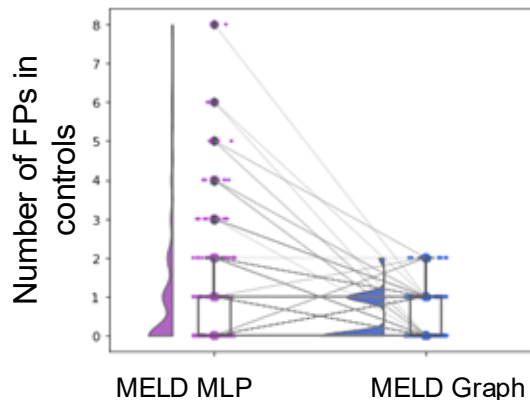
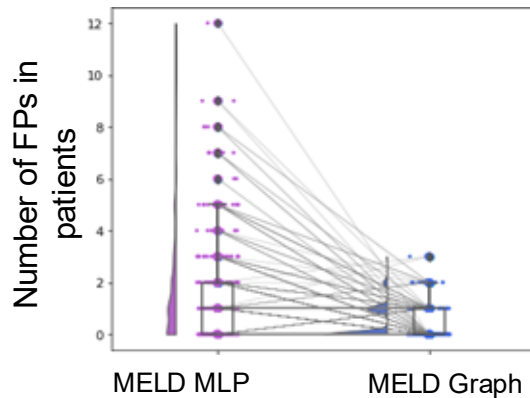
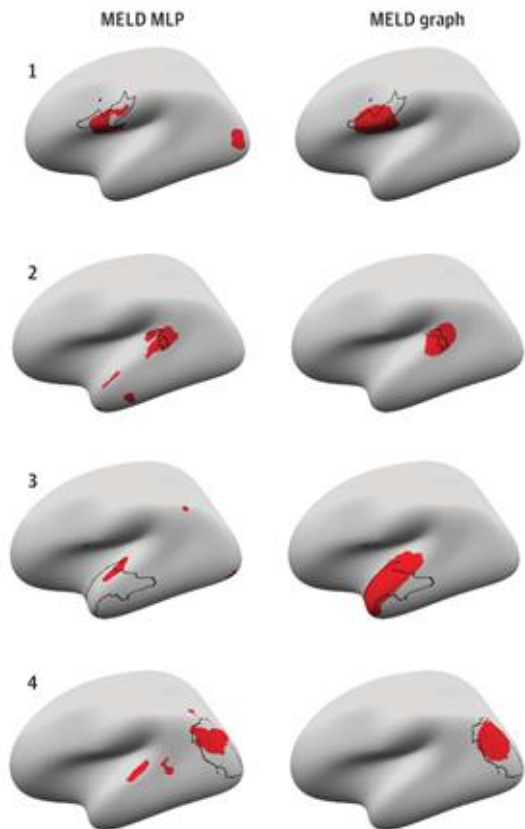
Table 2. Comparison of models on the test dataset.

Experiment	AUC (+/- std)	Sensitivity	Specificity	Run time (min)
MLP [10]	0.64 (n.a.)	67%	49%	n.a.
GC-nnU-Net	0.68 ^{†‡} (+/- 0.004)	67%	64%	396.1
GC-nnU-Net+c	0.74 [†] (+/- 0.008)	67%	66%	373.9
GC-nnU-Net+d	0.69 ^{†‡} (+/- 0.007)	67%	65%	426.9
GC-nnU-Net+dc	0.74 [†] (+/- 0.005)	67%	71%	564.9

†Model performance significantly improved compared to MLP.

‡Model performance significantly worse compared to GC-nnU-Net+dc.

Comparison of MELD MLP and MELD Graph

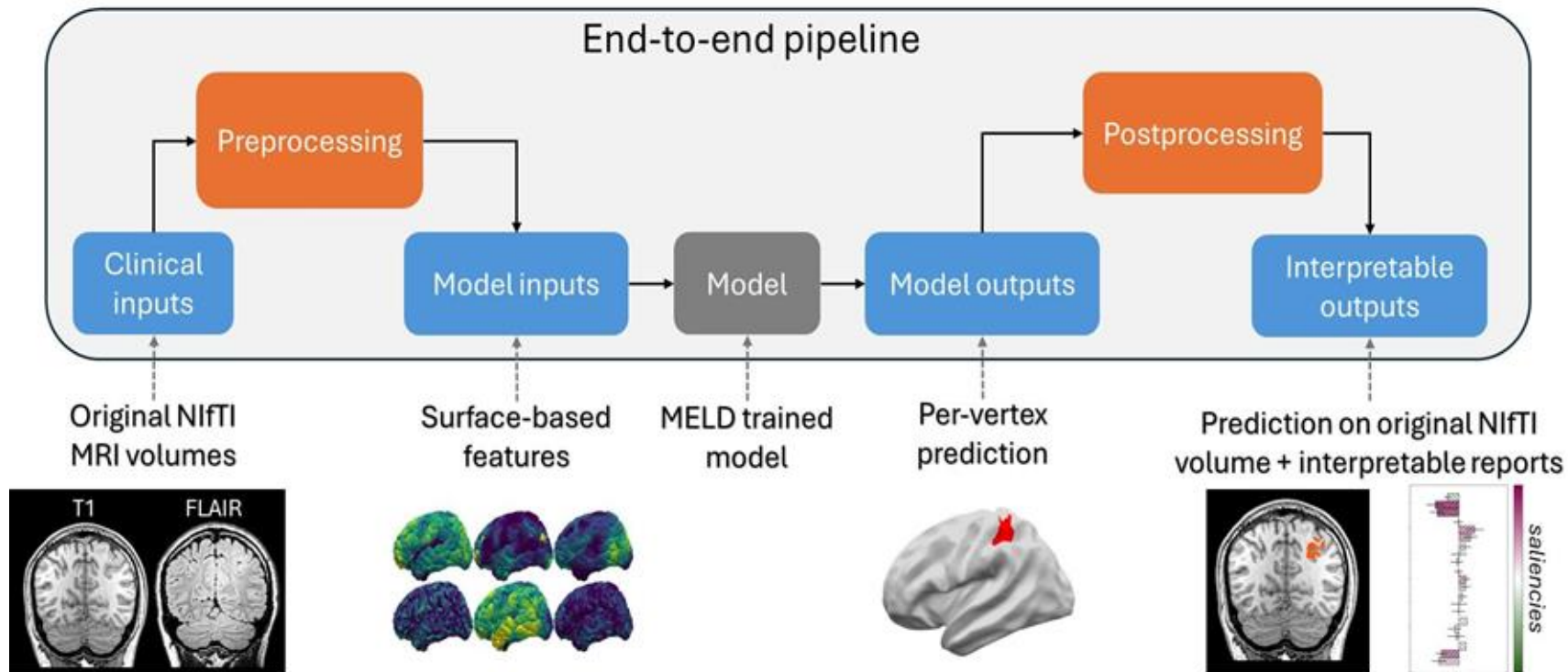


	MELD MLP	MELD Graph
Sensitivity	67%	72%
Specificity	54%	60%
PPV	39%	69%
Detection in MRI negative	63%	65%

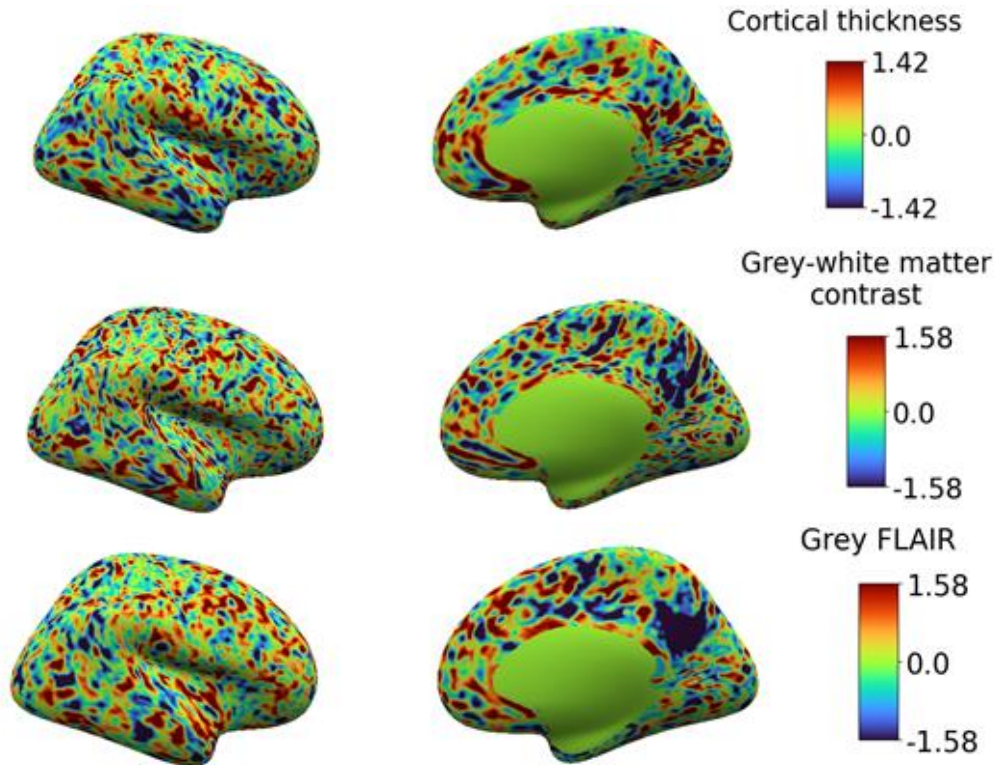
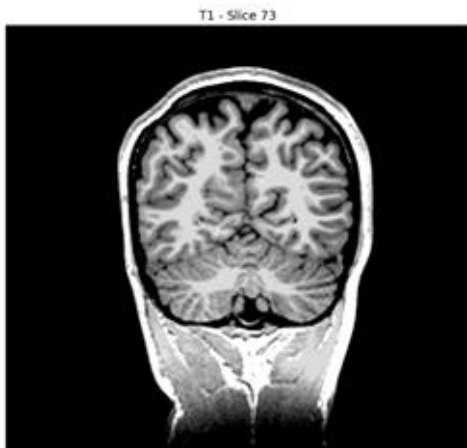
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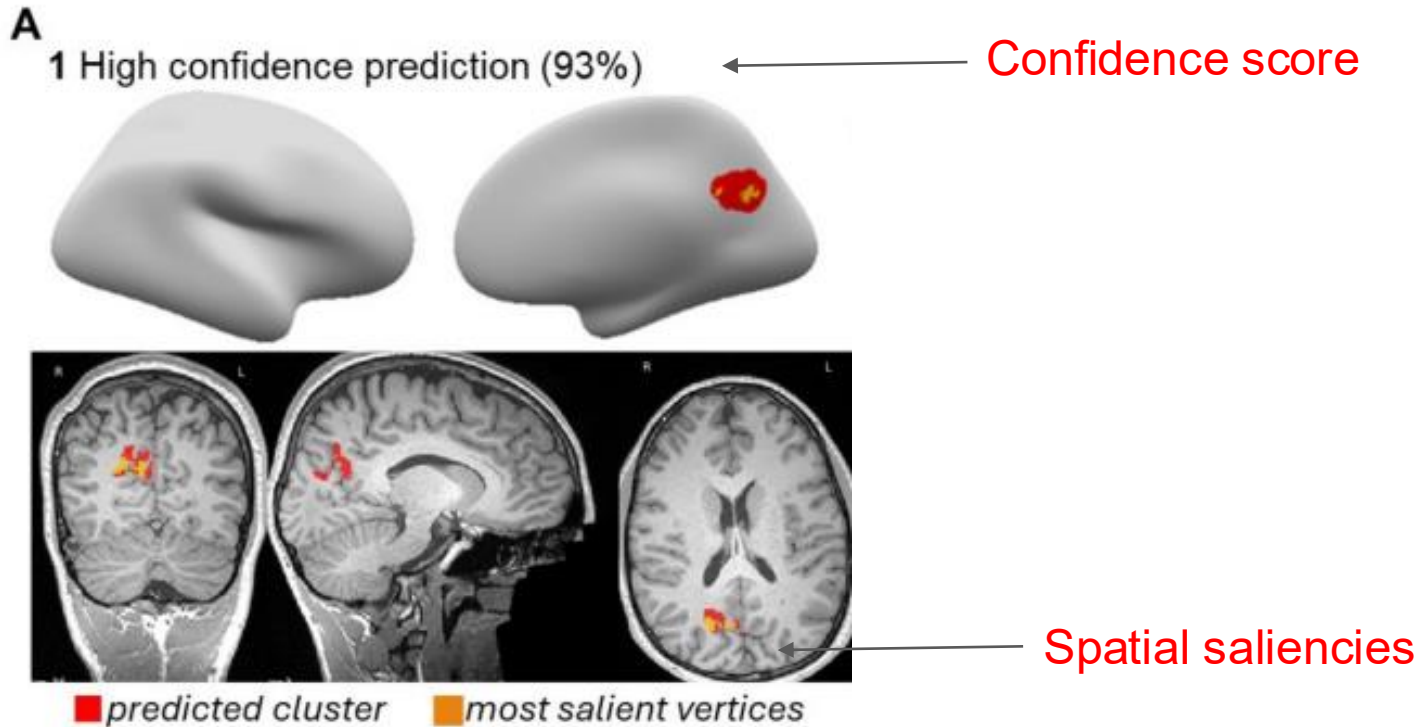
Integrating the model into an end-to-end pipeline



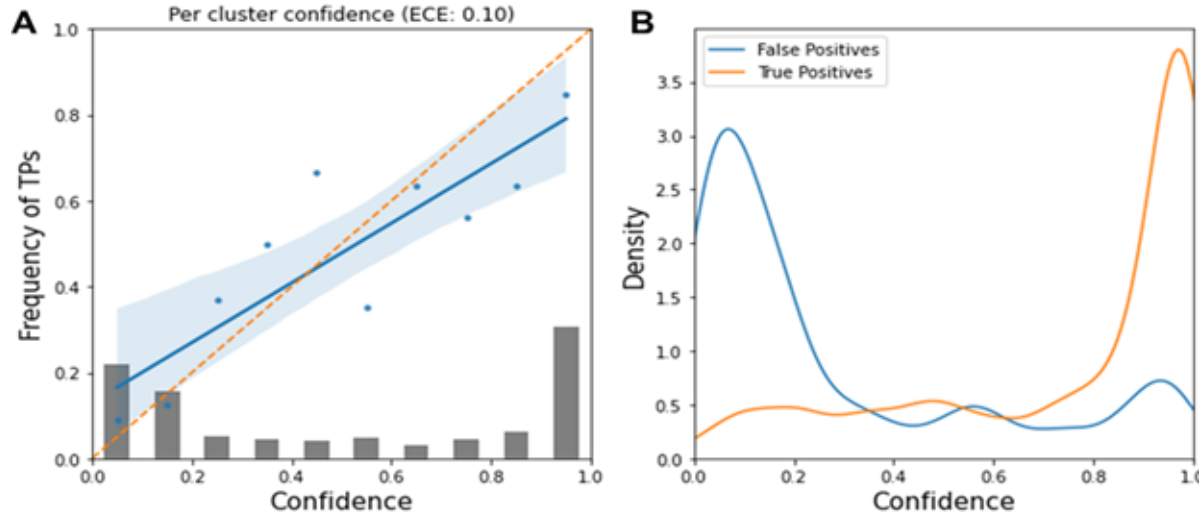
Can you find the lesion: back to the example



Can you find the lesion: Interpretable report

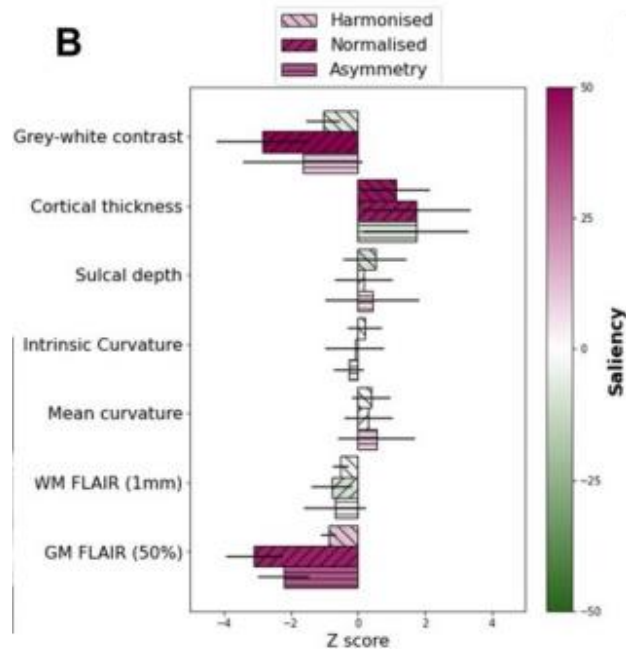


Calibration of confidence scores



Confidence scores = average of prediction scores for each vertex in the predicted lesion

Can you find the lesion: Interpretable report



← Feature saliencies

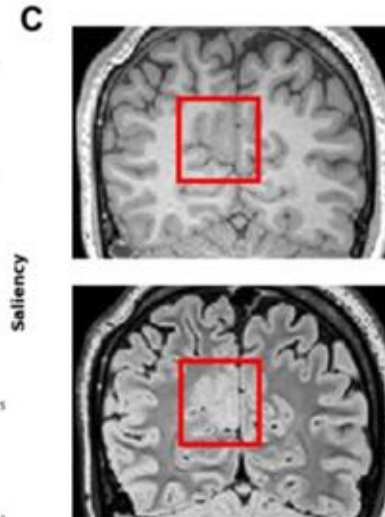
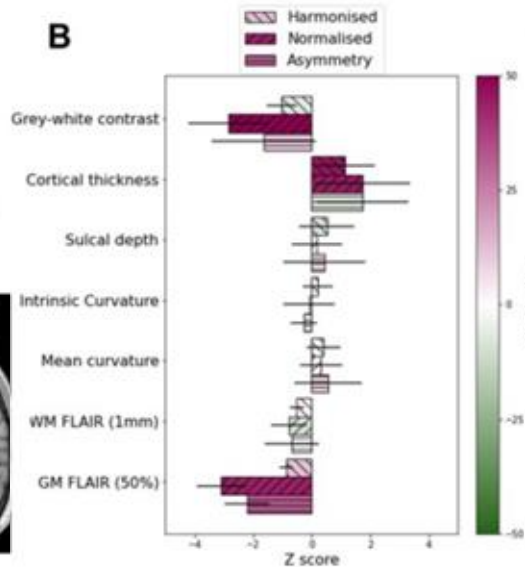
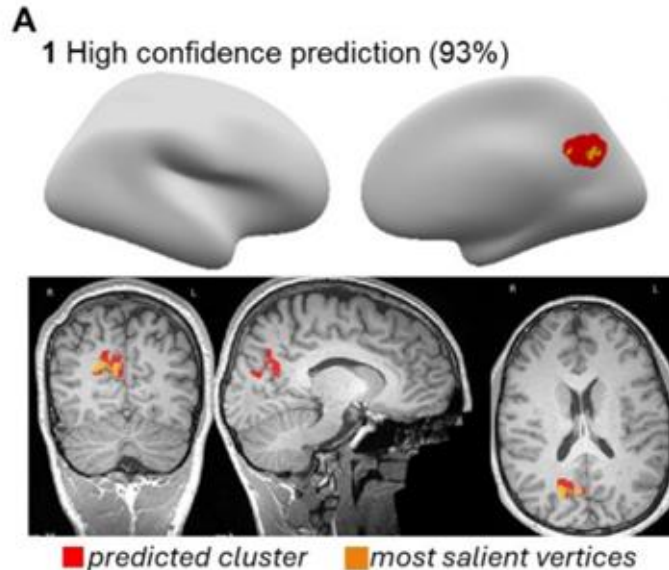
Can you find the lesion: Interpretable report

Q4. How confident are you about the lesion location?

1. Absolutely no idea
2. Everything looks a bit abnormal
3. One or two areas look suspicious
4. I think I know where it is
5. I am very confident



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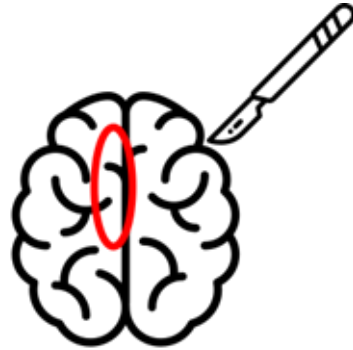


Translation into clinical practice

1. Neuroradiology



2. Neurosurgery



3. Global training



Disclaimer: **MELD Graph** is not a regulated medical device

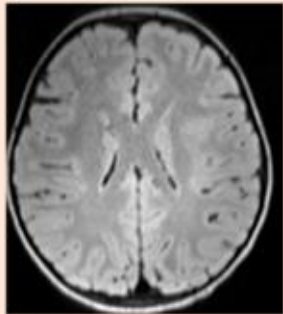
Translation into clinical practice: GOSH radiologists



Patient

8 year old girl
Referred to specialist
epilepsy surgery center
Drug resistant epilepsy

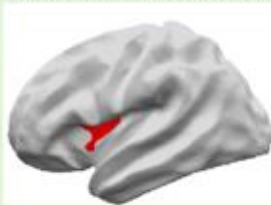
MRI negative



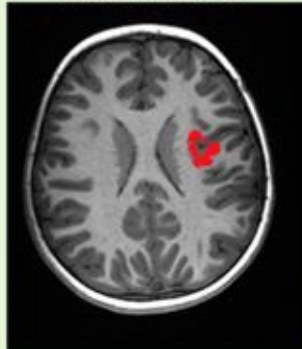
Original FLAIR MRI
No lesion detected

MELD FCD algorithm

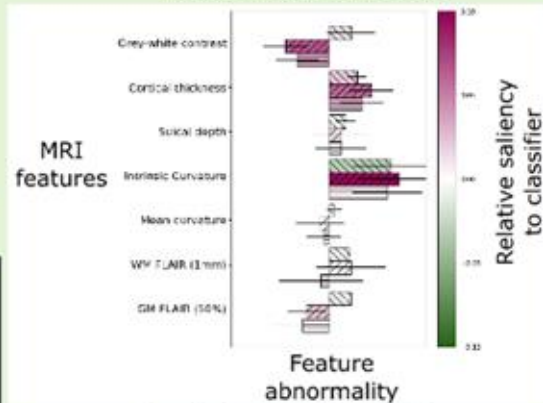
Surface-based view



Axial MRI view



Lesion feature report



Focal abnormality (red)
detected by MELD algorithm
and confirmed by radiologist.
Algorithm identified lesion from:
grey-white blurring,
increased cortical thickness
and folding abnormalities



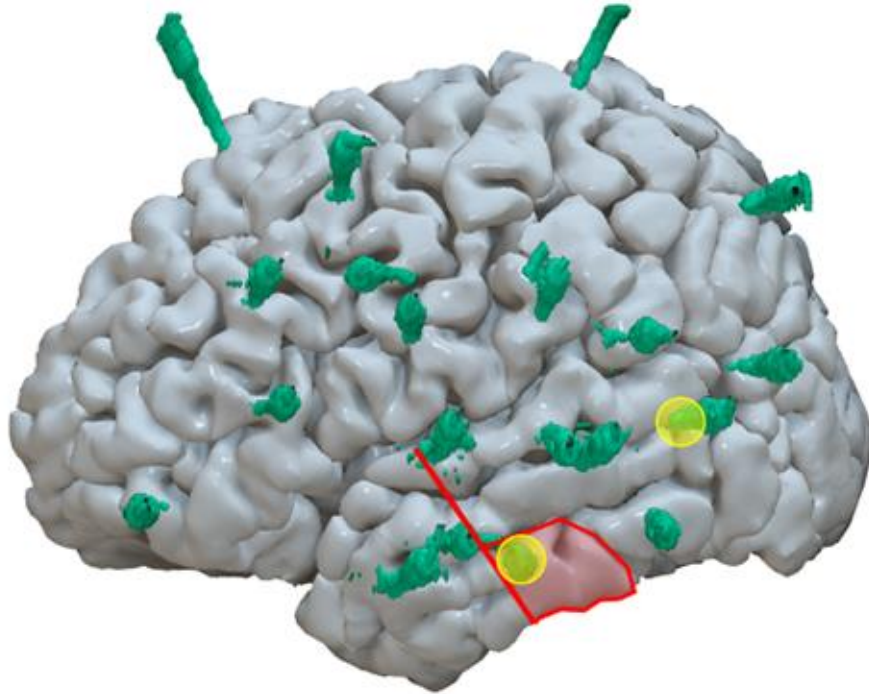
Dr Felice D'Arco



Dr Kish Mankad

Pastore et al.,
Neuroradiology 2024

Translation into clinical practice: GOSH neurosurgery



M . A . S . T

MELD as an Adjuvant for
SEEG Trajectories



Dr Aswin Chari



Dr Martin Tisdall

Chari et al., DMCN 2023

Translation into clinical practice: global training



Open source on GitHub

github.com/MELDProject



Documented on Readthedocs

MELD classifier Documentation

Search docs

INSTALLATION

- Docker container
- Native installation
- Singularity container

MELD FCD PREDICTION PIPELINE

- Prepare the data
- Compute the harmonisation parameters for a new scanner
- Predict FCD lesion on MRI data
- Interpretation of the MELD Graph pipeline results

OTHER USAGE

- Notebooks to recreate figures
- Training and evaluating a graph model
- Contributing guide

/ MELD Graph View page source

MELD Graph

Full documentation: [here](#)

Intro to MELD Graph and installation videos: [here](#)

Graph based FCD lesion segmentation for the MELD project.

This package is a pipeline to segment FCD lesions from MRI scans.

A Dataset
 $n_{patients}=703$
 $n_{controls}=482$
23 centres worldwide

B Training
Input features
MELD Graph

C Testing
Prediction vs Ground truth
Multilayer Perceptron
MELD Graph

E Interpretable output
Cluster size = 10.5 cm³
Confidence = 93%



YouTube Tutorials



MELD project

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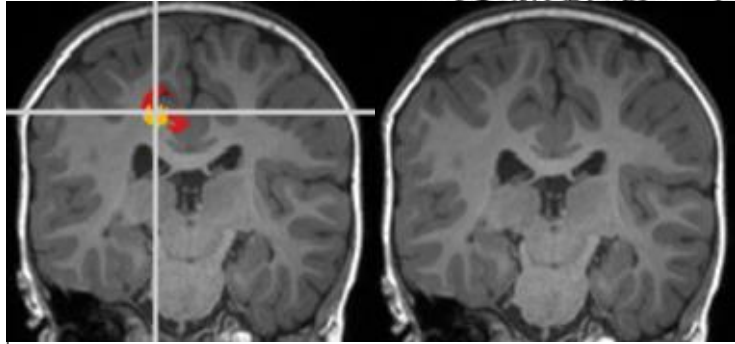
Next one:

ILAE School on Neuroimaging,
May 2026, Potsdam

MELD: cases from around the world



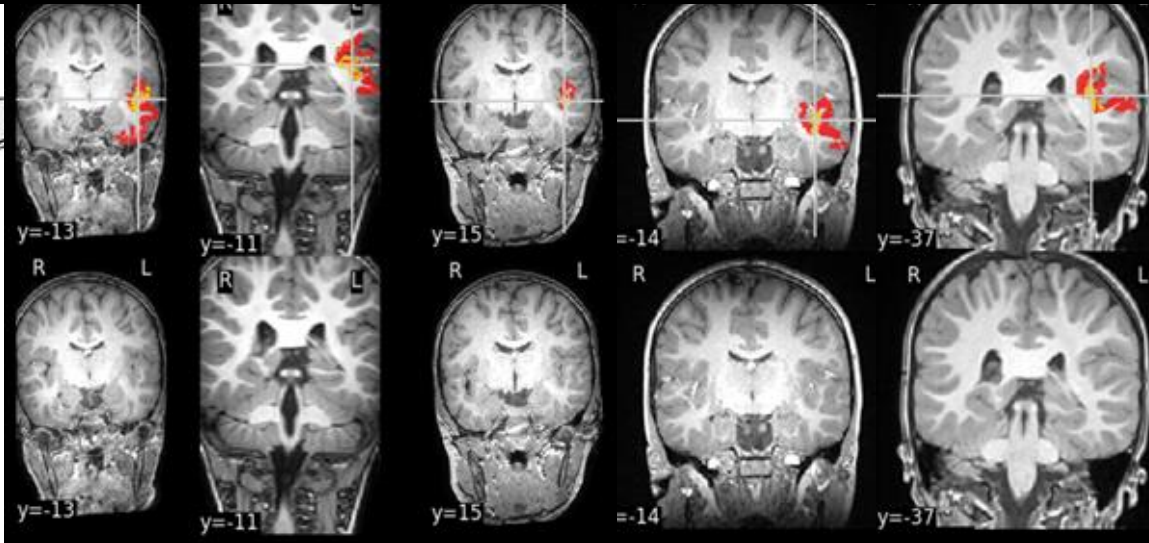
MELD: cases from around the world



20 seizures per day, MRI negative.
MELD cluster concordant with video telemetry
Thermocoagulated following sEEG
Child seizure free for a month after
A focal resection is planned.
Bristol Royal Hospital for Children, UK
Dr Marcus Likeman



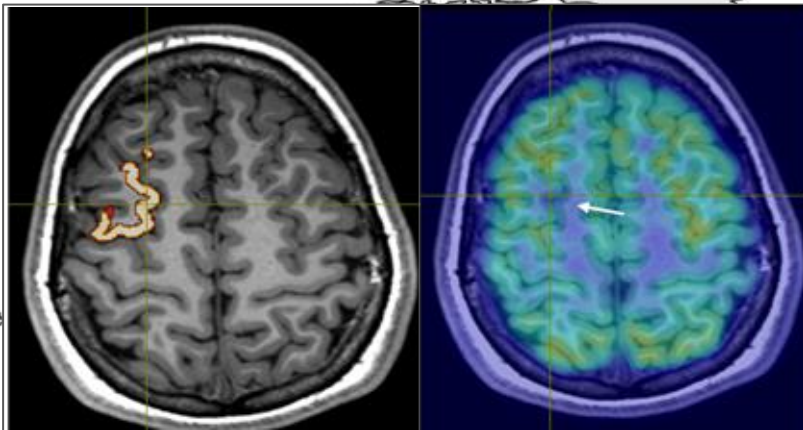
MELD: cases from around the world



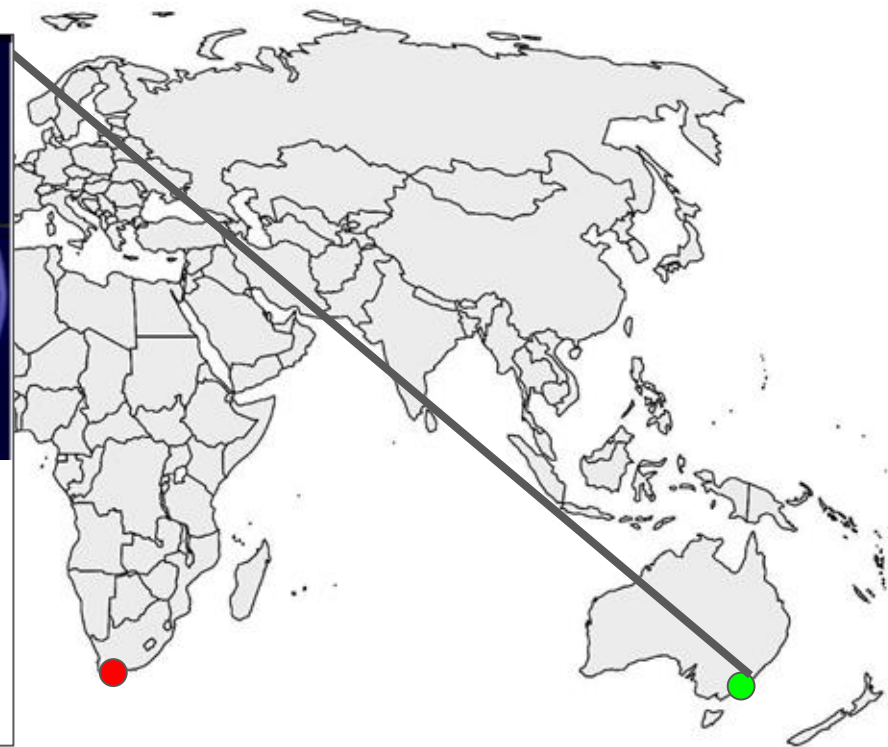
10yr old male. Up to 70 seizures/day. 7x MRI negative. MELD confirmed by neuroradiologist. **Mediclinic Constantiaberg, South Africa. Dr James Butler**



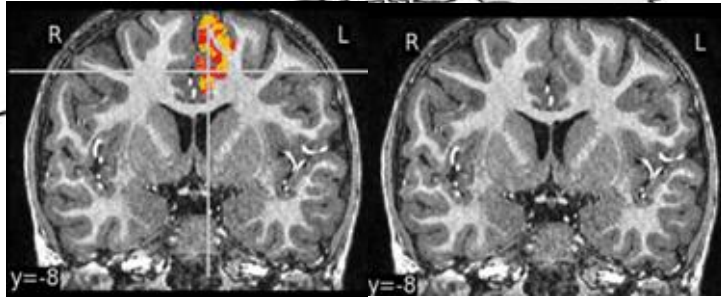
MELD: cases from around the world



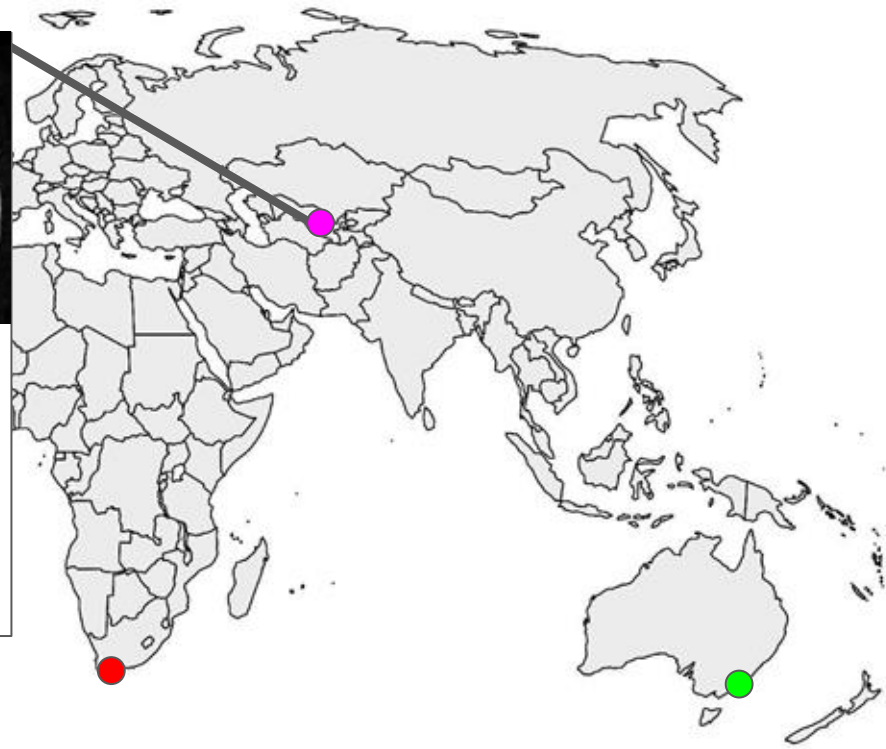
14 y old girl.
x3 MRI negative
MELD + PET: R frontal BOSD
Resective surgery - FCDIIA, seizure free off meds 6m
The Royal Children's Hospital, Melbourne
Dr Emma Macdonald-Laurs, Dr Aaron Warren



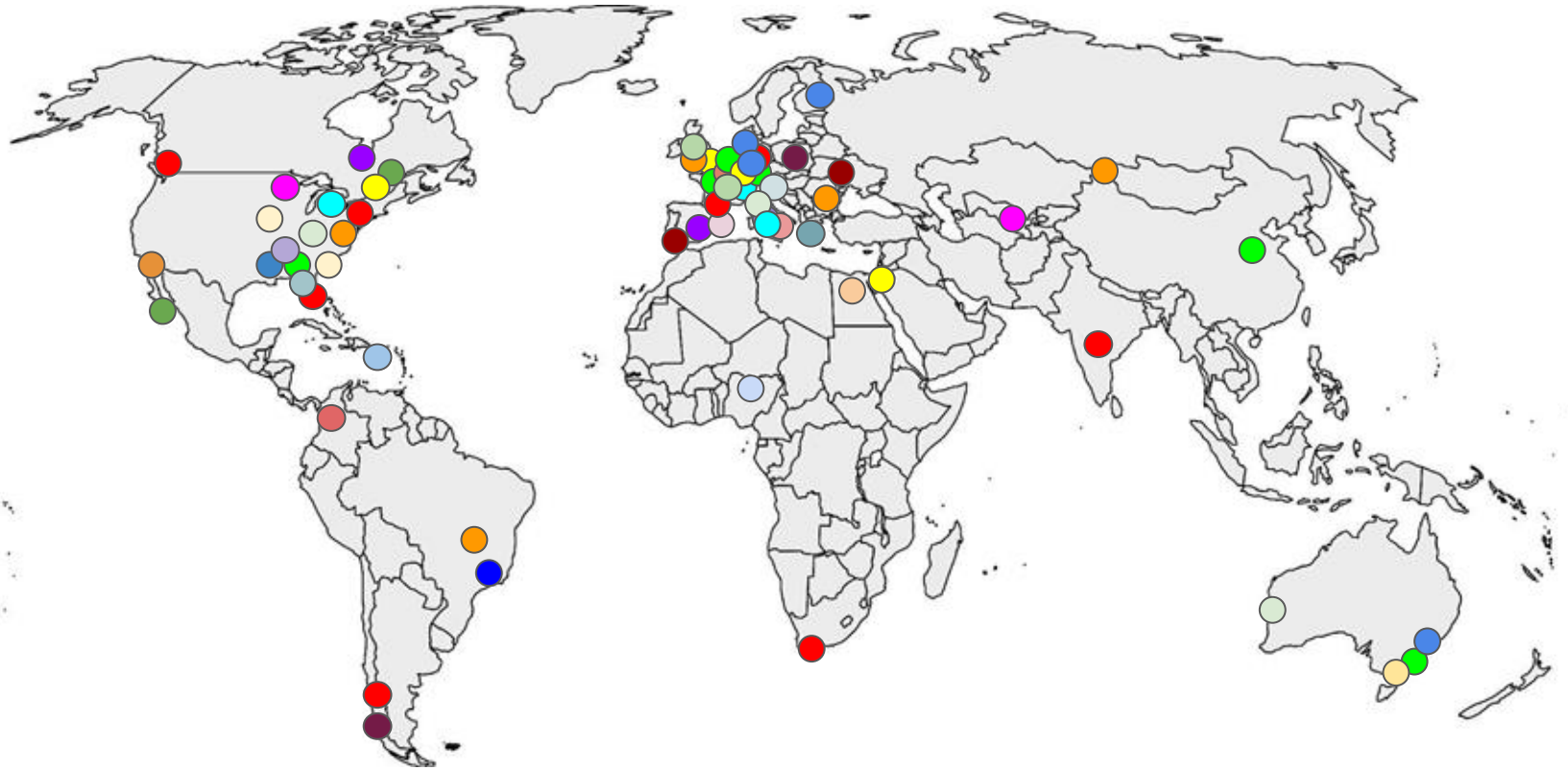
MELD: cases from around the world



4 y old, EEG - unclear, Non-localising
semiology: spasm, lip smacking
MRI negative
MELD cluster confirmed by Dr Felice D'Arco
**National Children's Medical Center,
Tashkent, Uzbekistan**
Prof Furkat Samadov



MELD: cases from around the world



Wider research impact



Dr Cornelius Kronlage

Can we use **ultra-high field MRI** for automated detection of focal epilepsy lesions ?



Collaborations:

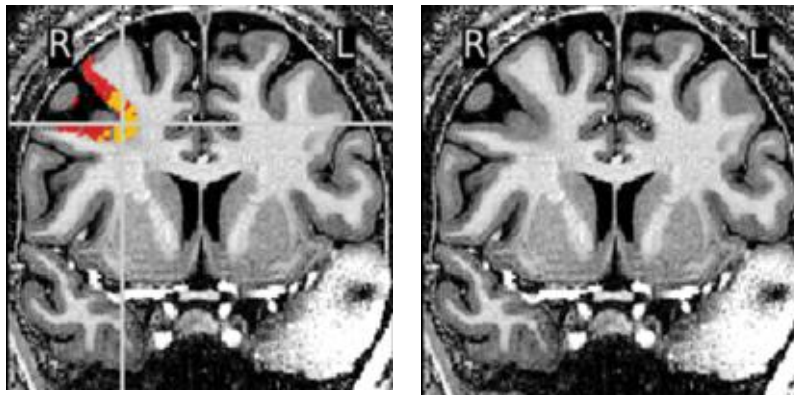
University of **Tübingen** (Martin, Hagberg, Scheffler)

University of **Bonn** (Rüber)

KCL (O'Muircheartaigh, Leary)

UCL (Piper, D'Arco)

University of **Cambridge** (Rodgers, Cope)



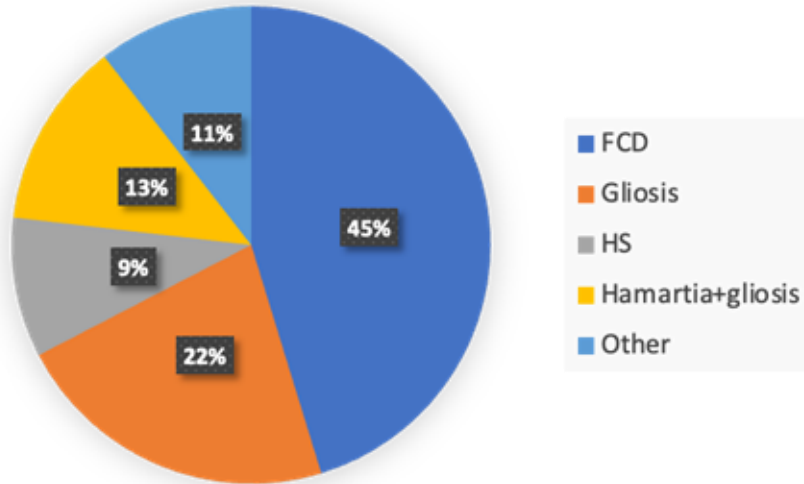
MELD-graph prediction on 0.8 mm 9.4T MP2RAGE

Key points of this talk

- AI can detect subtle focal cortical dysplasia (FCD) on MRI
- Graph Neural network for a more confident detection of FCD
- Translation into clinical practice
- Going further - a deep learning model for the detection of multiple epilepsy lesions

“MRI-negative” Epilepsy - “Missed” lesion

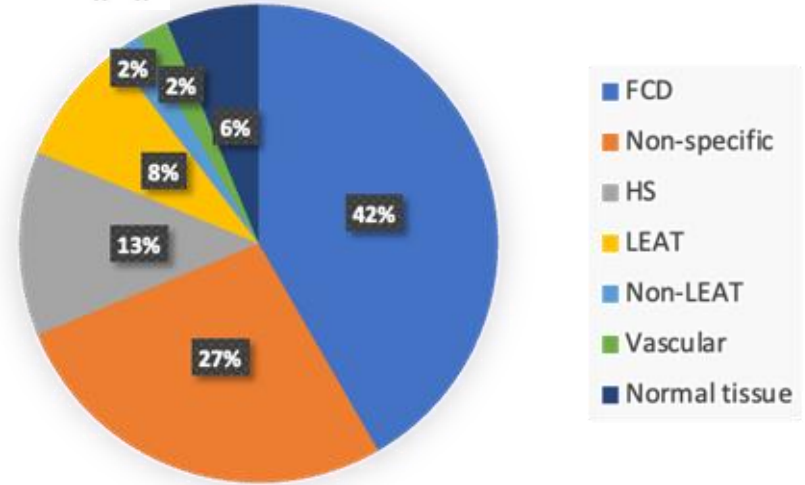
 Adults



95 MRI negative

Adapted from Wang et al., 2013 Mod Path

 Children



87 MRI negative

Eriksson et al., 2023 Epilepsia

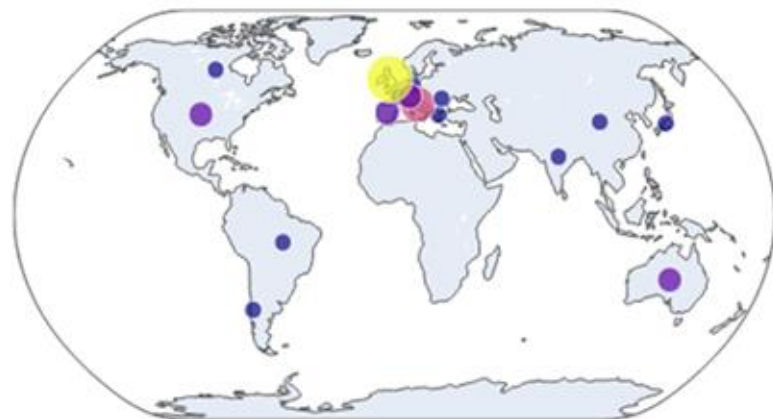
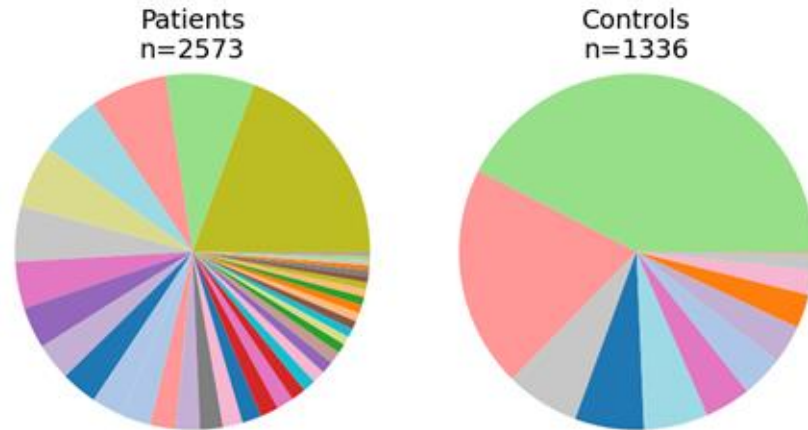
MELD Focal Epilepsies dataset

39 centres - 3909 subjects

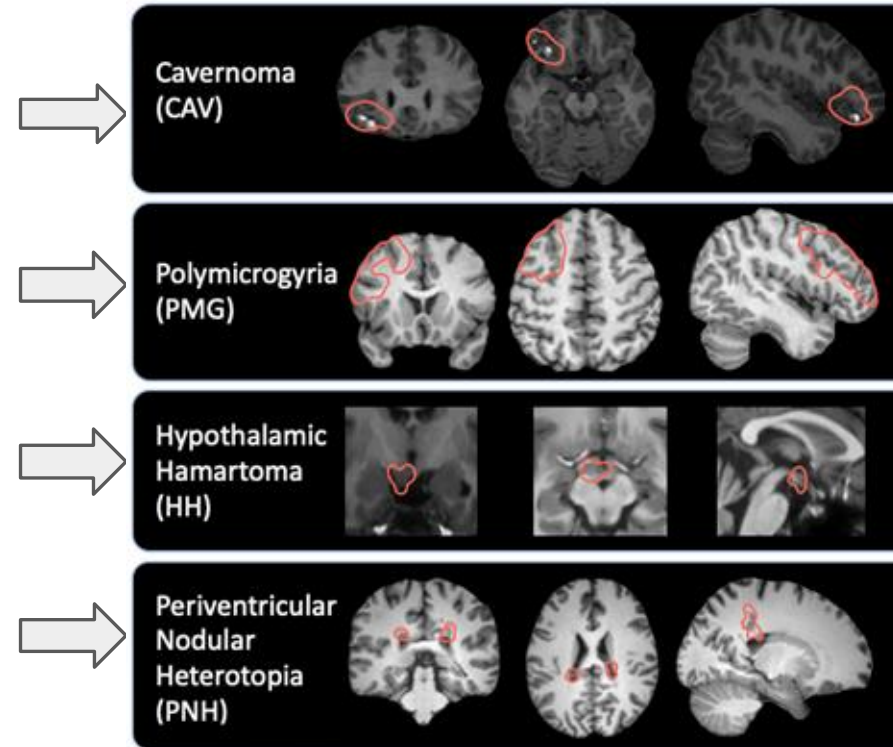
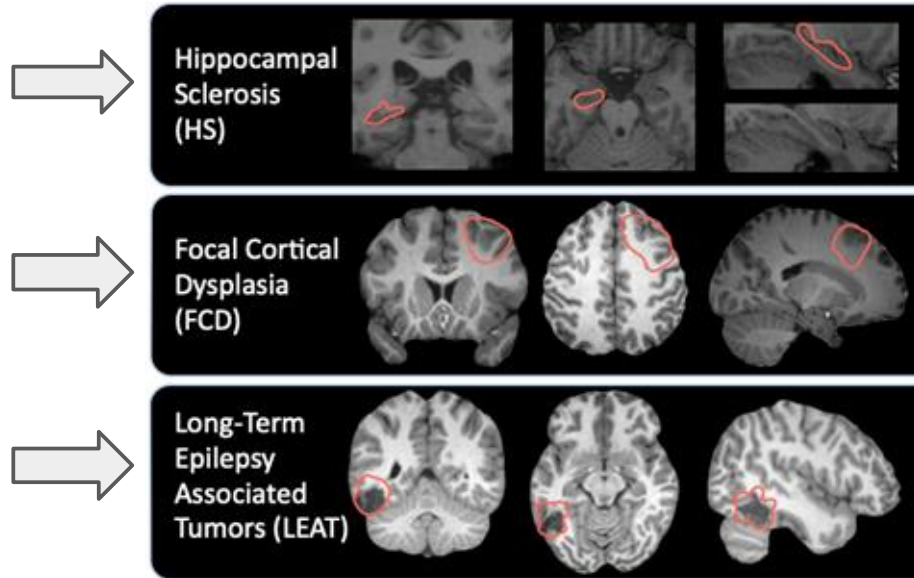
Over 2500 patients and 1300 controls

Large age range (0.1 - 69 years)

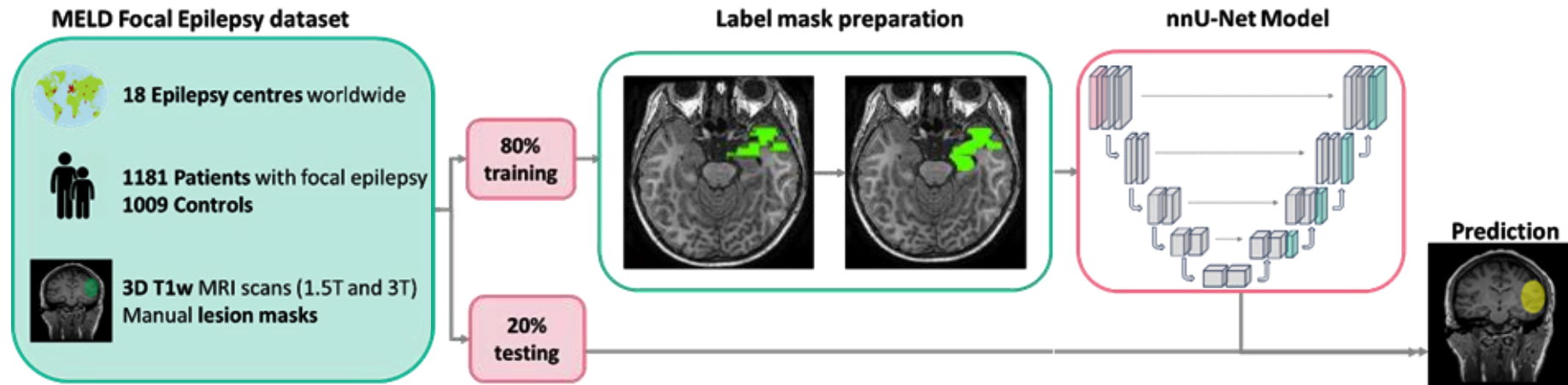
Multiple focal epilepsy pathologies



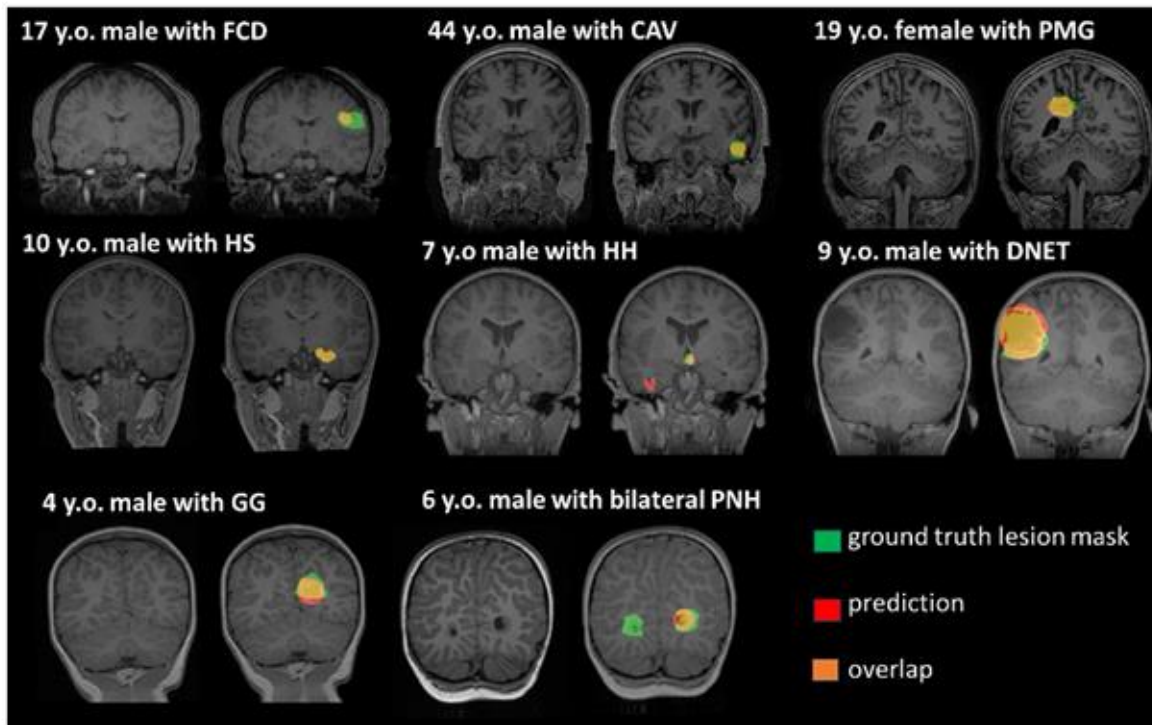
Focal epilepsies lesion location



MELD Find: Focal-epilepsy Imaging Network Detection

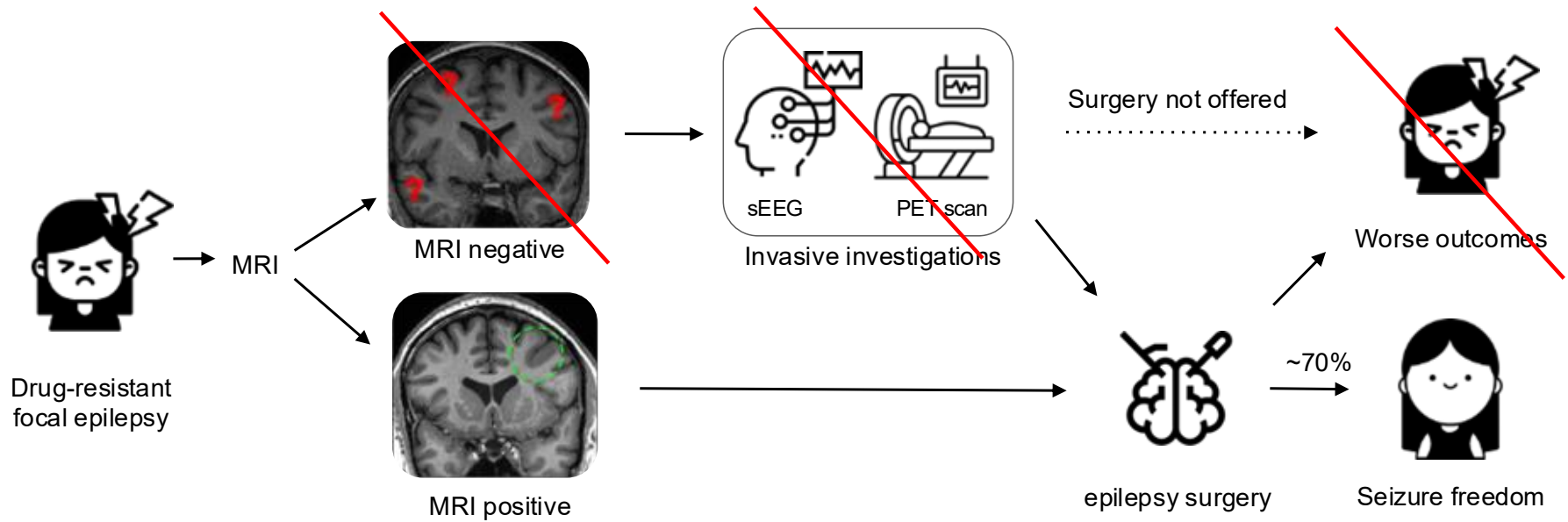


MELD Find: Focal-epilepsy Imaging Network Detection



73% sensitivity
90% specificity

Vision: A better chance of seizure freedom for all






Vision: A better chance of seizure freedom for all

 Earlier referral & faster diagnosis

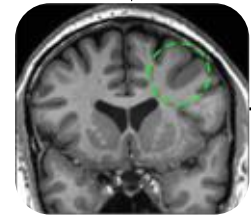
 Automated detection

 Help plan Intracranial EEG

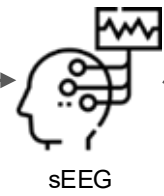
 Help refine surgery resection


Drug-resistant focal epilepsy

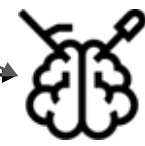
MRI



MRI positive



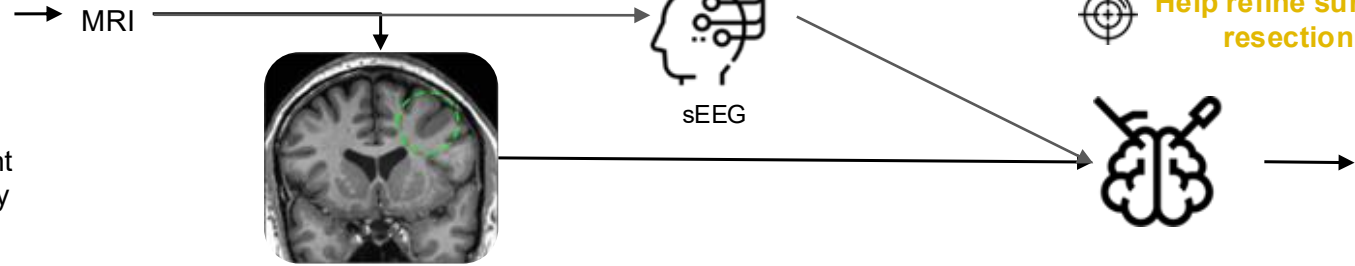
sEEG



epilepsy surgery



Seizure freedom



Acknowledgements



The **IMAGINE** Lab
led by



Dr Konrad Wagstyl



Dr Sophie Adler



MELD collaborators
& MELD tools users



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