



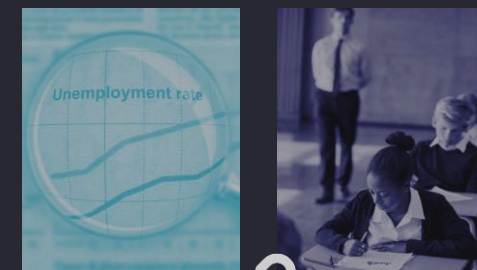
From capability to research: Meet the DARE UK Real-world Research Exemplars

Najmeh Modarres, Programme Manager

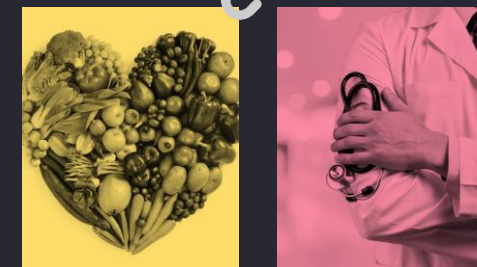
14 May 2026



Sensitive data: affecting people, communities, and populations...



Data linkage: safely connecting sensitive data to benefit people, communities and populations...



Vision

For all research and innovation to benefit from seamless, secure use of diverse sensitive data, at a pace, efficiency and scale approaching that of the open data ecosystem, that revolutionises research productivity and accelerates research to deliver public good

Mission

To put the UK at the forefront of sensitive data research and innovation by assembling the tools, technologies and standards needed to streamline secure data linkage and use

FUNDED BY

UK Research and Innovation (UKRI) – the UK’s largest public funder of data and research innovation – through the [UKRI Digital Research Infrastructure programme](#).



DELIVERED BY

[Health Data Research UK \(HDR UK\)](#) and [Administrative Data Research UK \(ADR UK\)](#).



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**Phase 1:
Design & Dialogue**

July '21 – March '24

£7.5m

**Phase 2:
Build, Test & Establish**

August '24 – March '27

£20.0m

**Phase 3:
Deliver, Optimise & Federate**

TBC



Phase 2



Partners
Build | Develop



TREs / SDEs
Test | Configure | Adopt



Researchers
Do science | Deliver impact



Community
Sharing | Best practice | Open science



Innovators
Test | Proof-of-concept | Pilot



Public
Involvement | Engagement



 Phase 2



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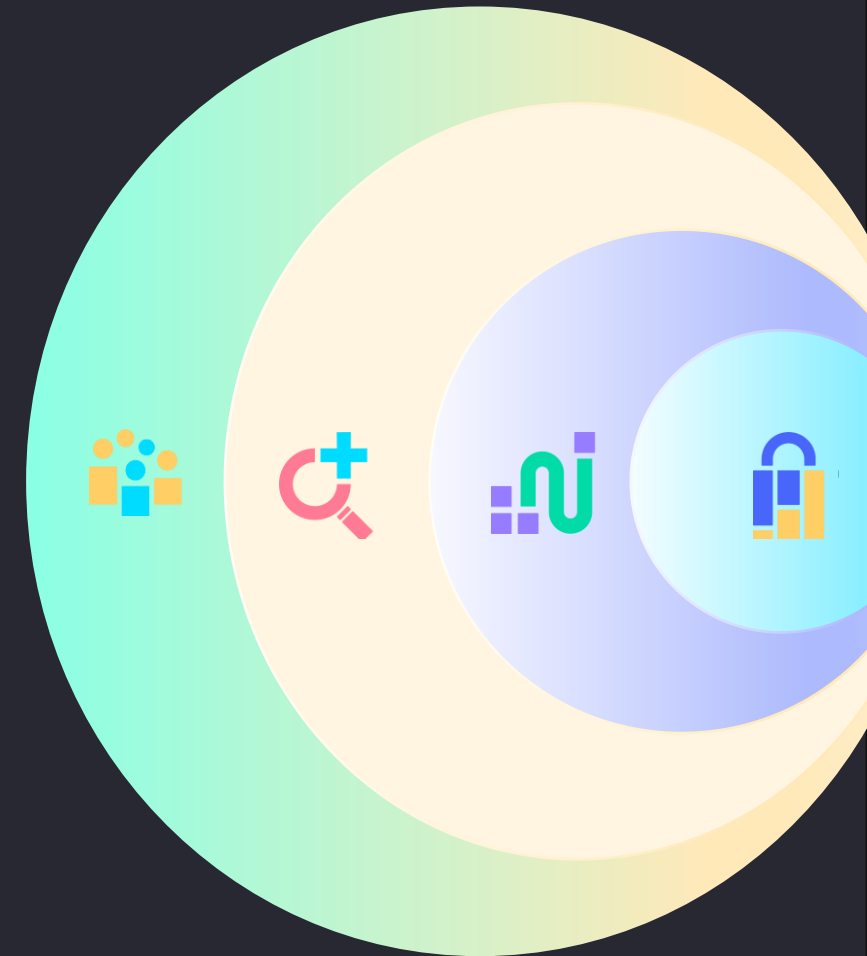
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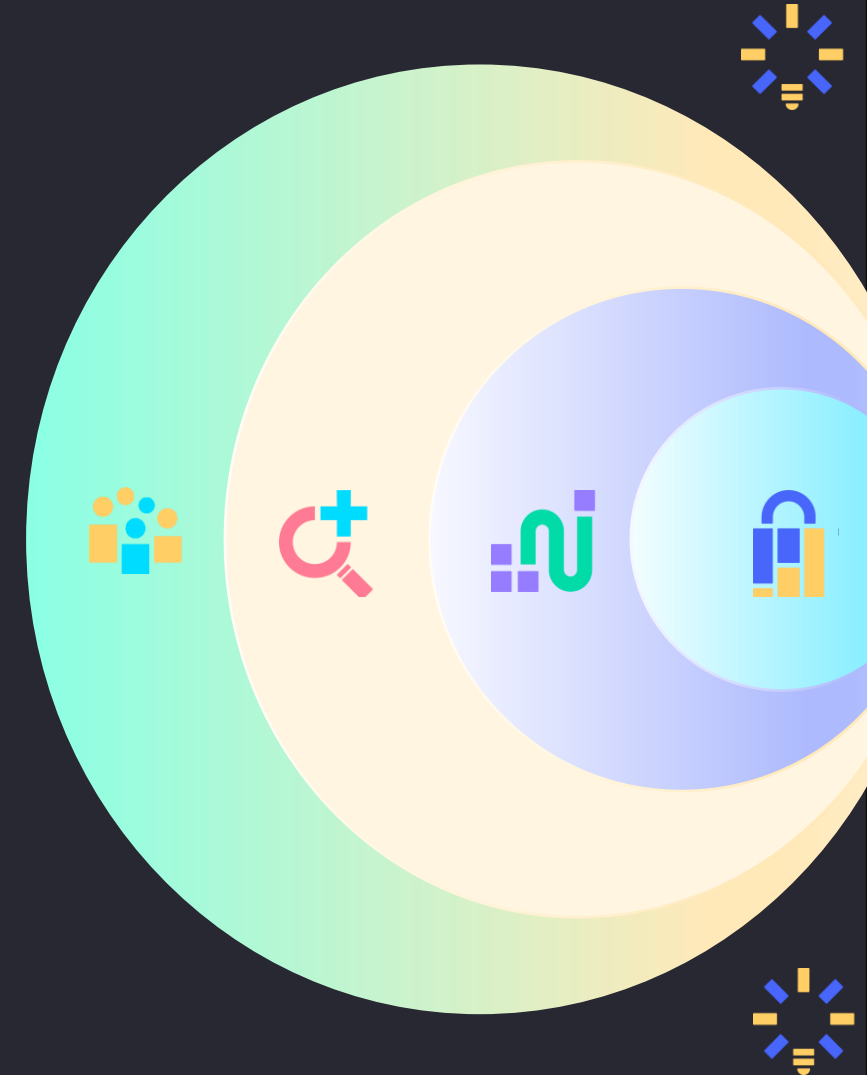
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Innovators
Test | Proof-of-concept | Pilot



Public
Involvement | Engagement





Real-World Exemplars (Mar 2026 – Mar 2027)

- **AIRR-BRIDGE: Bridging Research Infrastructure for Data Governance Extension on AIRR**
 - **The Alan Turing Institute**, King's College Hospital NHS FT, UCL, University of Bristol, University of Cambridge, UK Biobank
- **BRAID: Brain Frailty Integrated through Data Federation**
 - **University of Nottingham**, Nottingham University Hospitals NHS FT, Dementias Platform UK (DPUK)
- **CONNECT-AF: Cross-TRE Network for Evaluating Clinical Outcomes and National Trends in Embolic Complications of Atrial Fibrillation**
 - **Swansea University**, SAIL Databank, Eastern England SDE, DPUK
- **FRAME: Federated Research on Anti-hypertensives Maternity Emulation Trial**
 - **Swansea University**, SeRP, CPRD, King's College Hospital NHS FT, Guys and St Thomas NHS FT, University of Edinburgh, University of Birmingham, SAIL Databank
- **HEAL-Scot: Housing, Environment and Location Data Linkage in Federated TREs to Map Geospatial Inequalities**
 - **University of Edinburgh**, DataLoch, Health Informatics Centre
- **MELODY: Federated Machine Learning for Dermatology**
 - **University of Dundee**, Health Informatics Centre (HIC), Thames Valley and Surrey SDE, NHS Tayside, Oxford University Hospitals NHS FT
- **SAFEVID: Spasms Analysis using Federated Learning from Videos Across Multiple TREs**
 - **University of Glasgow**, Health Informatics Centre, West of Scotland Innovation Hub
- **TransPECT: Securing AI NLP-Transformer Models for Safe Release in TREs**
 - **University of Edinburgh**, DataLoch, Health Informatics Centre, King's College London, Public Health Scotland



Now to introduce the
DARE UK Real-world
Research Exemplars....



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Beyond the Airlock: TransPECT, Securing AI NLP-Transformer Models for Safe Release in TREs

Arlene Casey, University of Edinburgh, PI

Richard Walls, University of Dundee, Co-I

Pasquale Minervini, University of Edinburgh, Co-I

Stuart Dunbar, DataLoch (University of Edinburgh / NHS Lothian), PIE Lead



THE UNIVERSITY
of EDINBURGH



DataLoch



UK Research
and Innovation

HDRUK
Health Data Research UK



ADR UK
Data-driven change

**Health
Informatics
Centre**

Public Health
Scotland

KING'S
College
LONDON



Project Team



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DataLoch



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School of Informatics



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Centre



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Health Informatics
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Judit Kuti,
Data Analyst,
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Franz Gruber,
Data Analyst,
DataLoch



John McCullough,
Public Contributor



C,
Public Contributor



Stuart Dunbar,
Engagement Manager,
DataLoch



Fahrurrozi Rahman,
NLP Research Fellow,
DataLoch



Simba Doka,
Administrator,
University of Edinburgh

Natural Language Processing (NLP) Transformer Models – Opportunities

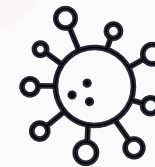
Sally lives alone in a 2-floor detached – skipping diuretic to avoid climbing stairs to bathroom at night. She has forgotten to take BP readings due to tiredness in evenings. Changed medications timing and referred for home health support.

Sensitive free-text data holds rich narratives about people



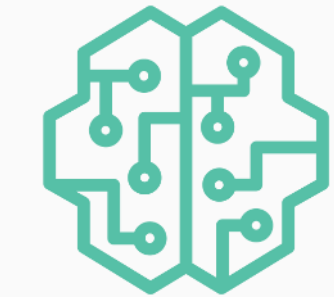
e.g. ChatGPT

NLP Transformer Models



Uses such as Clinical Summaries, Disease Detection, extracting key information

Natural Language Processing (NLP) Transformer Models



NLP Transformer Models for text

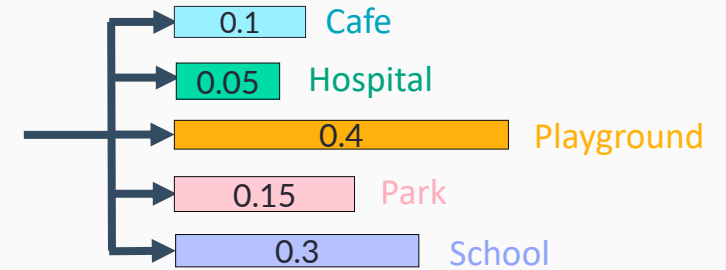


The girl went to the

Previous words (tokens) provide context



Learns patterns of how words appear together



Words being predicted



Challenge: What makes models powerful also means they can memorise tokens and reproduce small fragments of text which can lead to privacy leakage

Sensitive data is de-identified and we have the 5 Safe Frameworks but disclosure control at the safe outputs is not designed for assessing NLP Transformer models



Integration with TRevolution



Our Goals:

Measuring and understanding risks and mitigations for NLP Transformer Models:

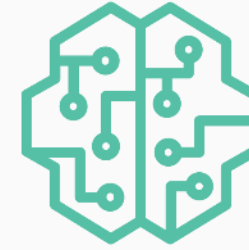
- Support TRE governance teams to make proportionate disclosure decisions
- Provide tools to support disclosure risks
- Help researchers design models in safer ways

Working with SACRO element: Building on gaps identified and aligning to existing tools and frameworks

TransPECT : Work Packages & Deliverables

WP1

- Embed public and stakeholder involvement and governance throughout the project, (4 public members)
- Public tested communications and policy-ready guidance mapped directly to existing frameworks in use within the TRE community



NLP Transformer Models

Understanding & Measuring Risks

Mitigations & Tools

WP2: Systematically measure data leakage giving evidence on privacy risks using UK-based sensitive free text

WP3: Test mitigation strategies in exemplar cases, and develop tools for SACRO-ML and approaches for disclosure checking

Thank You

Visit the project webpage





Brain Frailty Integrated through Data Federation



Characterising Brain Health is Challenging

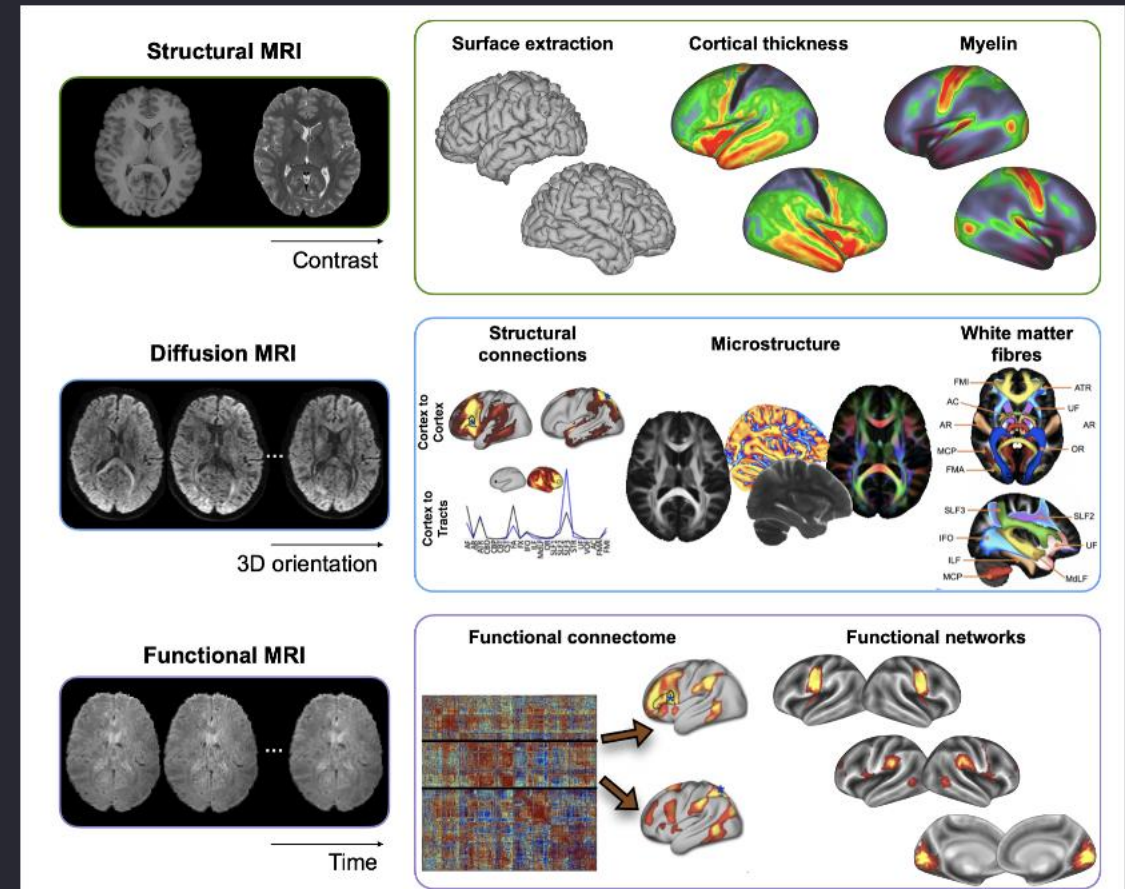
Large scale image analytics provide unique ways to do identify markers and individual phenotypes

Severely limited by:

- Inconsistent data recording
- Accessing data
- Poor data sharing
- Disconnected analytics

Consequence:

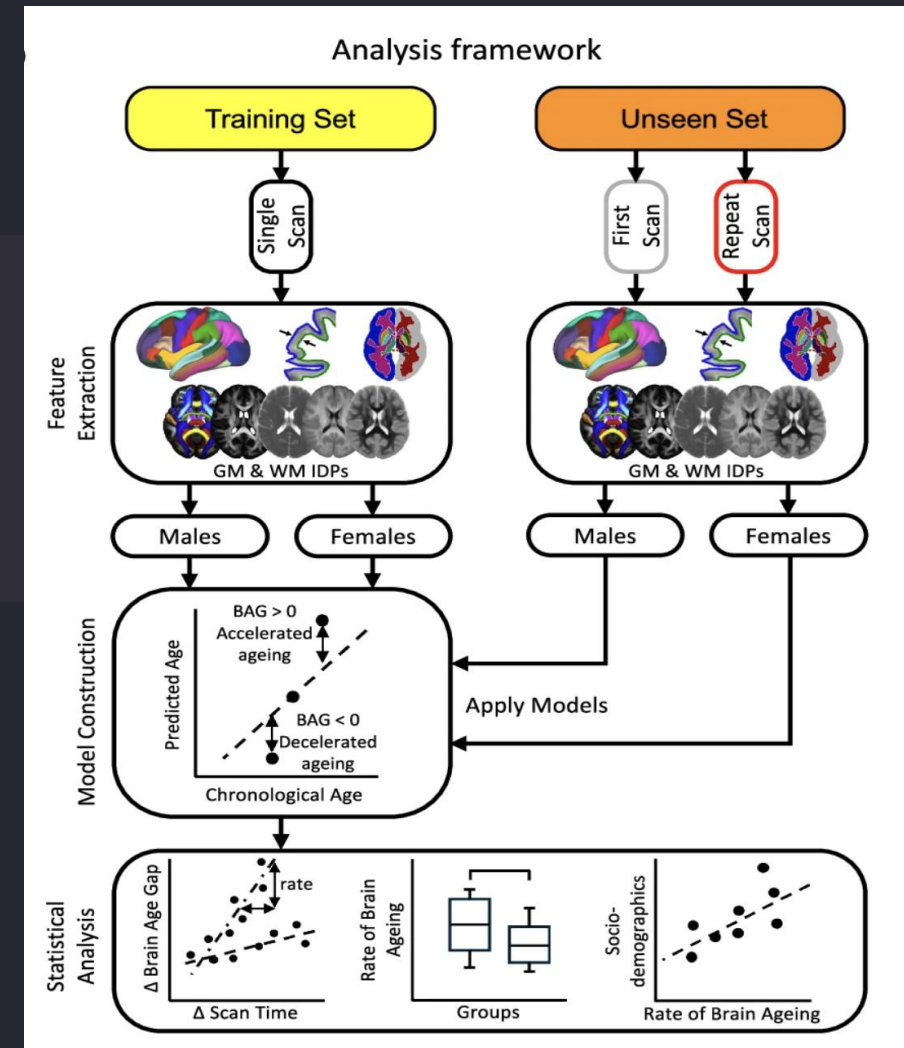
Not easy to translate research & scientific advances into clinic for patient benefit



Miller et al, Nature Neuroscience 2016
Alfaro-Almagro et al, Neuroimage 2018

Case Study: Brain Frailty

- Wear and tear of the brain, making it less able to cope with stress
- Imaging characterisation show early signs of brain frailty
- Combining brain scan with routine health data help predict outcomes and define treatment
- Our team has developed relevant analytics
- Data access and scalability for clinical translation remains a challenge
- DARE-TREvolution can help address that



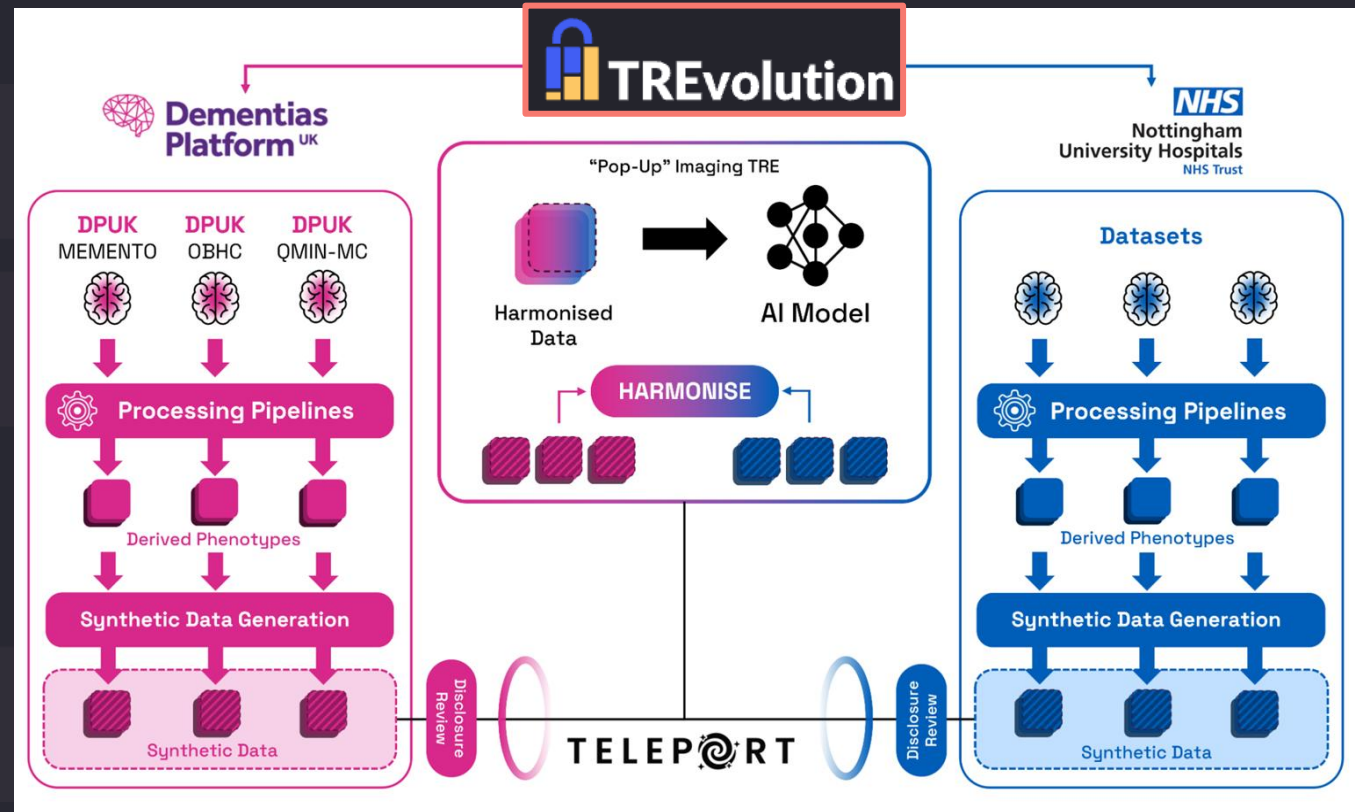
Project Goals

Support federated neuroimaging and multimodal analytics in TReVolution

Advance large scale, privacy-preserving research on brain health

Evaluate and contribute to TReVolution:

- Infrastructure readiness
- Disclosure control
- Public and patient perspectives on technology stack, reliability, security and patient data



Team



NIHR | Nottingham Biomedical Research Centre



Graziela Figueredo



Stam Sotiropoulos



Xin Chen



Dorothee Auer



Rob Dineen



Andy Rae



Emma Squires



Lewis Hotchkiss












Richard Preen



Claire Newman



 Machine learning	 Synthetic data generation	 PIE specialist
 Image analysis	 Disclosure control	 TRE provider
 Brain imaging	 Clinician & product owner	 TREvolution

PIE Activities

Aim: Co-design technical methodologies with members of the public.
Diverse PIE groups participating in each stage of the project



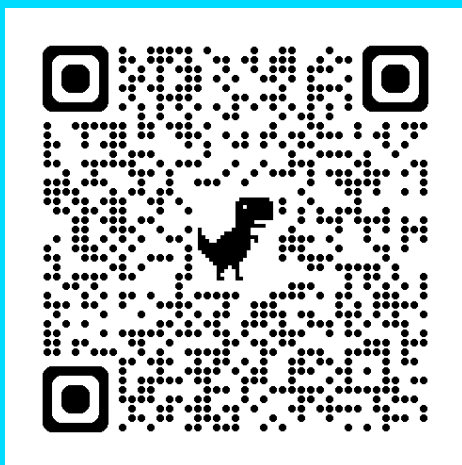
Project Contributions and Patient Benefit

- Enabling federated analysis of large-scale multimodal neuro-imaging data
- Improvement & feedback to TREvolution
- TRE-compatible FAIR analytics for brain data for a range of neurological and psychiatric disorders
- Foundations for phenotyping and precision imaging
- **Application to real-world data from NUH TRE node**



Stay in Touch

Visit our website



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Thank you and stay in touch





Cross-TRE Network for Evaluating Clinical Outcomes and National Trends in Embolic Complications of Atrial Fibrillation

14/05/26



UK Research
and Innovation



ADR UK
Data-driven change



Our Team



Dr. Fatemeh Torabi

 **Project Lead**

Assistant Professor of Healthcare Data Science & Senior Researcher




Prof. Ashley Akbari

 **SAIL co-lead**

Professor of population data science and Head of SAIL User and Data Support Services




Laura Clarks

 **Eastern SDE co-lead**

Senior Product Manager for Eastern England Secure Data Environment



Mark Avery

 **Eastern SDE co-lead**

Director of Health Informatics for Cambridge University Health Partners



Prof. Michael Gravenor

 **Analytics co-lead**

Professor of Biostatistics and Epidemiology and head of institute




Emma Squires

 **DPUK co-lead**

COO of DPUK Data portal and Head of Programmes and Innovation for SeRP



Chris Orton

 **UK SeRP co-lead**

Head of Business Development for SeRP and SAIL, Technical Lead for Population Research UK, and Technical Lead for HDR UK Inflammation and Immunity




Jillian Beggs

 **PPIE Co-lead**

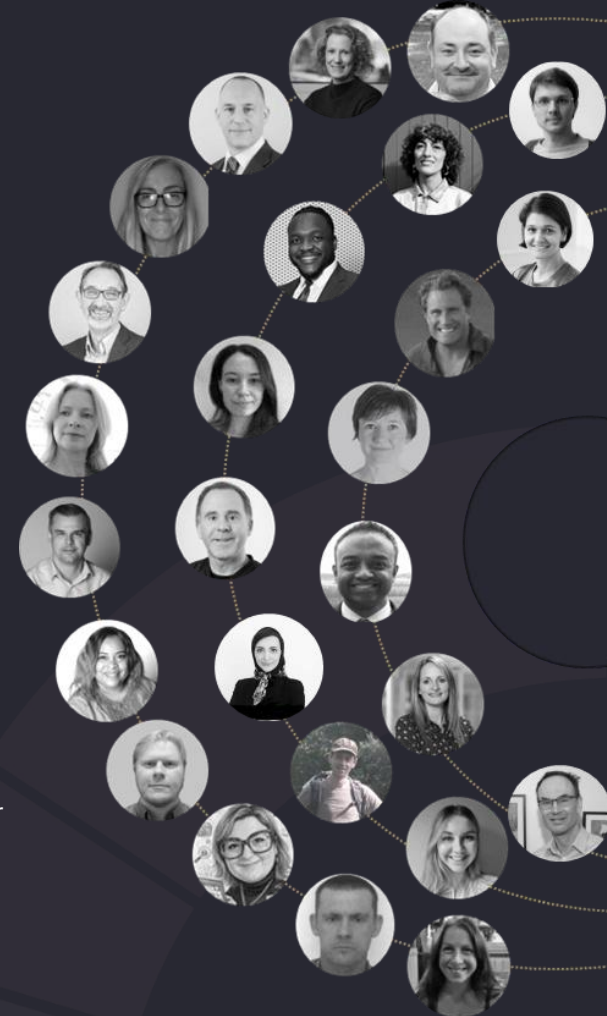
Expert PPIE research advisor



Antony Chuter

 **PPIE Co-lead**

Expert PPIE research advisor



Background and importance of this research topic

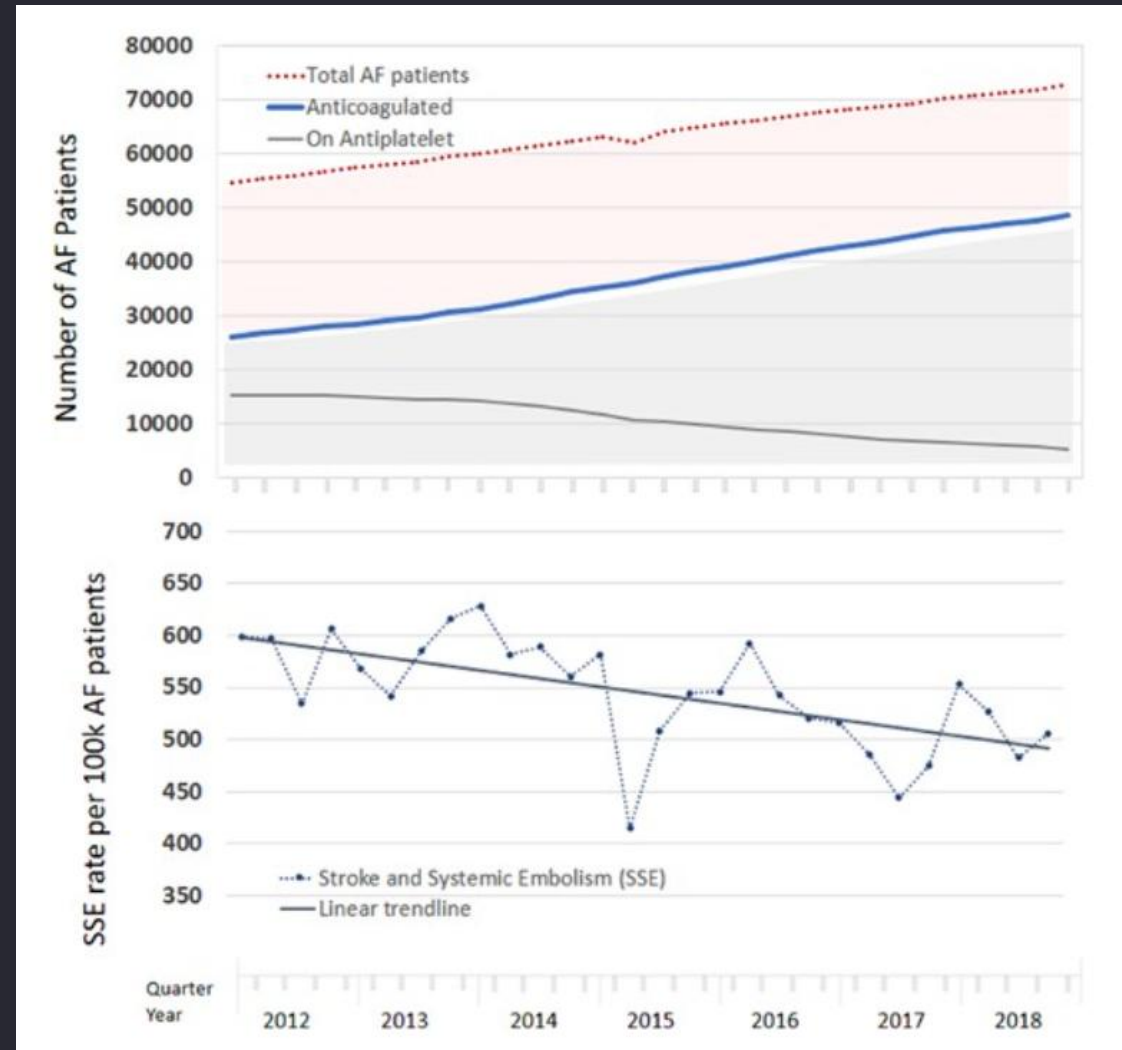
Atrial Fibrillation (AF) is the most common heart rhythm disorder.

1.5 million people in the UK

it increases the risk of stroke by 5 to 7 folds and the risk of dementia by 2 folds.

We can manage these known risks

Anticoagulant medication significantly reduces the risk of stroke in AF patients by two-third (~66%).



Torabi & Harris et al. 2022

AF patient pathways



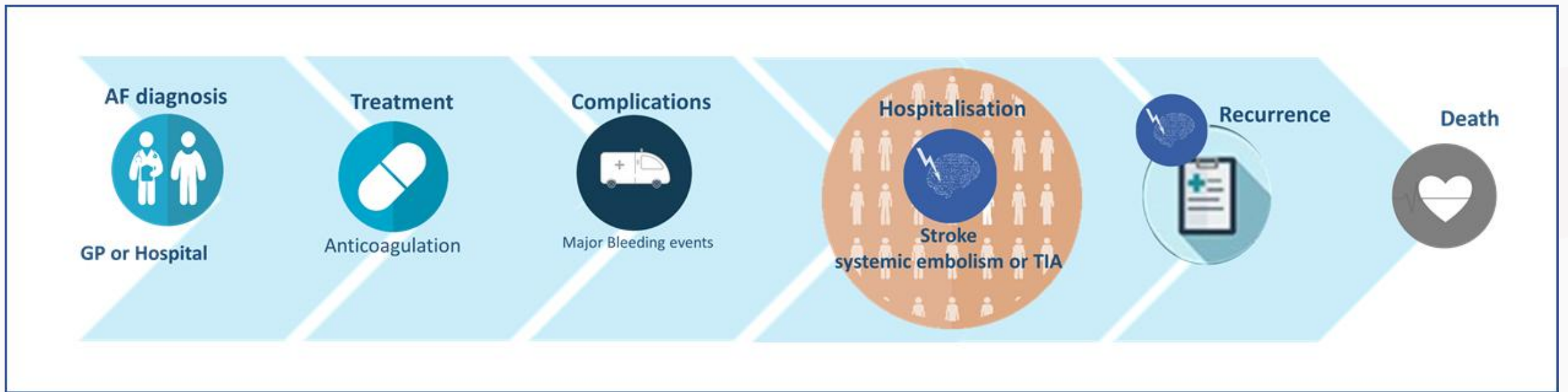
Entire pathway



Secondary care



Self-reported outcomes



WP1 – Scientific: quantify and compare hospitalisation rates for stroke, systemic embolism and Transient Ischaemic Attack among AF patients across Wales and the Eastern SDE

AF pathway in base layer



AF-Wales

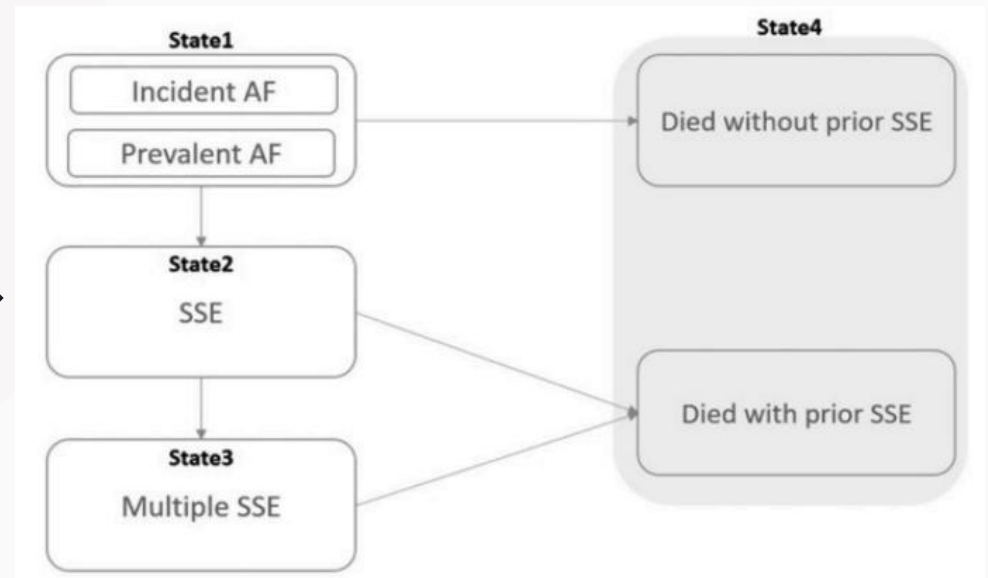
AF-ESDE

Self-reported outcomes

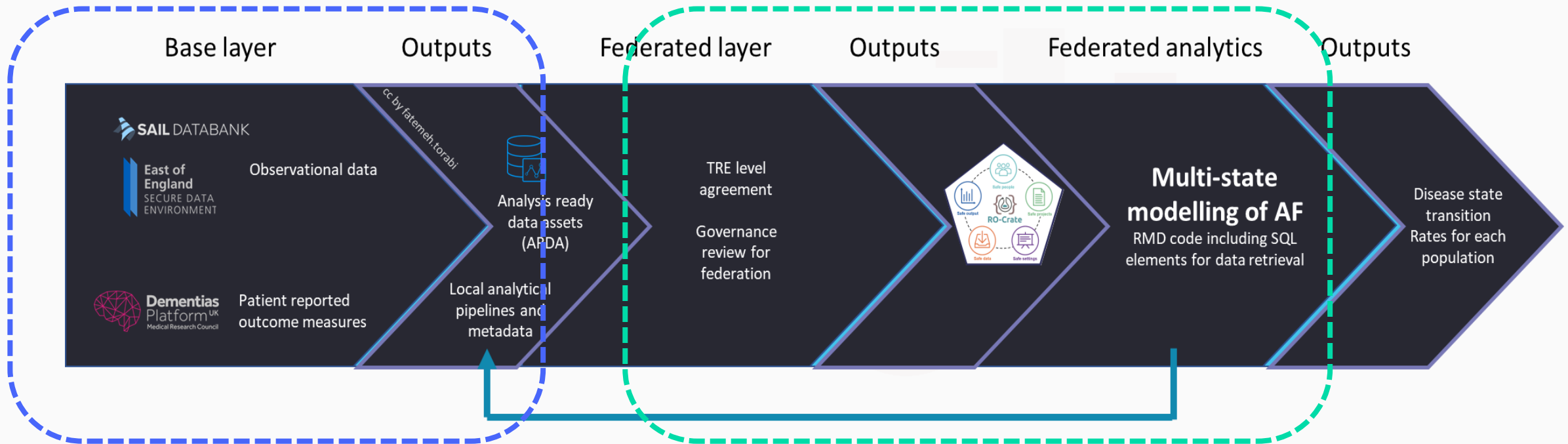
UKSeRP (TRE-FX)



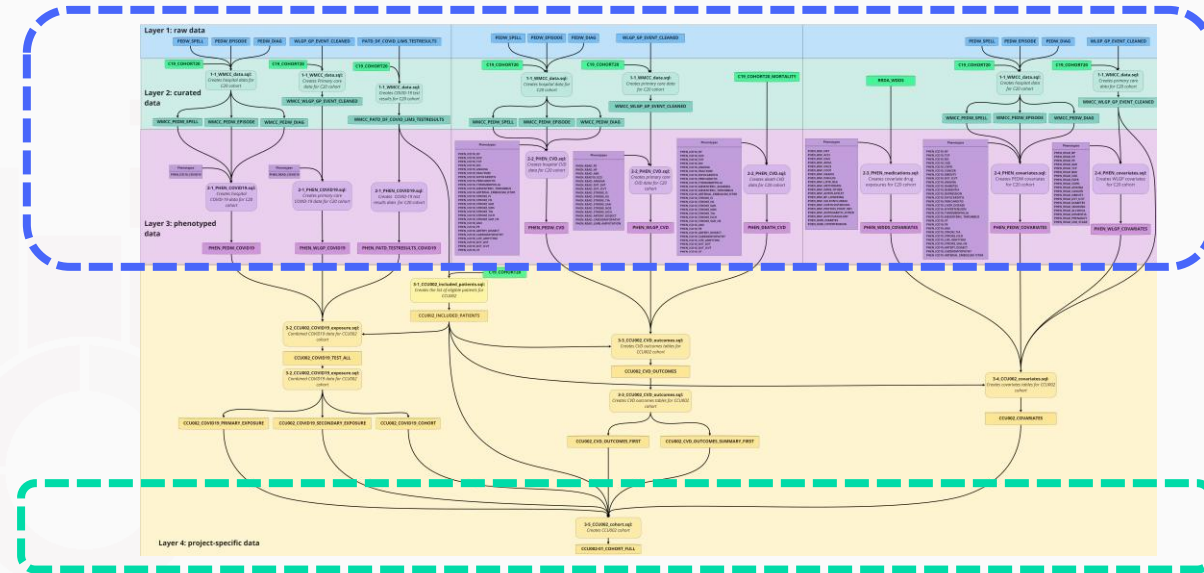
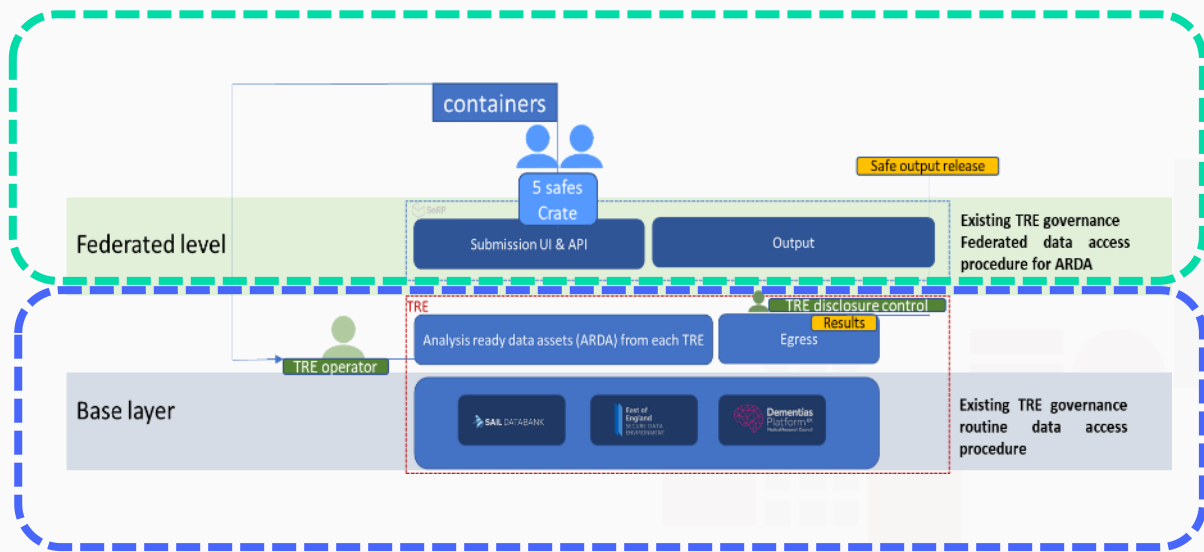
Federated analytics: Multi-state modelling of AF care



WP2 - Technical: Assess and analyse the systems effectiveness on each TRE node for adoption of DARE UK's federated analysis capabilities.



WP3 - Data Integration: Harmonise core AF cohort structures across both TRE nodes and develop containerised analytical pipelines that enable consistent, portable, and secure federated analyses across environments.



Abbasizanjani & Torabi et al. 2023



AF-Wales



AF-ESDE



Self-reported outcomes

WP4 - Engagement: Embed meaningful public and patient involvement and engagement (PIE) throughout the project to ensure transparency, accountability, and public trust.

Co-design from the start

Our PPIE representatives are those who experienced AF themselves and found value in federated way of accessing data for AF-research



Enablers



- SAFER: SAIL Advancing Federated Exploration and Readiness



- VISTA: Viable Implementation of SATRE, SACRO and Tools for AI



- 5S-TES: Delivering a federated network of trusted research environments to enable safe data analytics



- TELEPORT: Connecting researchers to big data at speed

Torabi & Orton et al. , 2023

International Journal of Population Data Science

Journal Website: www.ijpds.org




Common governance model: a way to avoid data segregation between existing trusted research environment

Fatemeh Torabi^{1,2,7}, Chris Orton^{1,3,1}, Emma Squires^{1,2,1}, Sharon Heys^{1,4}, Richard Hier¹, Ronan A. Lyons^{1,2,4}, and Simon Thompson^{1,2,3,4}

Torabi & Squires et al., 2024

nature medicine <https://doi.org/10.1038/s41591-023-02686-w>

A common framework for health data governance standards

Fatemeh Torabi, Emma Squires, Chris Orton, Sharon Heys, David Ford, Ronan A. Lyons & Simon Thompson Check for updates

A six-tiered governance framework for federated health data, with varying levels of data access and sharing, is proposed to facilitate the use of health data held within more than one secure environment, while preserving privacy.



Thank you!





AIRR-BRIDGE

A DARE UK PROJECT

Enabling large-scale health datasets
on UK national AI supercomputers

Martin O'Reilly, The Alan Turing Institute

Enabling large-scale health datasets on UK national AI supercomputers

3 TRE builders



2 supercomputers



AIRR: AI Research Resource

3 health datasets



Enabling large-scale health datasets on UK national AI supercomputers

The key idea

- To enable the **safe use** of large-scale **sensitive datasets** on the UK's national **AI supercomputers**
- Demonstrate this can be done for three **real-world health datasets** using **real-world research projects**
- While **validating solutions** developed in DARE UK programme against **real-world governance** requirements

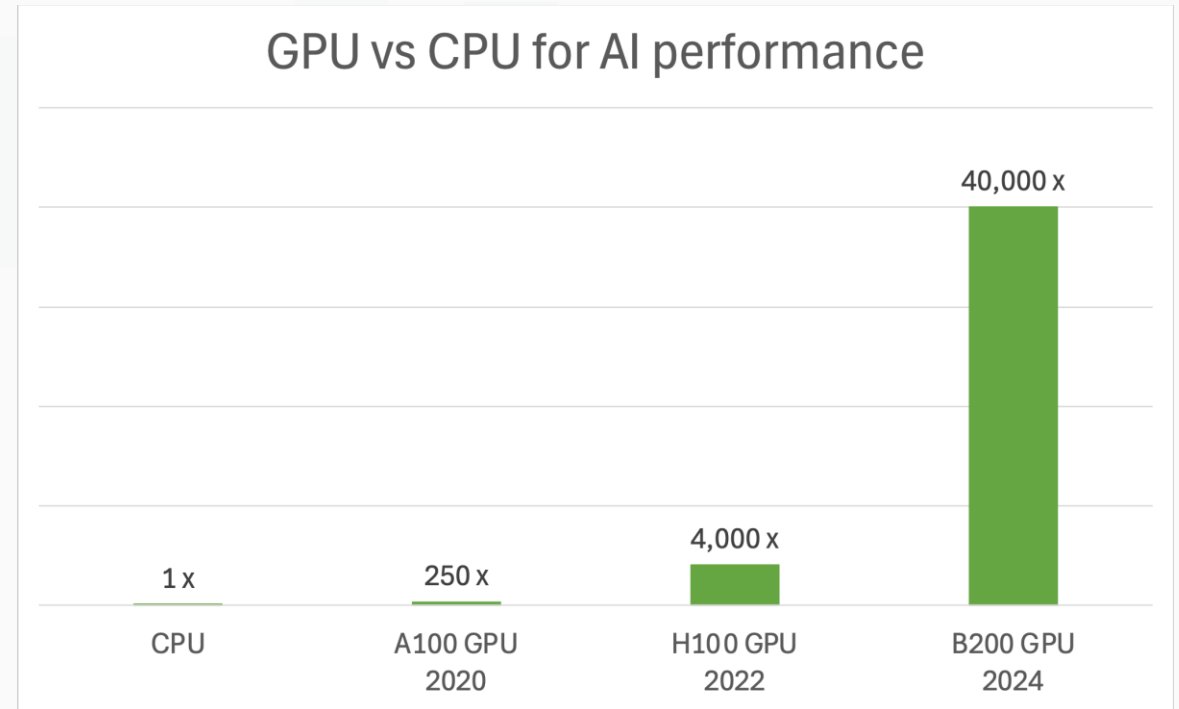
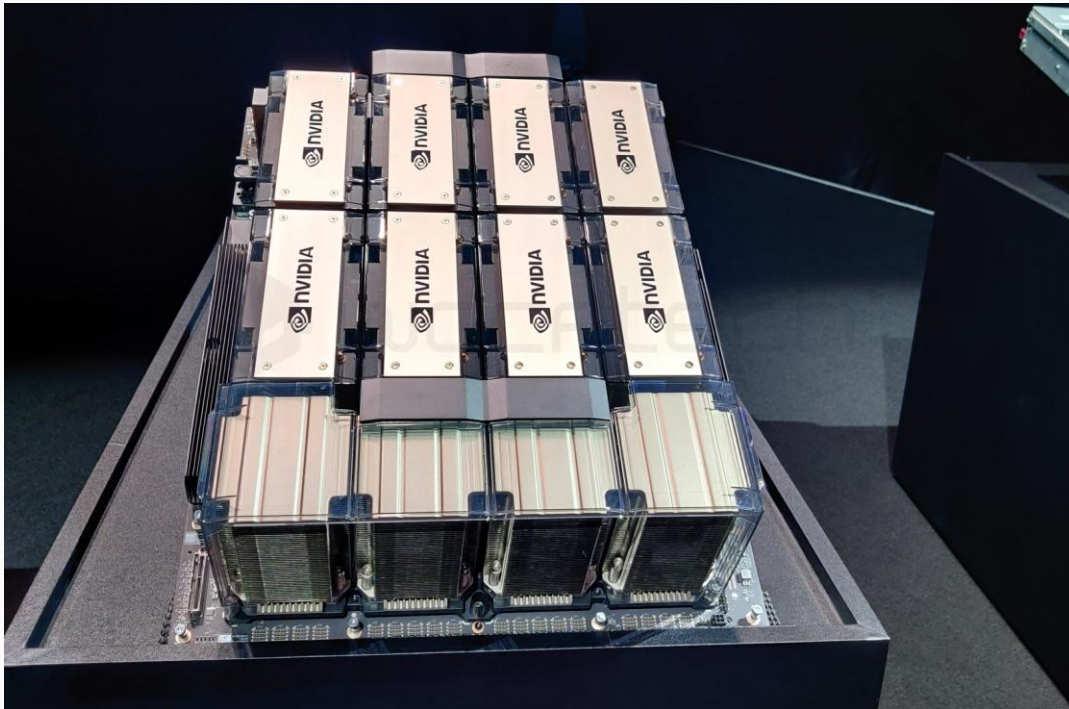
Enabling large-scale health datasets on UK national AI supercomputers

The datasets

- **UK Biobank**: Large 500,000 subject longitudinal research cohort. Long follow-up includes many different health outcomes.
- **PharosAI**: Large-scale multimodal data. 12,000 patients per year. Digital pathology, imaging, DNA/RNA, longitudinal clinical data.
- **AI Centre for Value Based Healthcare**: 170,000 brain MRI images from 40,000 patients with metadata and clinical radiology reports.

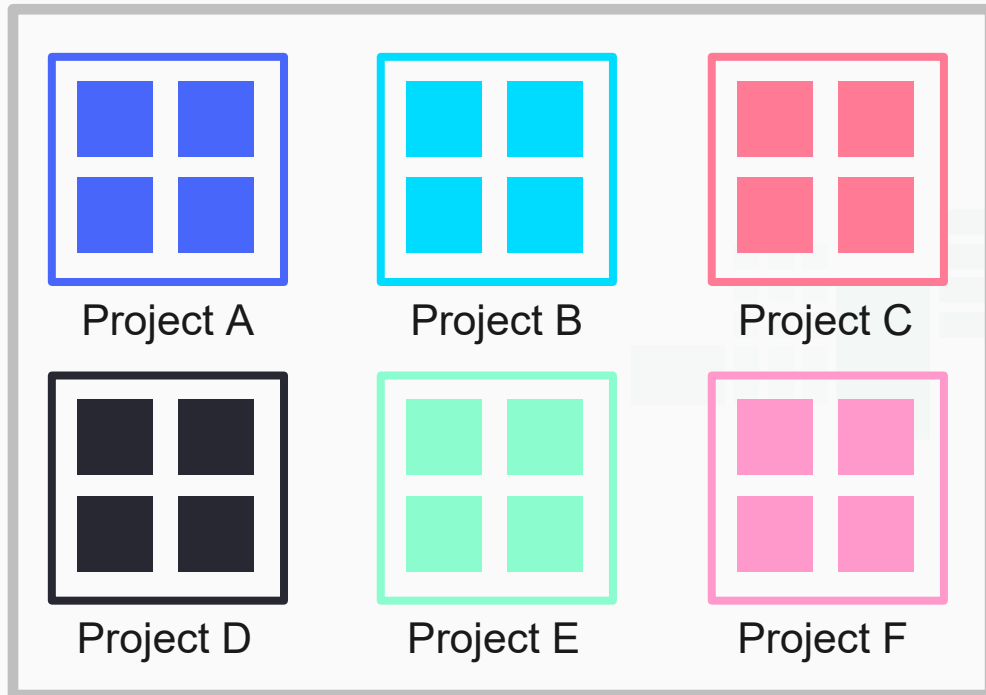
Enabling large-scale health datasets on UK national AI supercomputers

AI research needs specialised “GPU” computer processors ...



Enabling large-scale health datasets on UK national AI supercomputers

Sensitive research needs Trusted Research Environments (TREs) ...

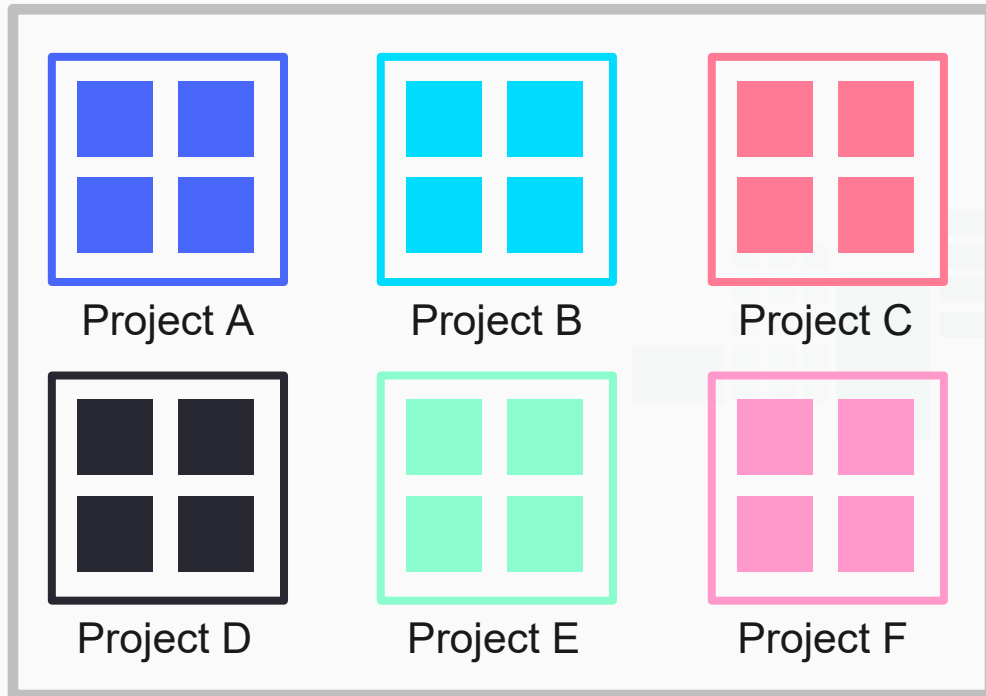


TREs combine strong **technical** controls and strong **information governance** controls to support **safe research** using sensitive data.

- Isolated projects
- De-identified data
- No internet or copy + paste
- Secure connections
- Approved projects
- Approved researchers
- Review of outputs

Enabling large-scale health datasets on UK national AI supercomputers

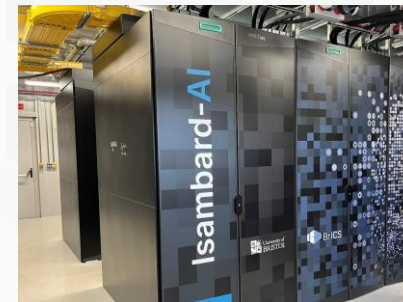
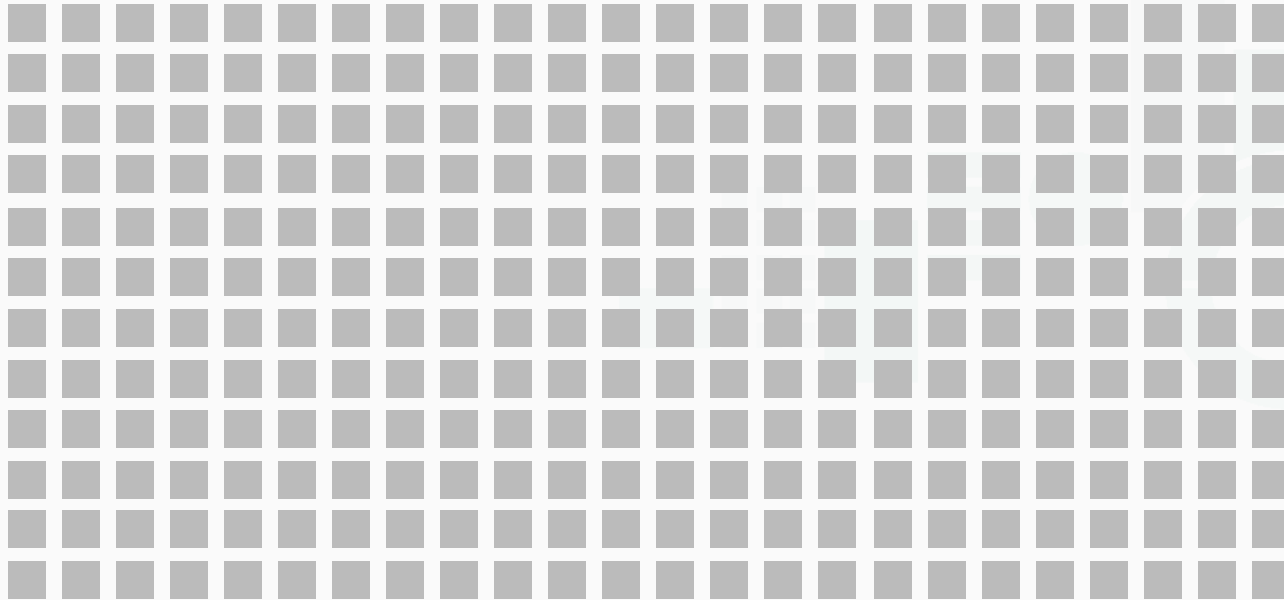
... but TREs don't have many GPUs



TREs provide a **small number of strongly isolated computers** per project

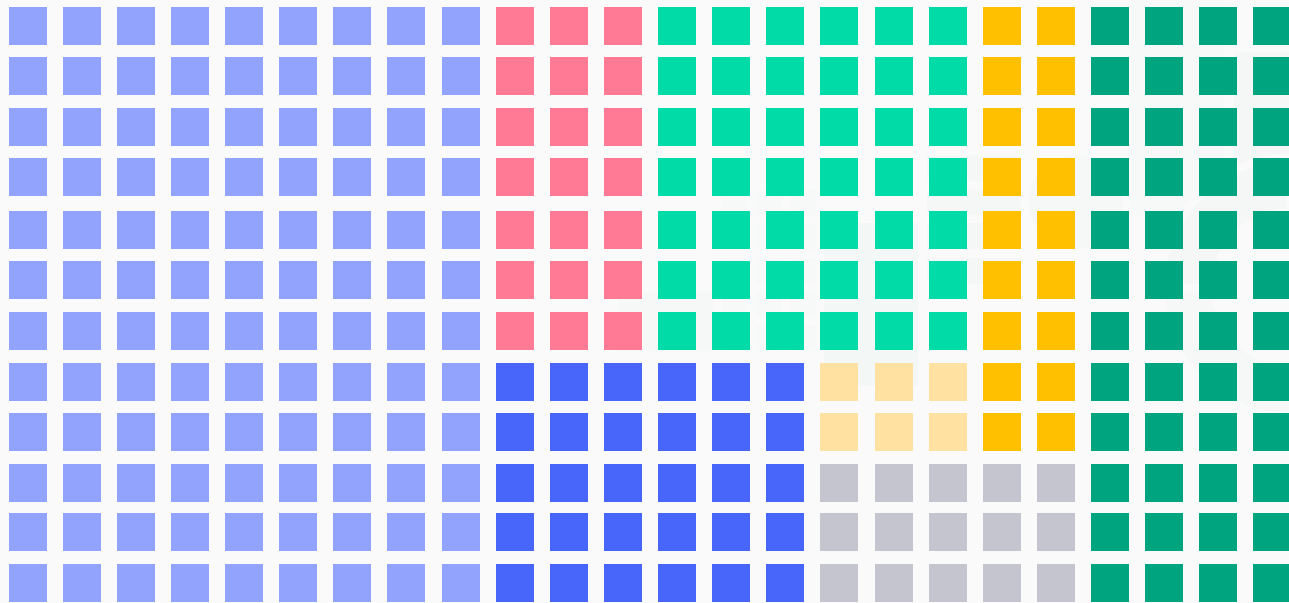
Enabling large-scale health datasets on UK national AI supercomputers

The AIRR supercomputers have lots of GPUs ...



Enabling large-scale health datasets on UK national AI supercomputers

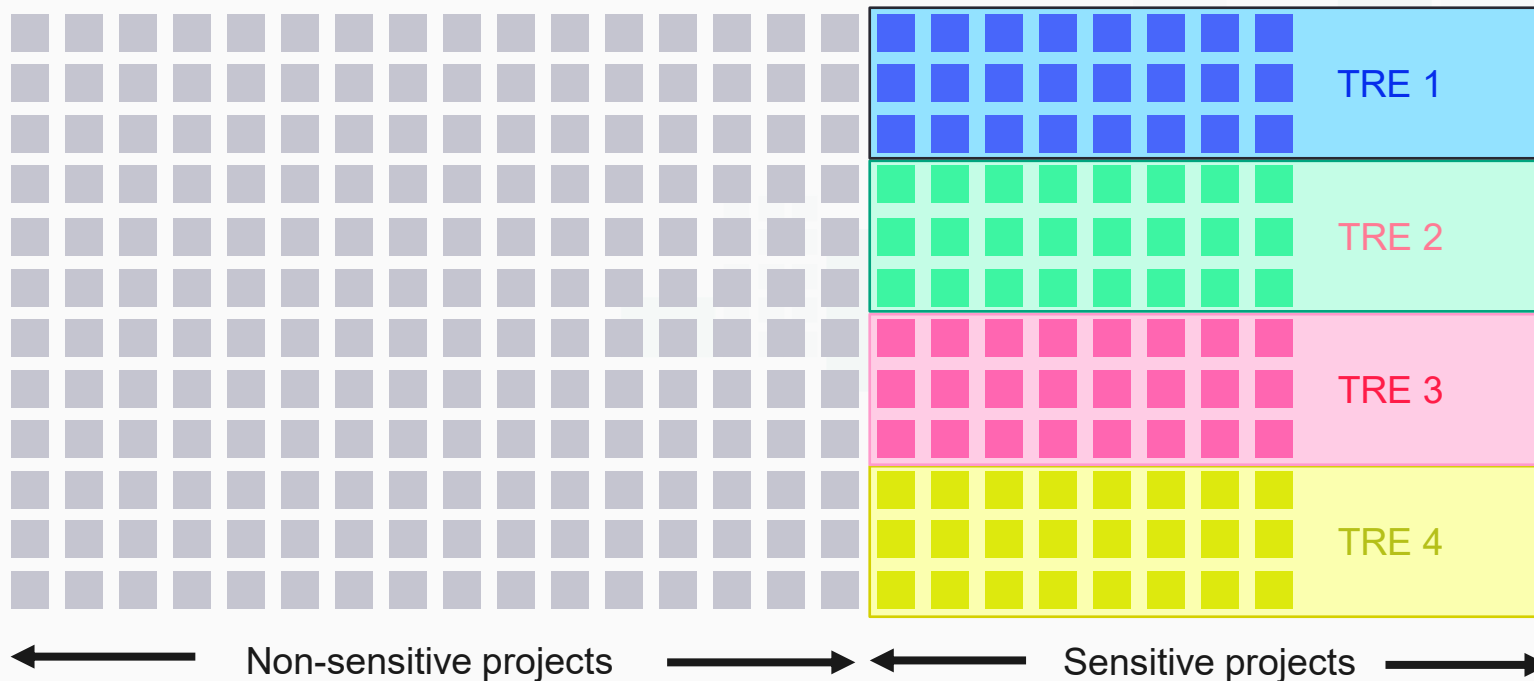
... but these supercomputers aren't TREs



Different projects and users are **not strongly isolated from each other** in the same way as they are in a TRE

Enabling large-scale health datasets on UK national AI supercomputers

... but these supercomputers aren't TREs ... yet



FRIDGE is a **satellite TRE** model that **isolates different parts** of a supercomputer for **different TREs**

AIRR-BRIDGE will **validate** this approach in a **production environment**

Enabling large-scale health datasets on UK national AI supercomputers

Demonstrate this with real projects using real health data ...

Federate the Turing
and KCL TREs...



...with the Isambard-AI and
Dawn supercomputers



Enabling large-scale health datasets on UK national AI supercomputers

... while leveraging and validating other TRevolution components

Build a **SATRE** compliant TRE-on-supercomputer solution



The Alan Turing Institute



Test the **K8TRE** “full stack” TRE on the AIRR supercomputers



Test the **SACRO-ML** disclosure control tools for AI models

Enabling large-scale health datasets on UK national AI supercomputers

PIE and community engagement

- Dedicated public member co-lead for PIE
- Turing TRE PIE group embedded in project
- Engagement with data provider and DARE UK PIE groups
- Wider public engagement via collaboration cafes
- Engaging with TRE and supercomputer communities



Federated Research on Antihypertensives Maternity Emulation Trial



What is the problem ?

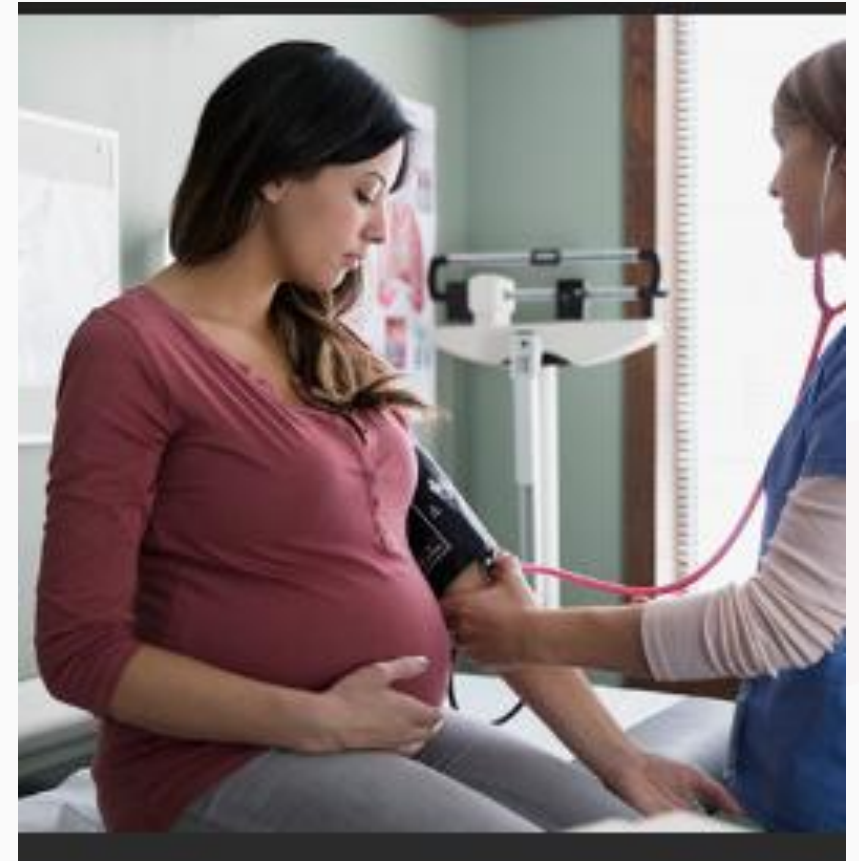
10%-15% of women have high blood pressure in pregnancy and this is harmful for the mother and baby

Trials are rarely big enough to be able to examine if recommendations for medications should differ by ethnicity and they rarely look at long term outcomes for mother and baby.

Medical records of women who have had high blood pressure could help answer these question.

We want to bring together the medical records from women across the UK so we can compare people with the same ethnicity who took different drugs in pregnancy.

We also want to see the long term health of the baby (in the first 2 years of life)



However, data can not just be shared across countries

We are using the technology developed by DARE to create a Pop-up Trusted Research Environment where data can be held and shared but still meet all the governance requirements of the data holders.

A space where all the data providers can come together.

This platform is called TELEPORT



Overview of the project

- **Four birth studies in UK have come together to make sure their data is the same.**
- There are 19 million pregnancies in these birth studies.
- This means we will have the power to compare women of the same ethnicity taking different drugs to see what works for African women, or Asian women or White women and we believe it will be different.
- We have the power to follow infants up for 24 months after their birth to examine:

If health visitors had any concern with growth and development

If infants were diagnosed with low heart rate, failure to grow or any heart or breathing disorders



Public involvement

- This work was developed because our Born In Wales Public Involvement group said research should improve safety of medicines in pregnancy.
- Women have taken drugs not knowing the long-term effects, we should learn from their experiences.
- The work is developed with the SAIL Consumer Panel
- The findings will be taken to a workshop to work with the public to interpret findings on medications, develop recommendations, develop community facing summaries. Working with existing PIE members in Born in Wales, Scotland and South London.
- It is vital that the findings are interpreted with the public to ensure messages are clear and can help women and their families.



What difference could it make ?

- This would make a major difference in medicines in **pregnancy, infancy** and long term follow up.
- Trials face ethical barriers and are difficult in to do in pregnancy or infancy, but people still need to take medication. We need to learn from their experiences
- Bring different data sets together is especially important when comparing smaller groups (e.g. different ethnic groups or people taking multiple medicines). In a single data source, the numbers in each group are often too small to spot real differences.
- It would also be used for rare conditions where you need to bring all data together to match individuals



The team includes:

- Obstetric physician
- Pediatric pharmacist
- Fetal and maternal medicine clinicians
- Data Scientists
- Epidemiologists
- Co-production Public Involvement lead



mireda
mother & infant research electronic data analysis



DataLoch

Health
Informatics
Centre



HEAL-Scot

A DARE UK PROJECT

Housing, Environment and Location data linkage in federated TREs mapping geospatial inequalities

Dr Laura Ward

lward001@dundee.ac.uk

Public webinar 14/05/26



UK Research
and Innovation

HDRUK
Health Data Research UK



ADR UK
Data-driven change

Our Team Linking Safe Havens



SOUTH-EAST

EAST



Kathy Harrison
Project Lead



Amy Chong
Data Analyst



Richard Walls,
Project Co-lead



Chris Hall
TRE



Magalie Guignard-
Duff, Data



Derek O'Sullivan
Developer



Stuart Dunbar
PIE/ Comms



Jenny Johnston
Governance

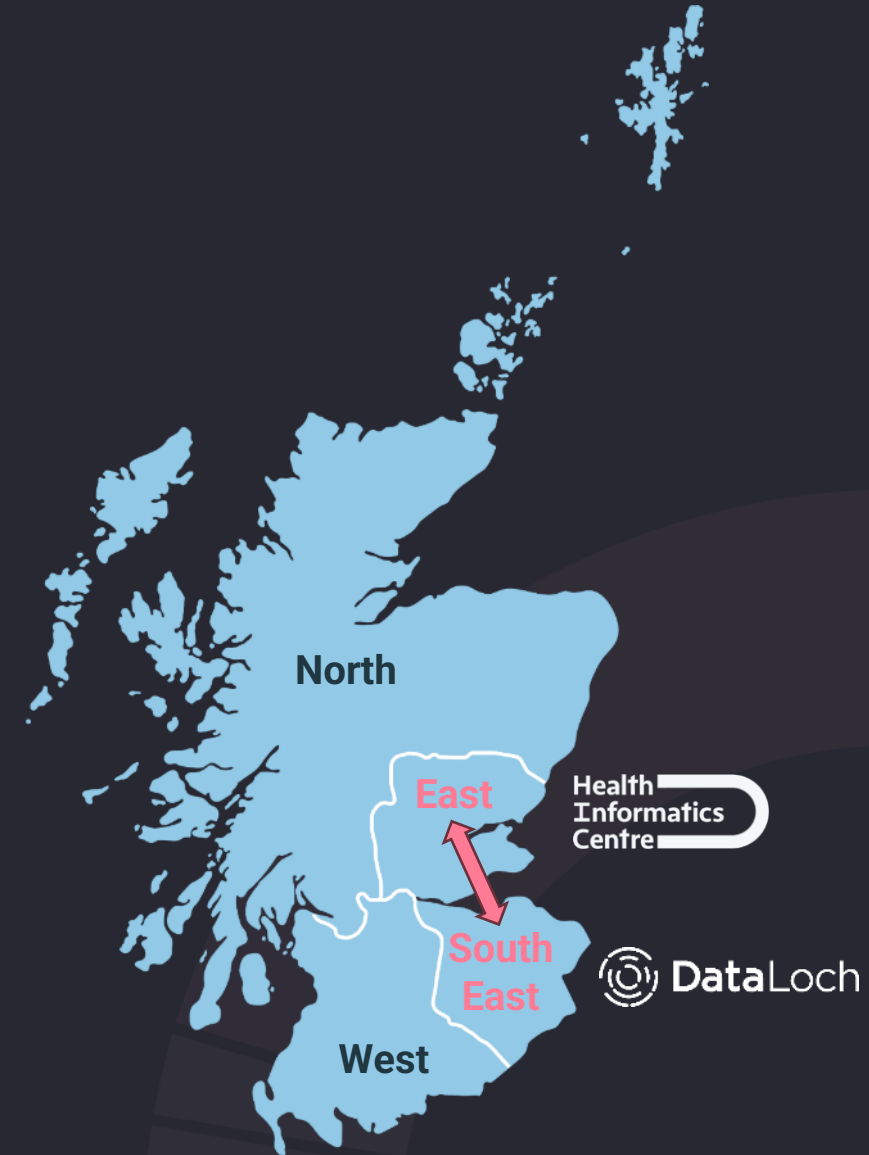


Aaron Jackson
Infrastructure



Laura Ward,
RWE Research

Kathy Kennedy & John McCullough are our 2 Public contributors



Social determinants of health

- **Prof Michael Marmot** founded the Institute of Health Equity
- His book, the “Health Gap” opens with this story of him training as a medic in the 1960’s

“‘Oh doctor’ said the patient, ‘my husband is drinking again and beating me, my son is back in prison, my teenage daughter is pregnant, and I cry most days, have no energy, difficulty sleeping. I feel life is not worth living.’ It was hardly surprising that she was depressed. ...The psychiatrist told the woman to stop taking the blue pills and try these red pills. He wrote out an appointment for a month’s time and, still a picture of misery, she was gone. That’s it? No more? To incredulous medical students he explained that there was very little else he could do. “
- **Social determinants of health are the ‘causes of the causes’**

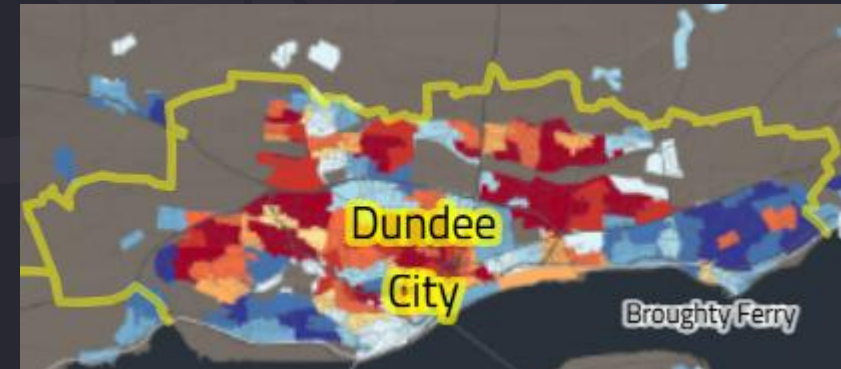
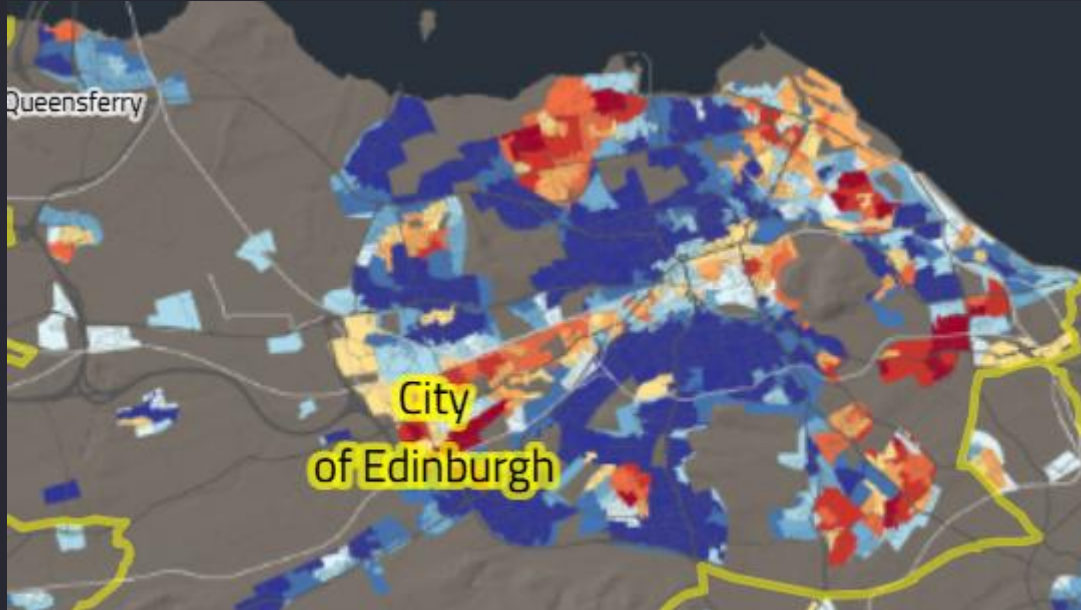


Professor Michael Marmot



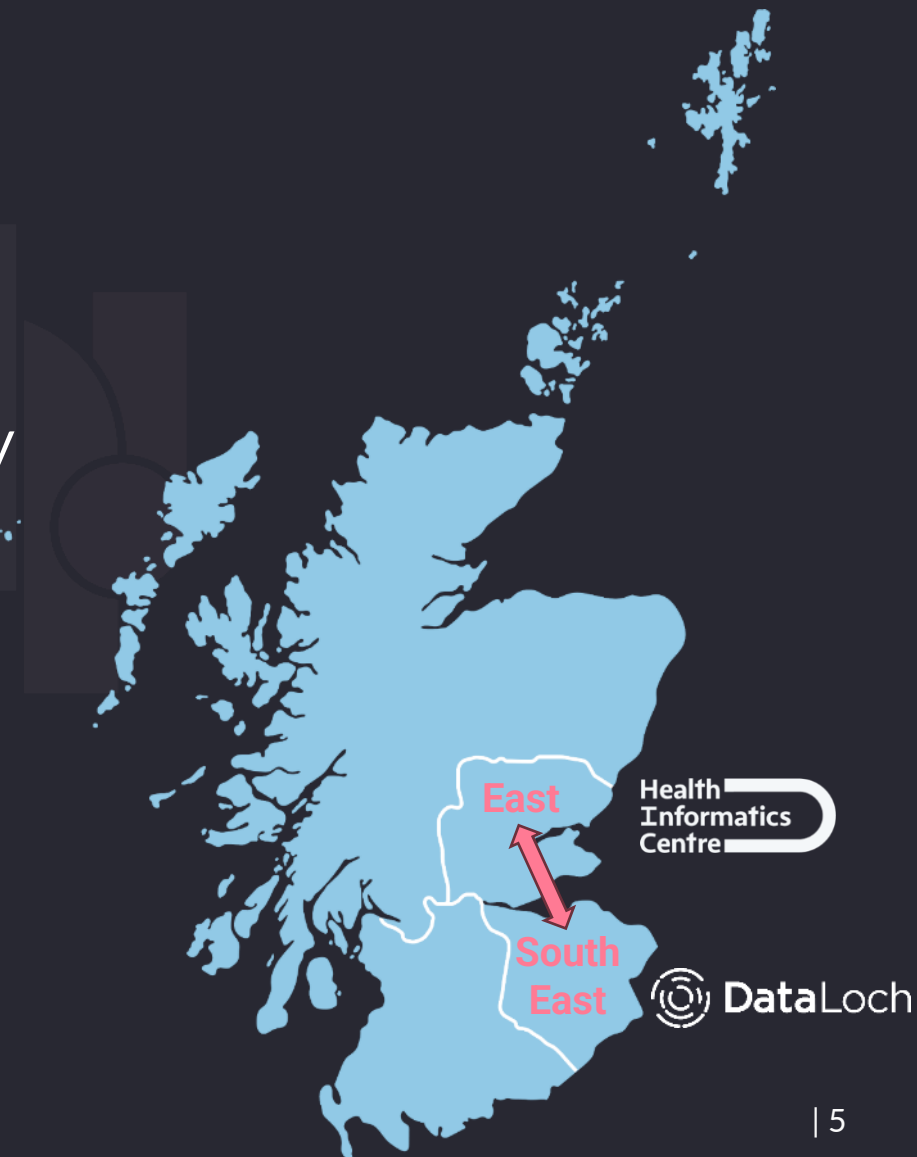
The health gap

If I am born, live, work, and age in a relatively deprived area, how much does this influence my health?



Geospatial and health inequalities

- Our social, economic, and environmental conditions are the main drivers of our health and health inequalities
 - For early deaths, they are responsible for **54%** variation
- **HEAL-Scot** links different sources of information across the **health and non-health**
 - Housing, neighbourhood access to health care, proximity to green spaces, air quality
- We will compare regional differences to begin to **try understand where differences are in terms of geospatial inequalities**
- This will be done in each safe haven (federated approach), hosting data for NHS Lothian, NHS Tayside, and NHS Fife,
- Research conducted inside a TRE



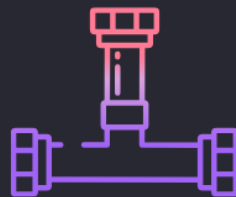
Technical pipelines
for cross-sector data



PPIE governance
safeguards



TRE federation and
scalability




Social determinants of
avoidable hospitalisations
and early death



 **LOCATION**

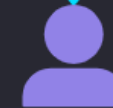
- Economic wellbeing
- Air pollution
- Access to services
- Proximity to green/blue space

 **HOUSEHOLD**

- UPRN coordinates
- Type
- Housing instability

 **INDIVIDUAL**

- Hospital
- Death
- Demography

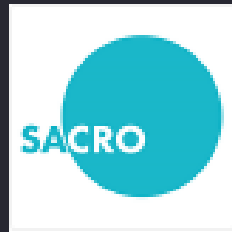
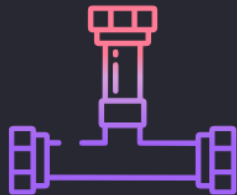


INCREASING
SIZE OF
AREA

Technical pipelines
for cross-sector data



TRE federation and
scalability



PPIE governance
safeguards



SATRE K8TRE



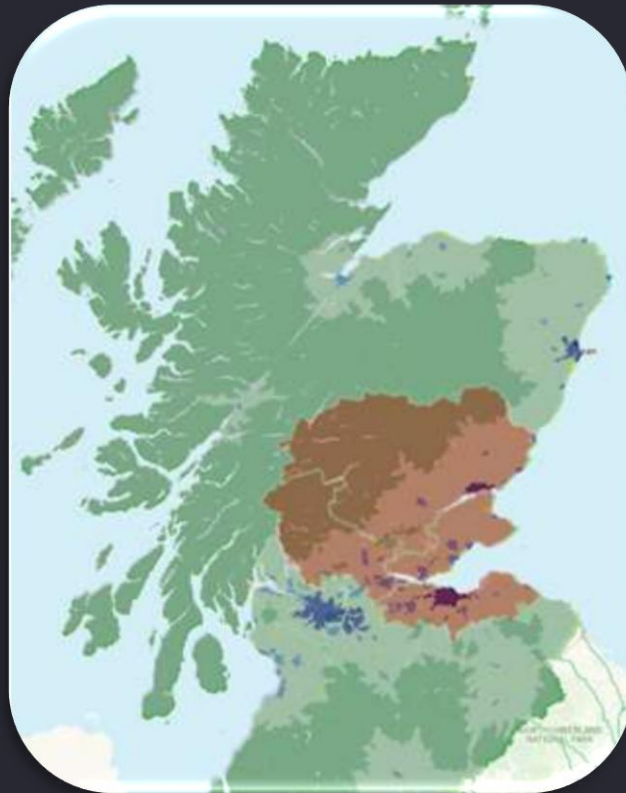
Social determinants of
avoidable hospitalisations
and early death

So what? How is public involvement shaping HEAL-Scot?

- **Team members:** 2 contributors will join the monthly team meetings and help shape decisions
- **Shaping safeguards:** public workshops will help define what is acceptable and what is not, when linking health and non-health geospatial data
- **Feedback loops:** governance designs, communications and results will be tested with public groups and feedback listened to
- How are we doing this? Working with **existing groups** -
 - **DataLoch** public panel
 - **RDS** public panel – Scotland Talks Data
 - **TREvolution** Collaboration Cafes



Stay in touch with our team



Visit our website



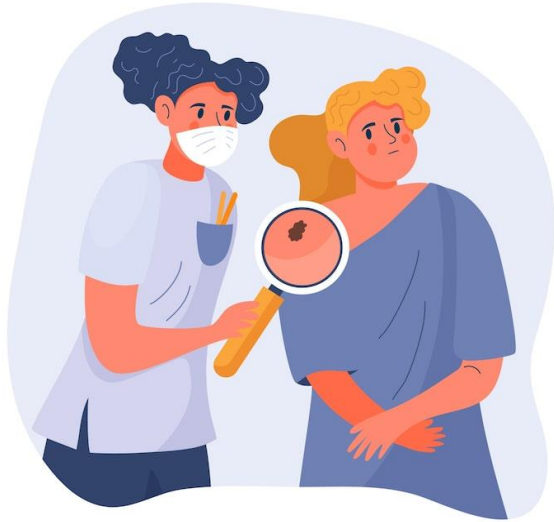
Thank You & Get in Touch



Kathy.Harrison@ed.ac.uk

MELODY :

federated **M**achin**E** Learning **f**or **D**ermatolog**Y** (**RWE** - **R**eal **W**orld **E**vidence Project)



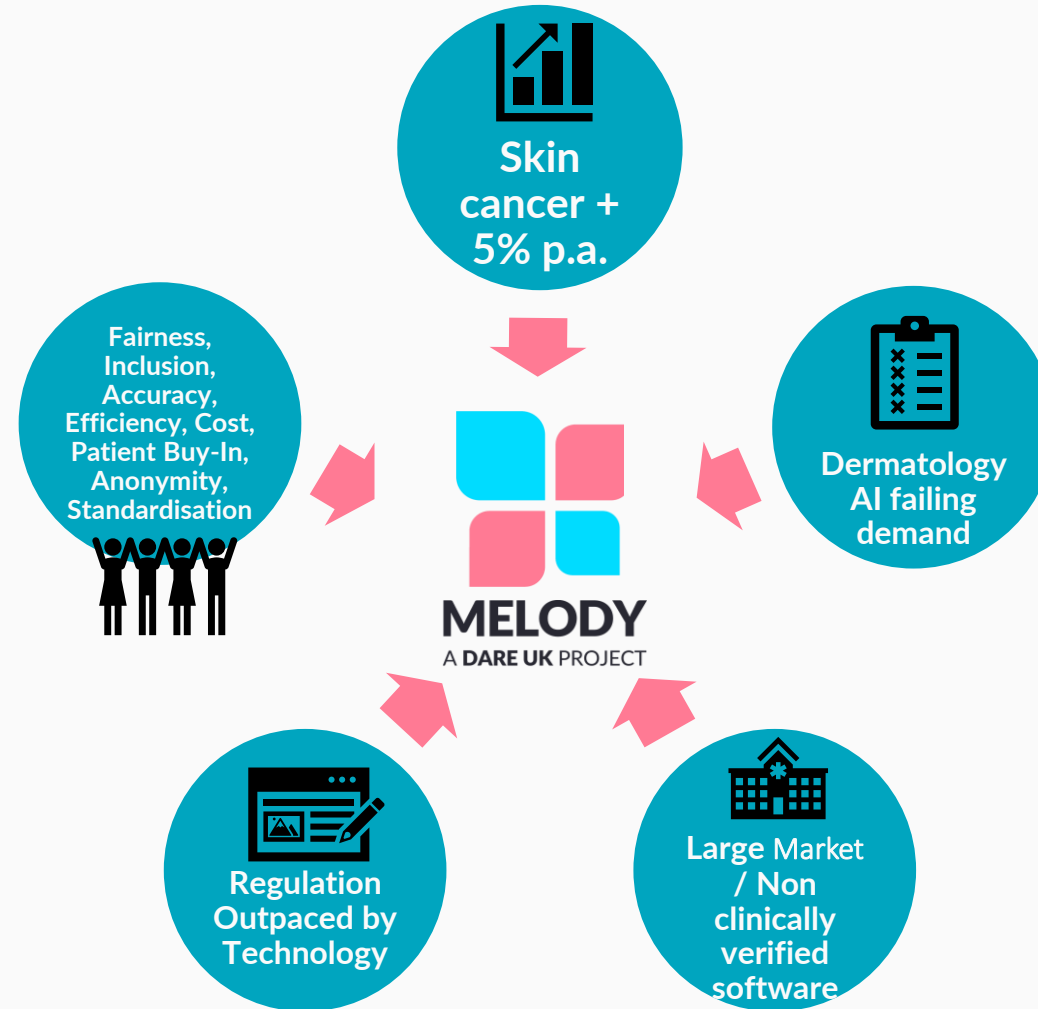
'From capability to Research' Webinar

Richard Walls

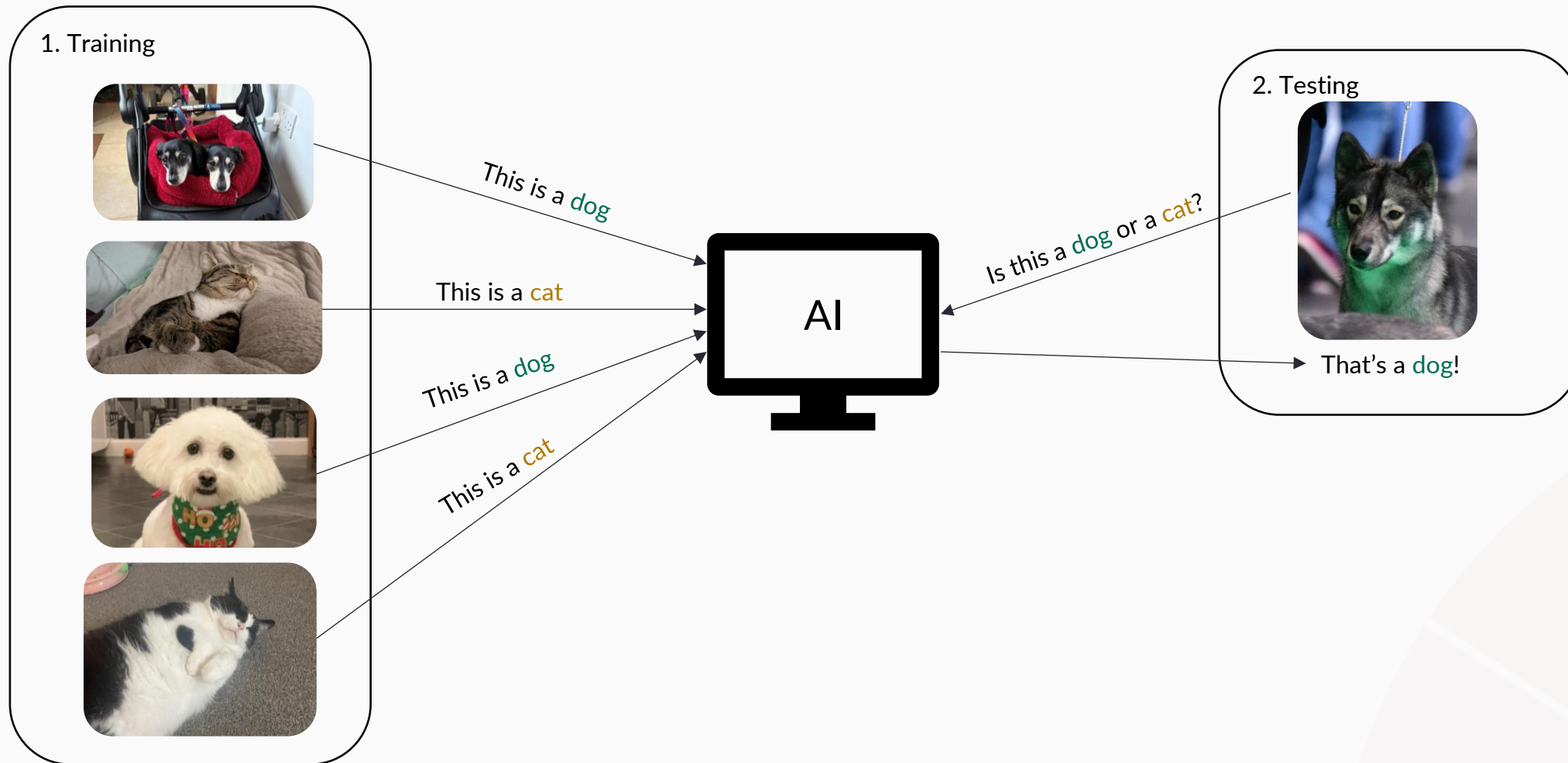
Operations Director, Health Informatics Centre (HIC), University of Dundee

14.05.2026

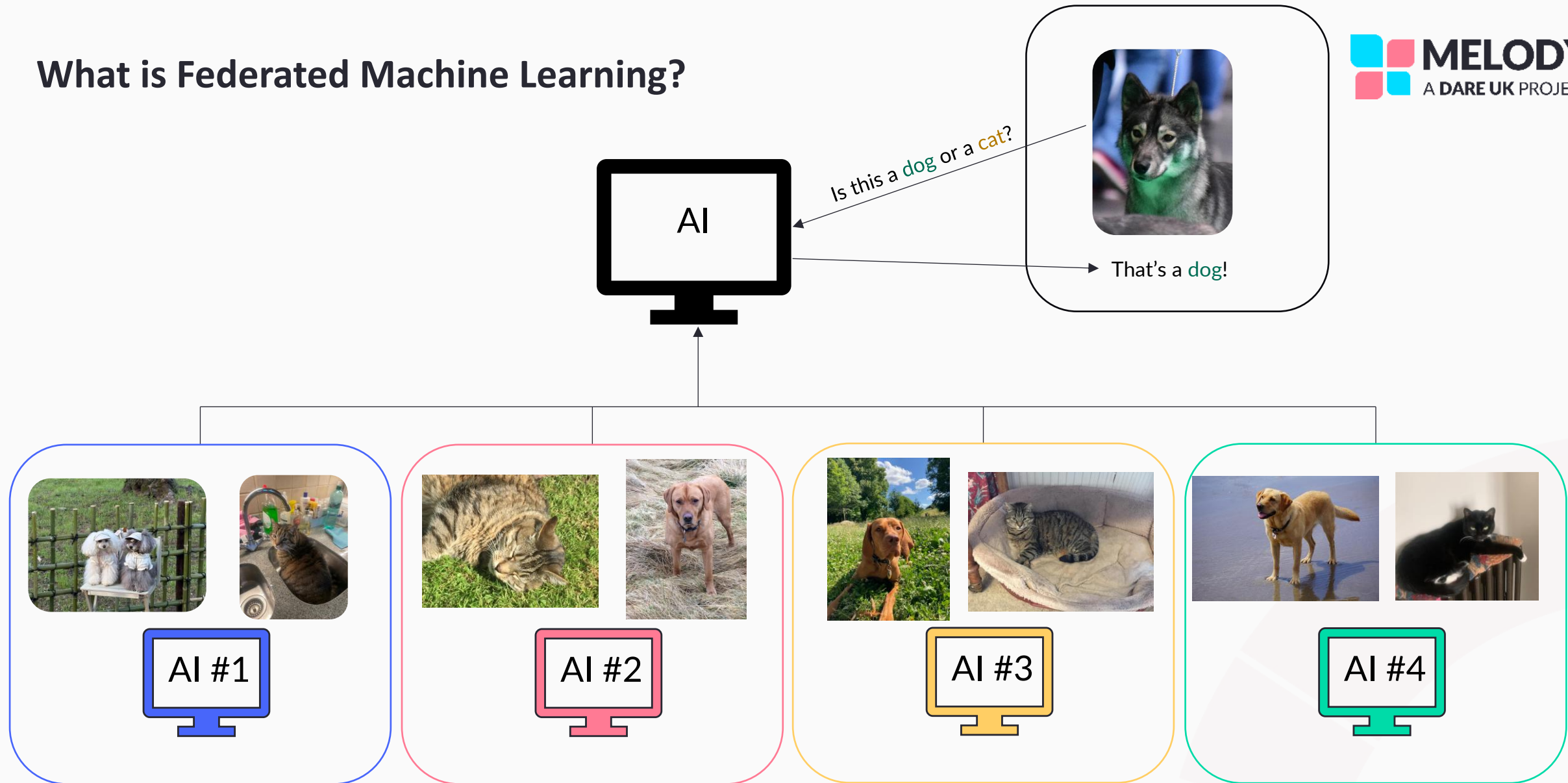
MELODY - federated Machine Learning for Dermatology



What is Machine Learning?



What is Federated Machine Learning?






Project Aims & Objectives

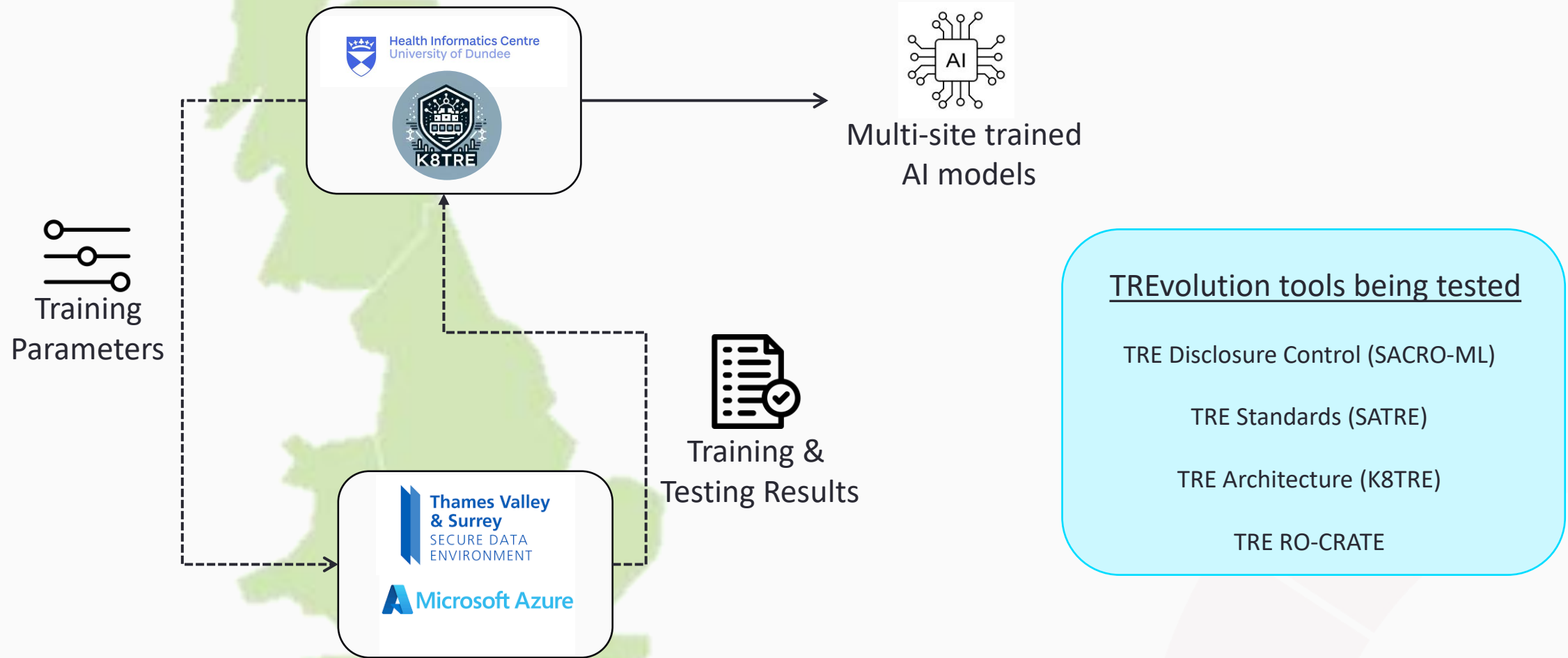
Aim: Develop a platform for federated machine learning (ML) across multiple TREs.

'Big Picture' Aim: To develop a national skin AI assurance platform (with real world, inclusive and carefully curated skin disease data) to use in testing and developing skin AI.

Project Objectives:

-  Develop **a solution to allow federated ML** across TREs.
-  Create **real-world dermatology datasets** across two locations (NHS Tayside and Oxford University Hospitals NHSFT).
-  Train and assess **a ML model on skin-lesions across the two sites**, using federated ML techniques as a proof of concept.

MELODY Approach: Federation



Public Involvement in MELODY

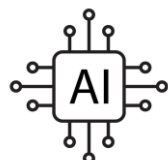
NHS SKIN
SHARE
UNIVERSITY of STIRLING
NIHR | Oxford Biomedical Research Centre
independent cancer patients' **voice**
RAD

Health Informatics Centre
University of Dundee
CO-CONNECT
Scotland Talks Data
Alleviate
The Advanced Pain Discovery Platform (APDP) Data Hub



How we would like the public to get involved in MELODY

Technology



- How should we approach federated ML in practice?
- What technical checks and balances should be put in place?
- How can we ensure data privacy for individuals?
- How should we monitor what researchers are doing with this new technology?

Data & Governance



- What dermatology images can be used?
- How can dermatology images be used?
- How can we make them non-identifiable?

Evaluation & Communication

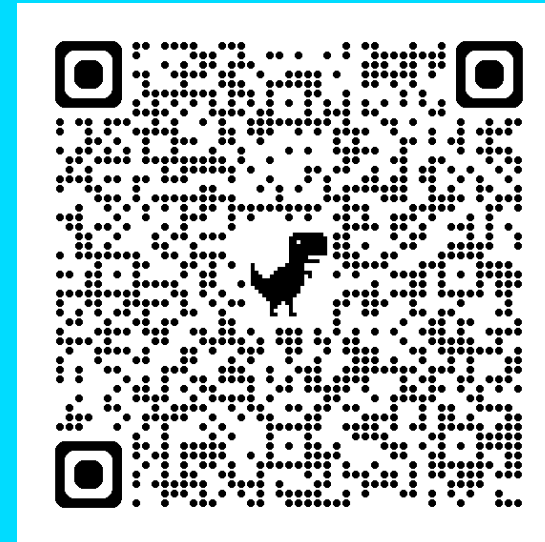


- Is this federated ML approach safe?
- Is this ML approach trusted by the public?
- Explore developing federated machine learning policies
- Co-lead engagement and dissemination events

If you are interested in getting involved in MELODY, please get in touch!



Sign up to our focus group



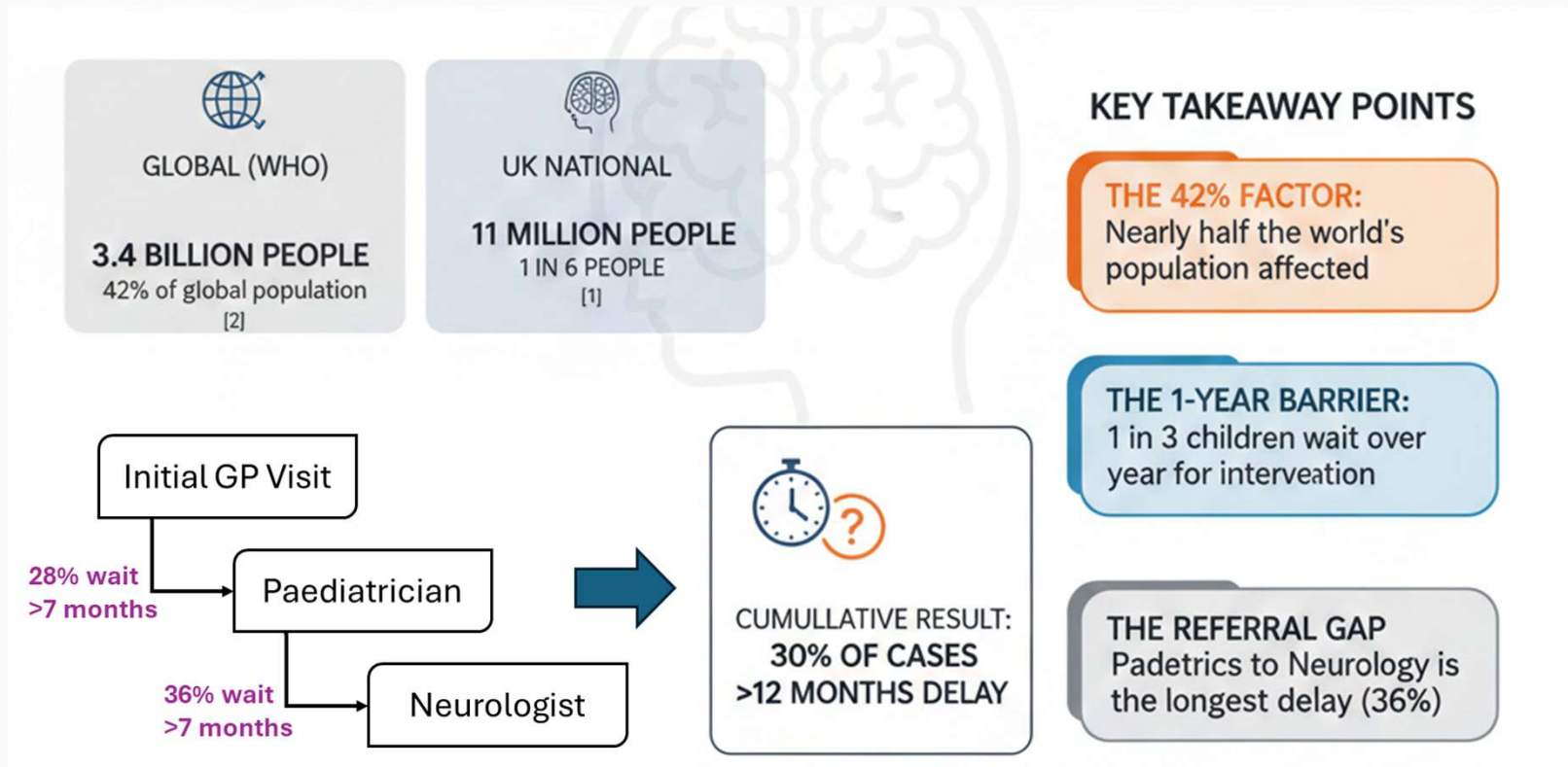


SAFEVID: Spasms Analysis using Federated Learning from Videos Across Multiple TREs

Edmond S. L. Ho, University of Glasgow



Neurological conditions - a major public health crisis



Infantile Epileptic Spasms (IES)

- IES is the most common epilepsy syndrome in early childhood
- **Lead time** - time from epileptic spasms beginning to treatment beginning
- Can smartphone videos be used to reduce Lead Time?

THE JOURNAL OF PEDIATRICS • www.jpeds.com

ORIGINAL
ARTICLES

The Effect of Smartphone Video on Lead Time to Diagnosis of Infantile Spasms

Chethan K. Rao, DO, MS^{1,2,3}, Douglas R. Nordli, III, MD^{1,2}, Joshua J. Cousin, MD^{4,5}, Danielle S. Takacs, MD^{4,5}, and Raj D. Sheth, MD^{1,2}

Table II. Lead times and treatment outcomes of video and nonvideo groups

Characteristics	Nonvideo group, n = 37	Video group, n = 43	P value
Lead time, d			
Infantile spasms start to first encounter	15 [2.5, 67.5]	6 [2, 10]	.02
Infantile spasms start to first EEG	23 [4.5, 67.5]	7 [3, 14]	.007
Infantile spasms start to diagnosis	25 [9, 84.5]	8 [4, 30]	.006
Infantile spasms start to treatment initiation	26 [9, 87.5]	9 [5, 30]	.008
Treatment outcomes			
Treatment responder	18 (49%)	32 (74%)	.02
Ongoing infantile spasms at last follow-up	13 (35%)	6 (14%)	.03
Developmental delay at last follow-up	34 (92%)	36 (84%)	.27
LGS at last follow-up	6 (16%)	1 (2%)	.03

Data are presented as n (%) or median [25%-ile, 75%-ile].

Shorter
Lead Time

Improved Treatment
Response Rate

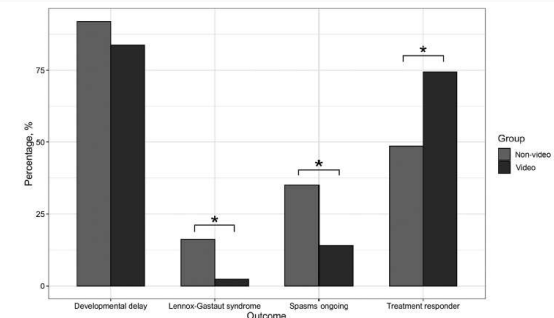


Figure 3. Comparison of treatment outcomes including developmental delay, LGS, ongoing infantile spasms, and successful response to treatment between video and nonvideo groups.

The Effect of Smartphone Video on Lead Time to Diagnosis of Infantile Spasms

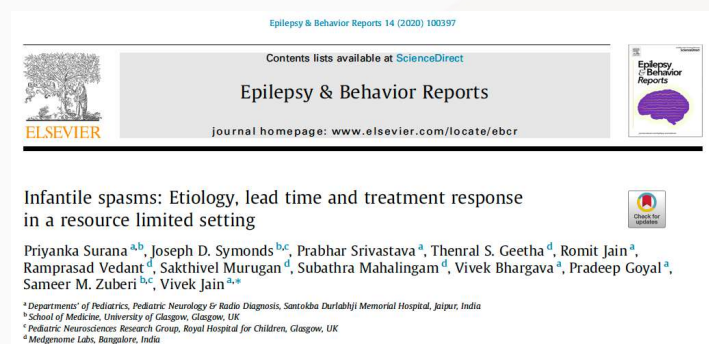
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Infantile Epileptic Spasms (IES) – Lead Time

- Median lead time to treatment was 60 days (in high resource countries it is typically days)
- 80% of people with epilepsy live in low-resource settings. 70% do not receive an appropriate diagnosis and treatment
- AI analysis of smartphone video could be integrated into a clinical decision support tool and support timely diagnosis around the world

Table 3
Lead time to treatment and treatment response.

Lead time	N (%) responding to first line therapy	Cumulative N (%) responding to second line therapy	N (%) with resistant spasms at 3 months	χ^2 test for trend (resistant spasms at 3 months)	N with missing information of treatment response
≤7 days	10/11 (91%)	11/11 (100%)	0/11 (0%)	$(\chi^2 = 10.0 (1 \text{ d.f.}), p = 0.0015)$	1
8-14 days	9/16 (56%)	15/16 (94%)	1/16 (6%)		0
15 days to 2 months	23/44 (52%)	30/44 (68%)	14/44 (32%)		2
> 2 months	19/38 (50%)	23/38 (61%)	15/38 (39%)		0

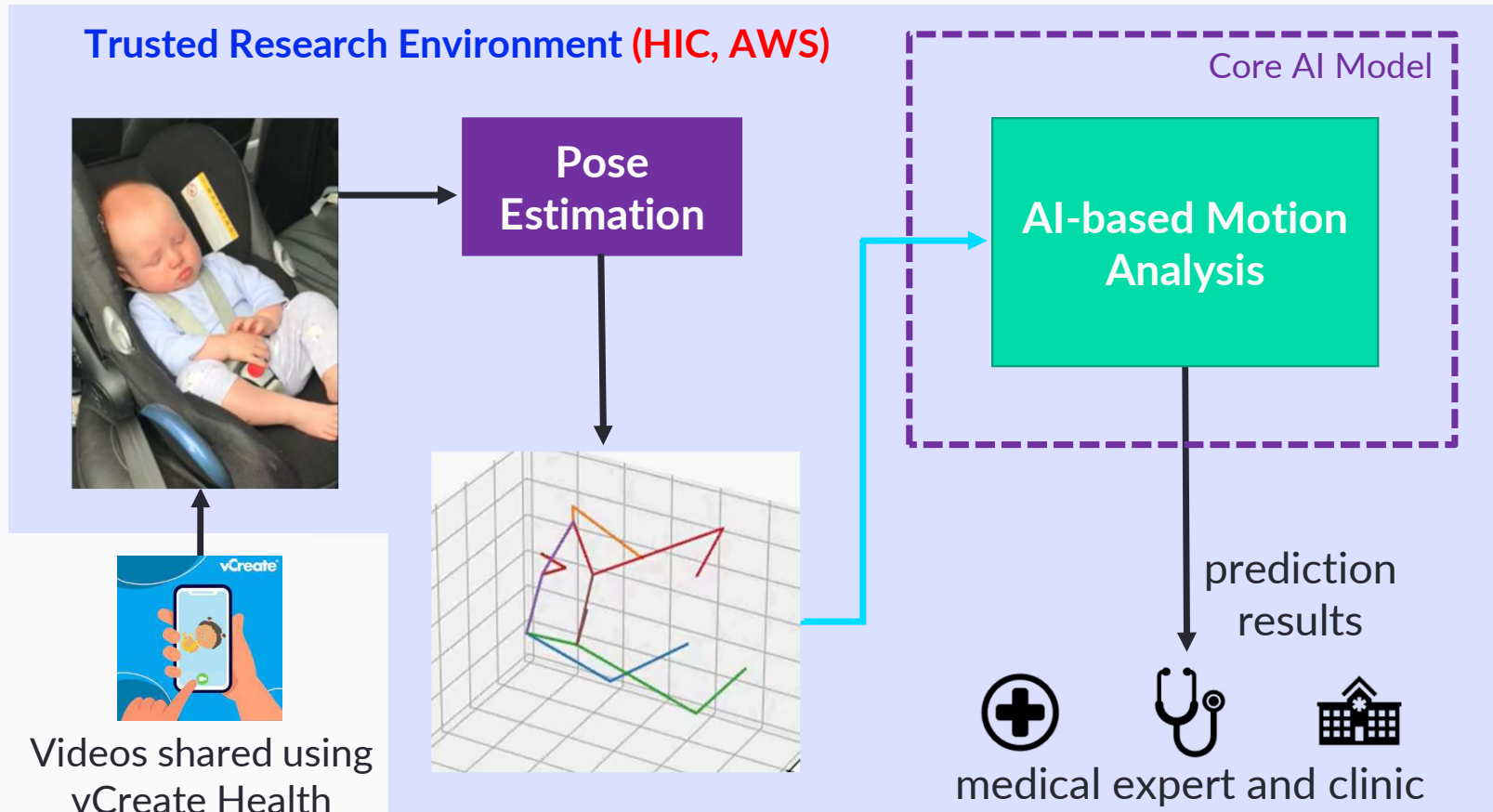


Infantile spasms: Etiology, lead time and treatment response in a resource limited setting

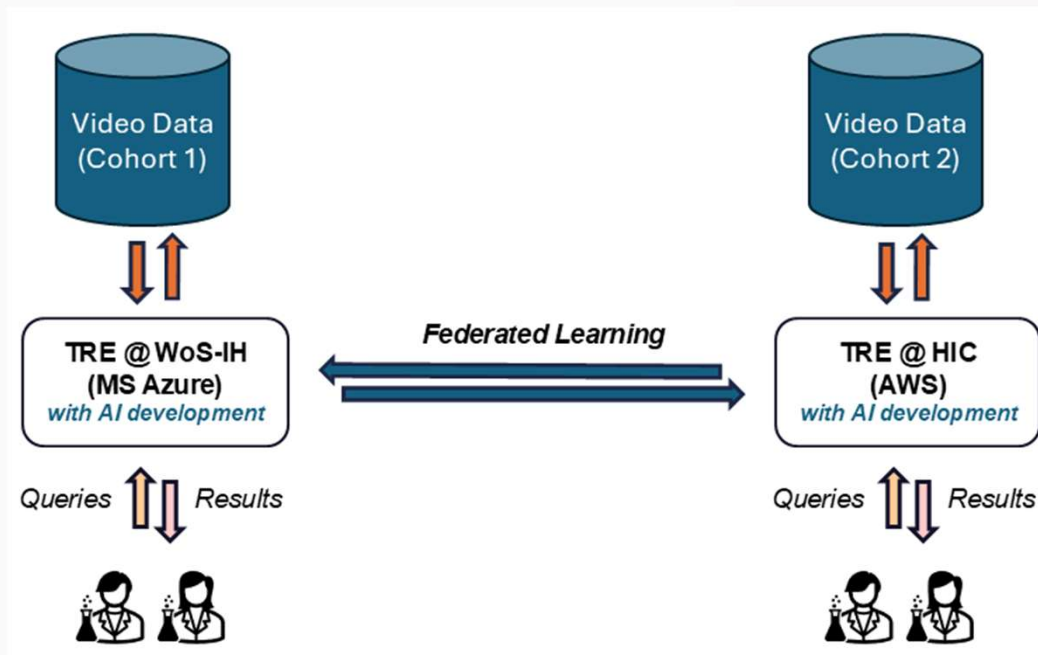
Priyanka Surana^{a,b}, Joseph D. Symonds^{b,c}, Prabhar Srivastava^a, Thenral S. Geetha^d, Romit Jain^a, Ramprasad Vedant^d, Sakthivel Murugan^d, Subathra Mahalingam^d, Vivek Bhargava^a, Pradeep Goyal^a, Sameer M. Zuberi^{b,c}, Vivek Jain^{a,*}

^a Departments of Pediatrics, Pediatric Neurology & Radio Diagnosis, Santokba Durlabhji Memorial Hospital, Jaipur, India
^b School of Medicine, University of Glasgow, Glasgow, UK
^c Pediatric Neurosciences Research Group, Royal Hospital for Children, Glasgow, UK
^d Medgenome Labs, Bangalore, India


Our Work - Detecting Infantile epileptic spasms (IES) from smartphone videos using AI





Project Overview and Objectives – upscaling our IES detection through Federated Learning across multiple TREs



 To evaluate the feasibility of federated learning between TREs on different cloud vendor platforms.

 To evaluate the feasibility of federated learning between TREs containing patient data from distinct cohorts.

 To evaluate the most feasible and secure federated learning options within the DARE ecosystem.

 To quantify the benefits of federated learning on our IES detection machine learning model through uncertainty estimation in the prediction results.

Enablers and Our Progress



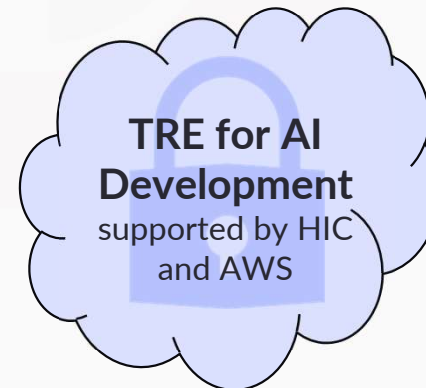
K8TRE – SATRE compliant TRE

Interoperability of TREs (eg with providers such as MS Azure, AWS, etc)



SACRO ML

Providing risk assessment on trained models



Patient, Public and Professional Involvement and Engagement (PPPIE) activities

- Established partners and ongoing PPPIE activities (since 2022) to inform our video-based AI research projects
- Proposed project activities
 - **Inaugural Workshop** – introduce our project and take early feedback
 - **Educational Workshop** - demystify TRE and AI, understand their views
 - **Research Showcase** – disseminate outcomes, encourage future engagement

Our Interdisciplinary Team



Edmond Ho¹

Senior Lecturer in
Machine Learning



Sameer Zuberi^{1,4}

Consultant
Paediatric
Neurologist



Neil Patel^{1,2,4}

Consultant
Neonatologist



Richard Walls³

HIC Operational
Director



Isla Birnie¹

PIE Lead
Grant Manager



Stay in touch! – QR codes

