

Transforming patient outcomes and clinical workflows with AI



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Mission

The National Health Service (NHS) is the publicly funded healthcare system in England. It is the second largest single-payer healthcare system in the world. The NHS provides healthcare to all residents, with most services free at the point of use . The NHS also conducts research through the National Institute for Health and Care Research (NIHR).



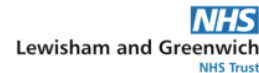
The AI Centre for Value Based Healthcare is pioneering AI technology for the NHS. AI will help the NHS increase efficiency, improve therapies, drive safety, and reduce costs all of which benefit patients, medical staff, and the wider society.



AI Centre Partners



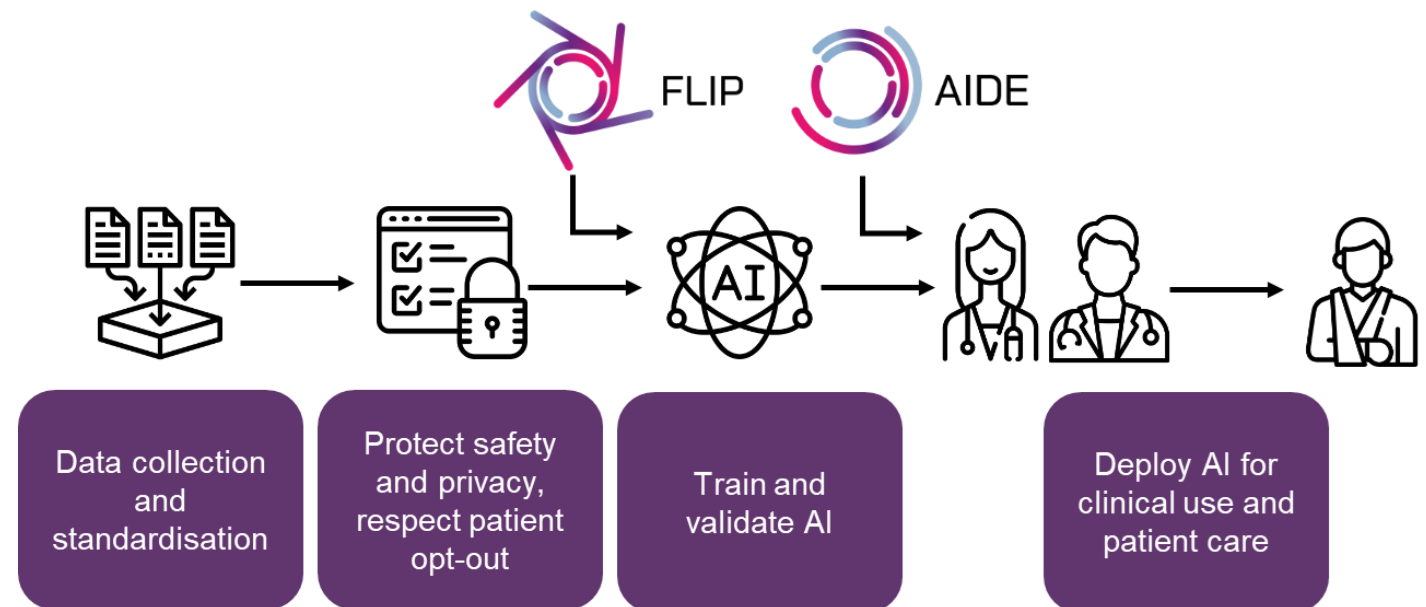
- 10 NHS organizations
- 17 Industry partners
- 4 Universities



Supporting the whole AI lifecycle

These in-house capabilities in our network and through the AI Centre - established a pipeline for supporting the end-to-end AI lifecycle. As part of our commitment to supporting the development and implementation of robust, secure, and trustworthy clinical AI we are also carrying out pioneering work in establishing ethics and data governance models.

- Collect and Standardise Data
- Safety and Privacy, respecting patient opt-out
- Training and Validating AI
- Deploying AI for patient care
- AI marketplace
- Standards (MHRA/BSI/MONAI)
- Scaling across the NHS



Infrastructure

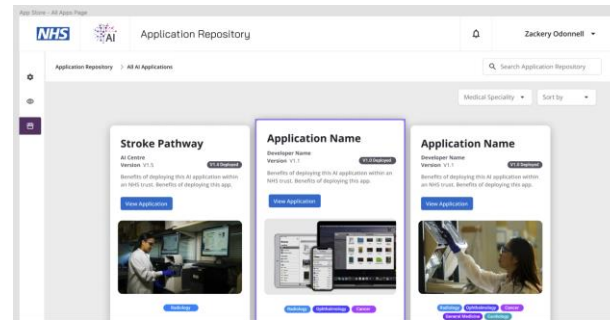
1) Hardware Infrastructure



Hardware already installed in 7 NHS orgs for training and running AI models

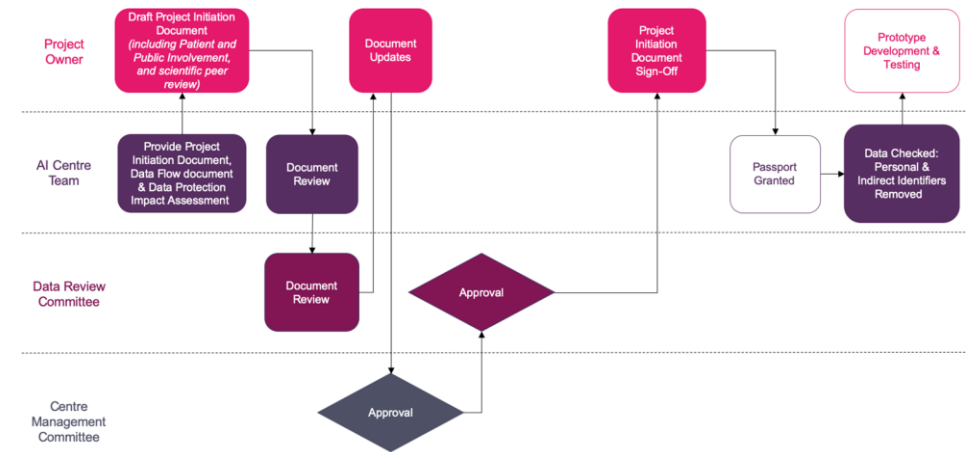


2) Software Infrastructure



Software platforms developed and installed to enable Federated learning and AI deployment

3) Regulatory Infrastructure



Approval processes for access to patient data for research purposes for our partners

Developing standards and policies

As AI technologies are still emerging – standards are being developed and shared across the healthcare ecosystem to ensure consistency, interoperability and security.

KCL and GSTT had been part of setting up the international collaboration with industry, academia and healthcare organisations - for *Medical Open Network for Artificial Intelligence* – [MONAI](#).

MONAI aims to close the gap from research and development to clinical production environments by bringing AI models into medical applications and clinical workflows -to improve patient care.

<https://docs.monai.io/projects/label/en/latest/whatsnew.html>

Fellowship in Clinical Artificial Intelligence

Technical skills, and data science/AI expertise in the workforce, are essential to a data enabled future for healthcare services.

Increasing numbers of clinicians are looking for 'portfolio' careers alongside their medical practice.

Funded by Health Education England, the Fellowship in Clinical AI is a year long programme which is integrated part-time alongside clinical work. Fellows are recruited from a diverse clinical workforce including: medical and dental specialty trainees, nurses & midwives, allied health professionals, and pharmacy professionals.

Fellows learn to adopt clinical AI technology, they gain experience deploying AI in clinical workflows in 12-month project placements, under expert supervision in multidisciplinary teams.

11 / 11

“ I am developing a clinical AI decision support system which enhances the acquisition of cardiac MRI. Translating AI science into practice is challenging; this Fellowship will give me a transferable skillset for deploying and implementing such projects. Looking ahead, I'd like to take on a hybrid cardiologist-data scientist role in the NHS and assist in delivering the AI Roadmap. ”



KAVITHA VIMALESVARAN
ST6 Cardiology



Transforming Patients outcomes

AI technologies provide opportunities to target diverse clinical pathways, from head to toe, in early life and old age. Our AI interventions create applications that enable faster and earlier diagnosis, automation of reporting, improved patient screening for disease, and personalised therapies.

AIDE platform

https://www.youtube.com/watch?v=7R2dLu2iG_s

start 9:33 ends 11:28

Model outputs for clinical review

Zackery Odonnell

Work List

Search task...

Filter

Sort

10/06/2021 · v1.2

Patient: Marcus Rashford

Name: Stroke Pathway Model

Mode: QA

10/06/2021 · v1.2

Patient: Marcus Rashford

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Mode: QA

10/06/2021 · v1.2

Patient: Marcus Rashford

Name: Stroke Pathway Model

Mode: QA

10/06/2021 · v1.2

Patient: Marcus Rashford

Name: Stroke Pathway Model

Mode: QA

Mode: QA - Not for clinical use

View Pipeline

REJECT

APPROVE

Patient: Marcus Rashford

Hospital ID: 11111111 (NHS ID: 0123456789)

Gender: Male

DoB: 24/08/1994

Study Date: 24/24/08/2003

Series



CT
Lorem ipsum S... (673)



MR (730)



Modality: DOC (673)



Hide Image Metadata

Manufacturer
Picker International, INC.

Institution Name
Boston Med Cent E.N.C

Conversion Type
WSD

Study Description
<VYHR> Path:dicom

Study Time
074737.0000000

SOPClassUID
1.2.840.10008.5.1.4.1.1.7

50

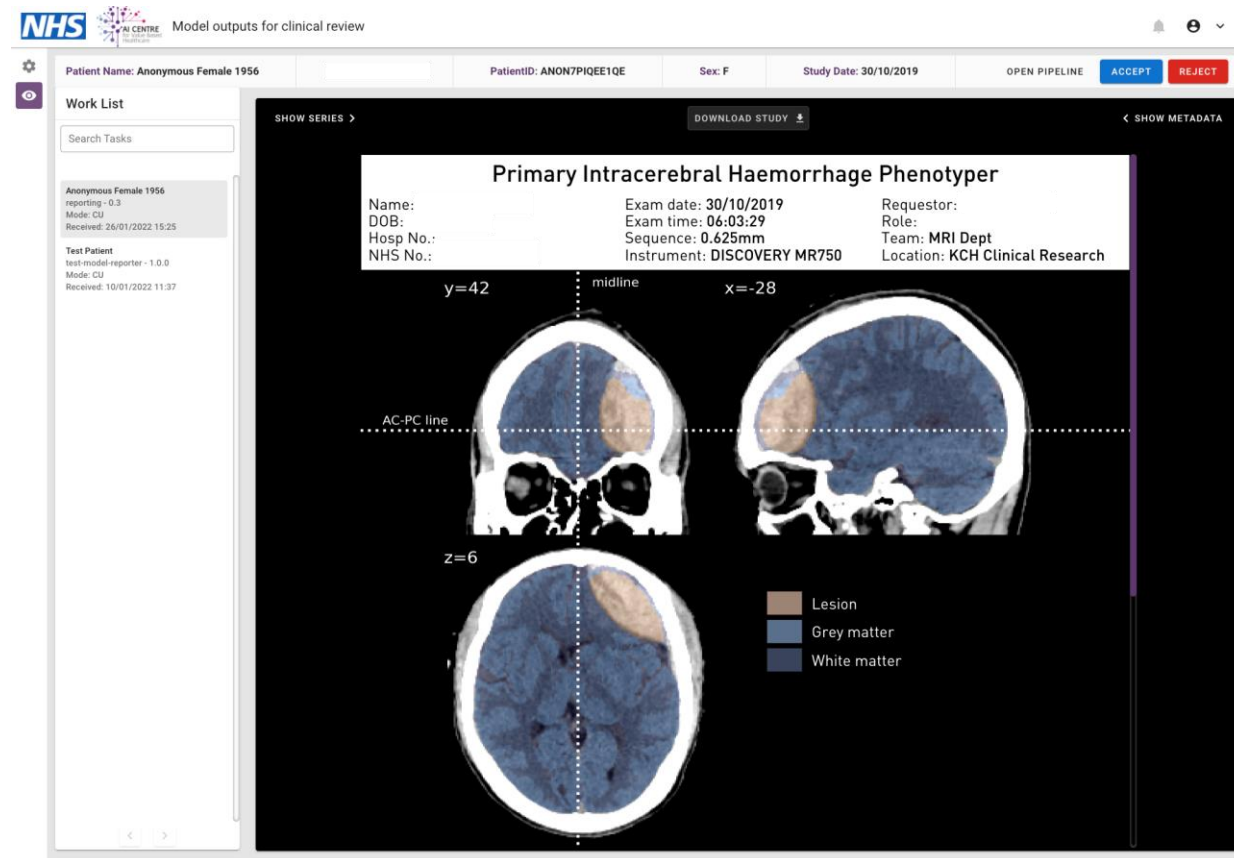
AI assisted Stroke Detection

Intracerebral haemorrhage Phenotyper (ICH)

Problem The severity of a stroke is highly dependent on which part of the brain is damaged. Simple measurements like stroke size are insufficient to predict outcomes. The complex relationship between stroke anatomy and outcome is a major obstacle to optimal care, especially in the initial hospital phase of most critical interventions.

Solution Developed in-house by KCL and KCH - Detects which parts of the brain are affected by the haemorrhage and the likely side effects, as a way to inform downstream treatment, including specific therapies needed for stroke rehabilitation.

Status It has been used in KCH hospitals for decision support since 2022, currently in use by clinical teams.





Work List

Search Tasks

Anonymous Male 1980
reporting - 0.3
Received: 09/16/2021 17:29

Anonymous Female 1956
reporting - 0.3
Received: 09/16/2021 16:58

Mode: CU

OPEN PIPELINE

ACCEPT

REJECT

Name: Anonymous Female 1956

Birth Date: 03/26/1956

Patient Id: ANON7PIQEE1QE

Sex: F

OPEN REPORT OPEN IN ORTHANC

UNDEFINED [20191030]

ORTHANC



20191030



Anonymous Female 1956
ANON7PIQEE1QE
19560326

20191030
3 | 0.625mm



Instance number: 72





Work List



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Anonymous Male 1980
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Anonymous Female 1956
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REJECT

Name: Anonymous Female 1956

Birth Date: 03/26/1956

Patient Id: ANON7PIQEE1QE

Sex: F

OPEN REPORT ⌵ OPEN IN ORTHANC ⌵

UNDEFINED [20191030]

ORT HANC

Primary Intracerebral Haemorrhage Phenotyper

Name: Anonymous Female 1956 Exam date: 30/10/2019 Requestor: Ione Pelton
 DOB: 26/03/1956 Exam time: 06:03:29 Role:
 Hosp No.: Sequence: 0.625mm Team: MRI Dept
 NHS No.: Instrument: DISCOVERY MR750 Location: KCH Clinical Research

y=42 midline x=-28

AC-PC line

z=6

- Lesion
- Grey matter
- White matter

Global lesion parameters	Focal grey involvement	Focal white involvement	Focal CSF involvement
Component: 2, 6 Vol: 172.00	Component: Lat: N 0% Det: Mid Front Gyral Grd: 20x1 0% Mid Front Gyral P: 20x1 20x1 0% Mid Front Gyral: 20x1 0%	Component: Lat: N 0% Det: Mid Front Gyral: 20x1 0% Mid Front Gyral: 20x1 0% Mid Front Gyral: 20x1 0%	No quantified focal CSF involvement



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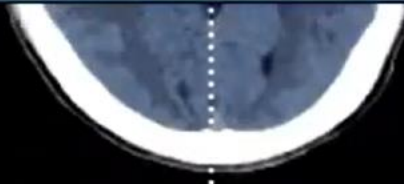
OPEN IN ORTHANC

UNDEFINED [20191030]

ORT HANC



20191030



Global lesion parameters

Centre of mass: -28, 42, 6
Volume: 72.3 ± 11.8 ml (72nd)

Tissue distribution	%	Cent
Grey	35±2	35 th
White	7±1	7 th
CSF	7±1	7 th

Global tissue parameters

Compartment	ml	Cent
Grey	676	100 th
White	457	100 th
CSF	213	100 th
Bone	654	100 th
Soft tissue	1376	100 th
Cranium	1367	100 th

Focal grey involvement

Compartment	Lat	% hit	Cent
Mid Front Gyrus Orbit	L	50±3	50 th
Inf Front Gyrus P Triang	L	50±3	50 th
Mid Front Gyrus	L	33±4	33 rd
Inf Front Gyrus P Orbit	L	29±2	29 th
Sup Front Gyrus Orbit	L	27±1	27 th
Sup Front Gyrus	L	19±2	19 th

Focal white involvement

Compartment	Lat	% hit	Cent
Inf Front-Occipital Fasc	L	25±1	25 th
Ant Thalamic Project	L	23±1	23 rd
Uncinate Fasc	L	12±0	12 th
Sup Longitud Fasc III	L	11±0	11 th
Sup Longitud Fasc II	L	6±1	6 th
Cingulum Ant	L	4±1	4 th
Corpus Callosum	-	4±0	4 th
Sup Longitud Fasc I	L	4±1	4 th

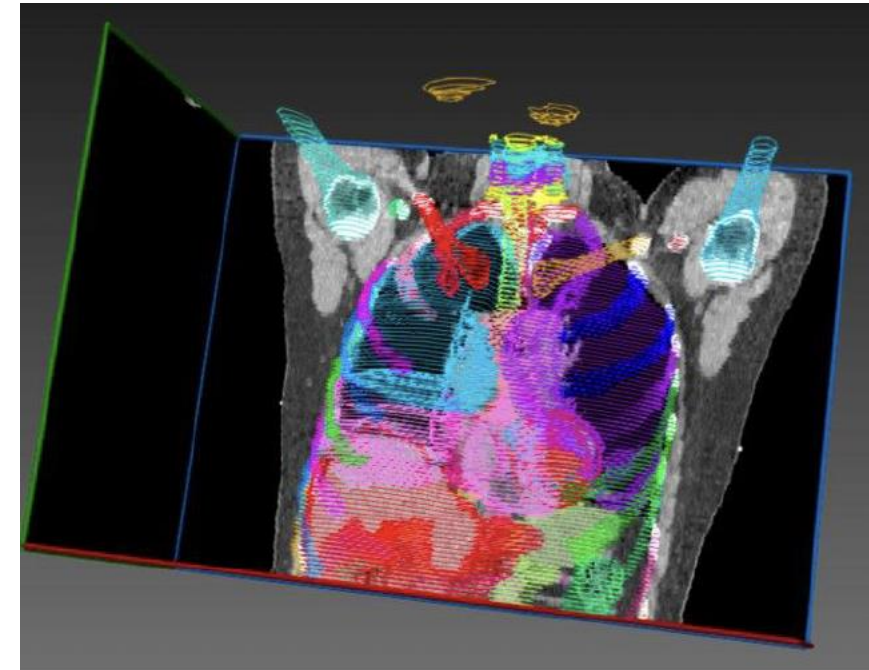
Focal CSF involvement

No quantified focal CSF involvement

AI for automated radiotherapy contouring

AutoSegCT–Auto-contouring for radiotherapy

- Problem** Manual contouring can take 30mins –3 hours and is heavily reliant on staff focus and time.
- Solution** AutoSegCT tool is developed based on open-source code. It enables robust segmentation of >100 important anatomical structures in CT images. The resulting RTStructureSet DICOM is returned to the radiotherapy treatment planning system (TPS) within 20 mins of receiving the original CT. they are subsequently used for radiotherapy treatment planning purposes, such as beam optimisation to target dose only at the tumour and high-risk regions of interest and minimise the dose to healthy organs.
- Status** Used in Radiotherapy Oncology at Guy’s and St Thomas since July 2024 by radiotherapy teams – significantly reducing clinician workload and contouring turnaround time by 75%.



Automated clinical trials feasibility

Problem

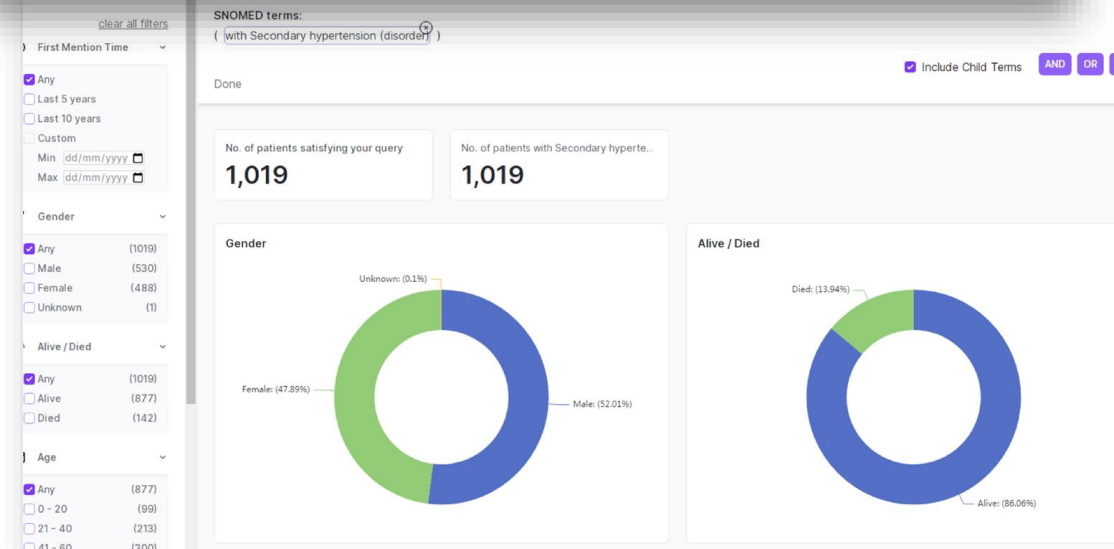
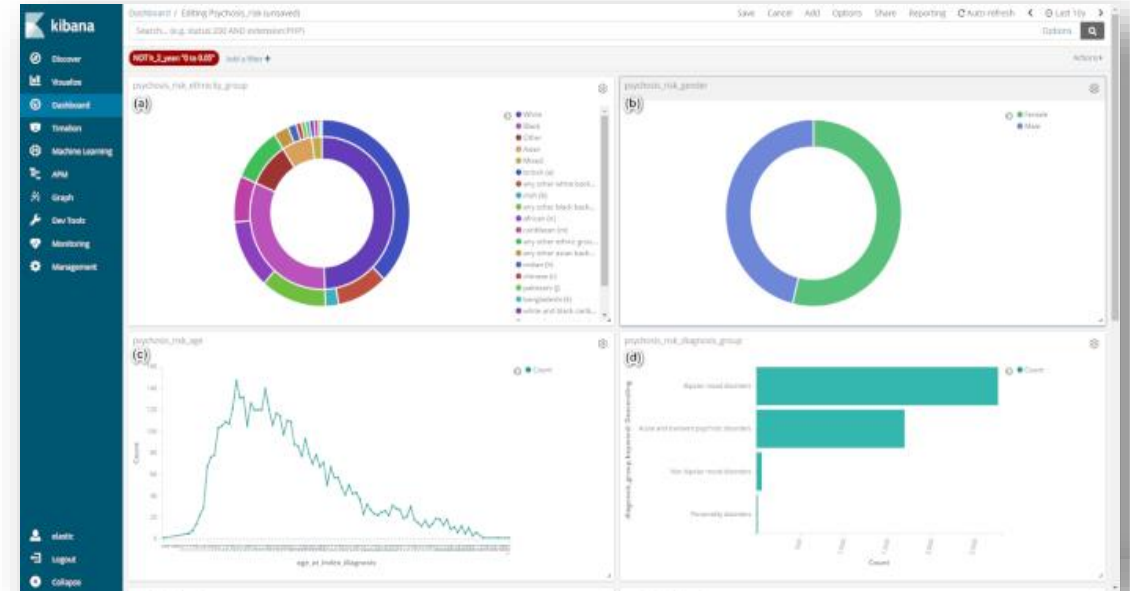
For clinical researchers and pharma companies, finding the patients that are eligible to participate in trials is critical. Patients have specific inclusion criteria, which include structured data findings and indicators that are typically found in the unstructured data.

Impact

A large and more accurate cohort of patients are identified, enabling 'instant feasibility' and shortening lead times for patients participating and benefitting from new drugs and treatments.

Solution

We use 'CogStack Watcher' model to track defined criteria, and surface as dashboards or fed back to the Electronic Health record (EHR) to be flagged automatically.



Waiting List Prioritization- Rheumatology

Analysis of Clinical Documentation to enhance structured data

Problem

- Demand for Rheumatology Service review outstrips capacity, the patient scheduling processes are not robust enough to ensure patients who are clinically in need of follow up are prioritized for booking.
- Structured data held on hospital core systems is not complete enough to provide reliable scoring for prioritization.

Solution

- Clinical Prioritization Tool has been developed enabling scoring of patients based on the clinical priority criteria.
- Series of diagnosis terms and vulnerable patients are flagged in the system.
- Clinic letters are analyzed for presence of the flags enabling more accurate classification of patients

Impact

- Patients with urgent needs are prioritised and are seen and treated quicker
- The automation enables clinicians to spend more time delivering direct patient care.



NHS 'in-house' AI capabilities



Dr Hammad Khan, Neonatologist
Guy's and St Thomas' NHS Foundation Trust

“GSTT is a leading centre of tertiary neonatal care and looks after (approximately) 250 very preterm infants every year. All of whom are at risk of abdominal pathology.

X-ray is our current best imaging modality for diagnosis. Abdominal x-rays are interpreted by neonatal staff in urgent medical situations, but reported non-urgently by radiologists.

We want to build a machine learning model using large data sets of neonatal abdominal x-rays for the most common diagnoses – i.e. NEC, perforation.

Having our in-house clinical data scientists, means we are able to create an algorithm that identifies the abdominal perforation and alerts clinical teams straight away.”