Local evaluation of an AI augmented surgical waiting list prioritisation tool

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INTRODUCTION

- In March 2024, 7.5 million patients were on waiting lists for treatment (1).
- Triaging patients at risk of deterioration before treatment could provide clinical benefits.
- C2-AI have developed software to support waiting list triage through scoring patients on a variety of parameters, including historical diagnoses.
- All hospital inpatient stays are coded for financial renumeration, healthcare analysis and service planning in the form of Hospital Episode Statistics (HES).

PREVIOUS ADMISSIONS

- A key limitation are outpatients with no previous inpatient admissions and therefore no coded past medical history in HES available.
- Using Python, Jupiter, Pandas packages the following process was followed to ascertain number of patients with hospital admissions.



TECHNOLOGY

C2-AI utilises inputs from HES and Referral to Treatment Time (RTT) waiting lists, both in CSV format, provided by the hospital on a weekly basis.
C2-AI calculates scores from this data, using ICD-10 codes from previous admissions within the HES data as risk predictors.
The software is currently pending registration as a Class IIb medical device.
The UHBW present the scores as a dashboard within PowerBI (Fig3)



Fig. 1. Diagram depicting the data sources and output required for the triage software

WEIGHTED MATRIX SCORING SYSTEM for Triage

- Core product output
- Risk matrix score applied to each patient on outpatient surgical waiting list.
- Six Parameters
- Allocated weighting according to consensus clinical opinion in organisation.
- Highlighted are the parts of the risk score which use a statistical model (the AI component)
- Aims to prioritise patients based on a risk of increased harm should they deteriorate while waiting.

Dashboard

system.

23.51

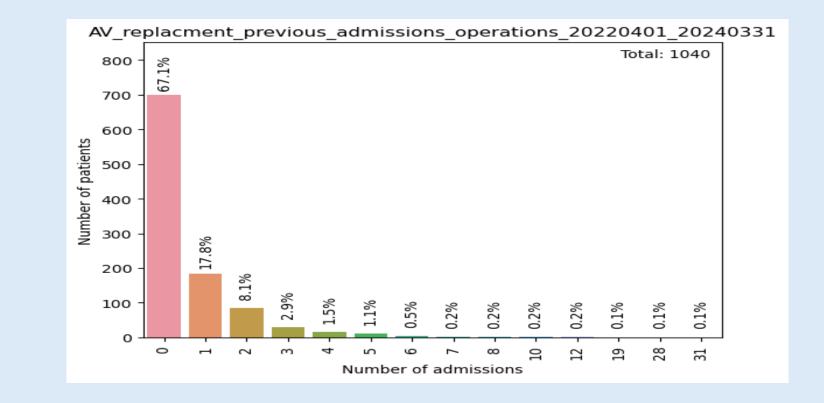
Fig. 3. Dashboard presenting the C2-AI scoring

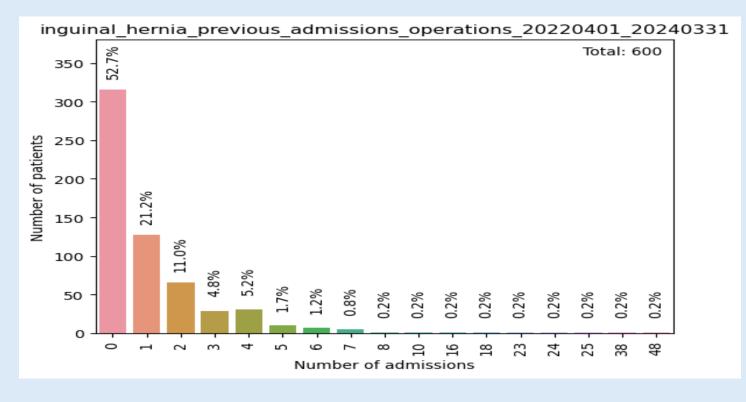
Priority Code (@IBlank) (@P2 (@P3

19.49

| Elective Procedure | Number of patients (2022 – 2024) | Proportion with no past medical history (2020-2024) |
|-------------------------------------|-------------------------------------|---|
| Aortic Valve replacement | 1024 | 67.1% |
| All elective lower GI procedures | 14704 | 36.2% |
| Colectomy | 256 | 9% |
| Cholecystectomy | 1124 | 20.8% |

Table 1. Results of filtered patients undergoing specific procedures and counted admissions prior to their operation





• Up to 30 of the maximum total 100 Points

| Risk Matrix Points | 4. Deterio | 4. Deterioration | |
|------------------------------|-----------------|------------------|--|
| Maximum 100 Points | possible | possible | |
| 1. Waiting time | o Yes | 10 | |
| ○ >52 weeks 20 | 0 No | 1 | |
| o 26 - 52 weeks 15 | 5. Change | in mortality | |
| • 17 - 26 weeks 10 | ○ >10.0 | 1% 10 | |
| \circ 5 – 16 weeks 5 | o 5.01 - | · 10% 7 | |
| 0 0 - 4 weeks 1 | 0 1.01 - | -5% 5 | |
| 2. Surgeon priority | 0.01 - | -1% 3 | |
| Urgent / | o 0% | 0 | |
| P2 30 | 6. Change | in | |
| • Soon / P3 15 | complicat | ion | |
| o Routine / P4 1 | ○ >20.0 | 1% 10 | |
| 3. Procedure priority | 0 10.01 | - 20% 7 | |
| • P2 20 | o 5.01 - | · 10% 5 | |
| • P2/3 15 | 0 2.01 - | · 5% 3 | |
| • P3 10 | 0.01 - | 2% 1 | |
| • P3/P4 5 | o 0% | 0 | |
| • P4 1 | | | |
| | | | |

Fig. 2. Table of the C2AI risk matrix points.

AI COMPONENT

- The change in mortality and morbidity rates are based on a statistical model using previous coded patient co-morbidities collected from HES data.
- Two complication / mortality risk calculations are made:

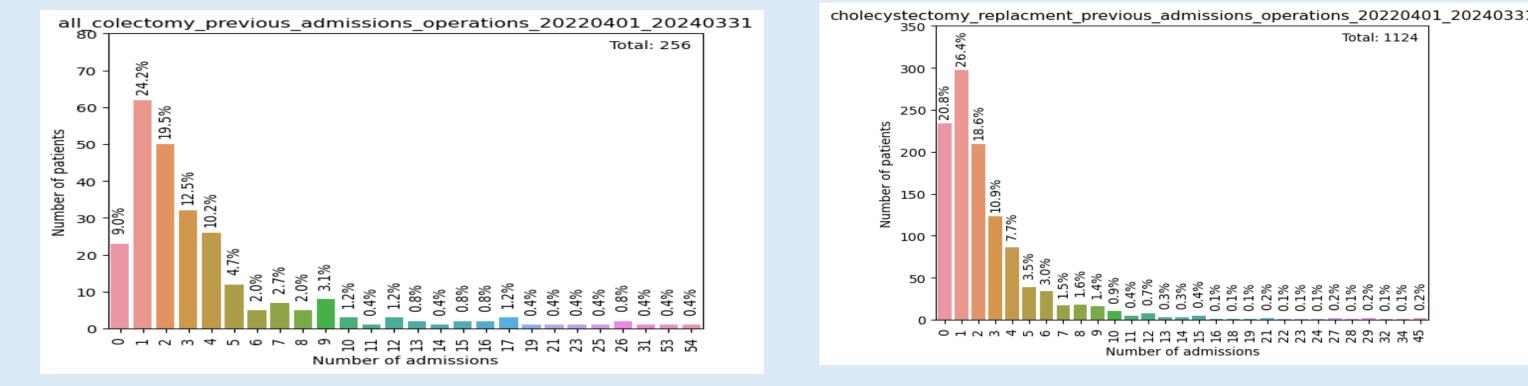


Fig. 4. Bar charts depicting the number of prior admissions within 2020-2024 for four operation groups.

DISCUSSION

- A significant minority of surgical patients cannot be accurately allocated a risk score due to missing medical history data.
- This could be due to:
 - Tertiary or Quarternary referral from their main medical center.
 - $\circ~$ No previous admissions in the time period.
 - Patients moving to the area recently.
- A solution given to this could be all hospitals in a geographic locale buying and implementing the product.

PLAN FOR FURTHER EVALUATION

- In order to assess the model's predictions of post operative complications further work is being undertaken.
- Baseline risk for the patient and procedure when admitted electively.
- Additional risk for the same patient and procedure when admitted and operated as an emergency .

LIMITATIONS

- The AI component of the model requires HES data for each patient. If this data is unavailable, the risk score cannot be personalised to the patient.
- The AI scoring component calculates increased mortality and morbidity for patients should the present as emergencies needing surgery, cannot not provide the probability of such an emergency admission occurring.
- The model generates a matrix of scores, with the aggregate score determined by a locally defined weighting system. This allows for the creation of millions of customized versions of the tool.

- Cholecystectomy had the greatest number of patients with a previous admission.
- HES data for their operative admission and 30 days afterwards was collected with analysis of post operative complications undertaken.

References:

1)

www.england.nhs.uk. (n.d.). *Statistics» Consultant-led Referral to Treatment Waiting Times Data 2023-24*. [online] Available at: https://www.england.nhs.uk/statistics/statistical-work-areas/rtt-waiting-times/rtt-data-2023-24/.

