Faculty of Natural and Mathematical Sciences
Department of Informatics

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#### 7CCSMPRJ

### Individual Project Submission 2024 - 2025

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Degree Programme: MSc. Advanced Computing

Project Title: GluCORRECT - Harnessing Artificial Intelligence to

scrutinize Hypoglycemia in hospitalised patients with

diabetes to classify, anticipate and analyse hypoglycemic

episodes [Knowledge Exchange Project with NHS England]

**Supervisor:** Dr. Rita Borgo

Word Count: ==== Word count goes here ====

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Signature: Signature Date: August 5, 2025



## Department of Informatics King's College London United Kingdom

7CCSMPRJ Individual Project

GluCORRECT - Harnessing Artificial
Intelligence to scrutinize Hypoglycemia in
hospitalised patients with diabetes to classify,
anticipate and analyse hypoglycemic episodes
[Knowledge Exchange Project with NHS
England]

Name: **Siddharth Kishor Samarth** Student Number: K24012370 Course: MSc. Advanced Computing

Supervisor: Dr. Rita Borgo

This dissertation is submitted for the degree of MSc Advanced Computing.

## Acknowledgements

I would like to express my sincerest gratitude towards my project supervisor, Dr. Rita Borgo, for her invaluable advice and consistent direction throughout the course of this project. Her mentorship and ideas have been instrumental in shaping the development of this work, leading to its successful completion.

I am also deeply thankful & appreciative of my industry advisor, Dr. Piya Sen Gupta, for providing the dataset that has served as the foundation of this work. Her contributions have significantly enhanced the practical relevance and quality of this project.

Ultimately I would like to thank my friends and my parents, especially my dad, without whose sacrifices I would not be where I am today.

### Abstract

<u>Project Variant:</u> Variant 4 - Develop a weighted score and design score to predict risk of a hypoglycaemic episode before it occurs.

It is well known that hypoglycemia as well as hyperglycemia are common adverse events in patients who receive blood sugar control medication, and they are also one of the most frequently cited causes of hospital admissions in people with diabetes. National quality improvement programmes from the Healthcare Quality Improvement Partnership (HQIP) and the study of ambulance call-out data have shown that lack of awareness by both affected individuals and their attendants is associated with a dramatically increased rate of complications, amongst other factors. Guy's & St. Thomas' NHS Foundation Trust (hereafter referred to as GSTT) has found, after careful deliberation and departmental review, that hypoglycemic episodes have been occurring with unusual frequency. The Trust now seeks to take measures to resolve such problems with a greater focus on prevention combined with early corrective action. This research project has been undertaken in close collaboration with GSTT, one of the largest NHS trusts in the UK and an indispensable element of London's healthcare system, with almost 24000 staff across 5 major hospitals, handling over 3 million patients a year and generating an annual turnover of over £3 billion.

This analytical study serves as a foundation and proof-of-concept to aid GSTT in preemptively reducing hypoglycemia within hospitalised inpatients, by utilising statistics & machine learning techniques. Through exploratory data analysis I draw out relevant conclusions about the dataset around patient age, ethnicity and

I go on to identify the significant factors responsible for hypoglycemia within the dataset provided by the industry advisor from GSTT through exploratory data analysis, while also . I explore how they can be utilized to devise a risk score, to classify patients based on their risk of hypoglycemia.

RESULTS: HbA1c values identified as risky - 56 or so eGFR ethnicity major is glucose value

In conclusion, exhibit my findings with potential ways of applying them in practise in hospitals.

All abbreviations and symbols used in the report must be listed and defined in alphabetic order.

## Nomenclature

GSTT Guy's and St Thomas' NHS Foundation Trust HQIP Healthcare Quality Improvement Partnership

"Hypo" or "Hypos" Hypoglycemic episode(s)

"Inpatient" Referring to the the fact that a patient is required to stay overnight

in order to be treated (in case of surgeries or long term observation for example)

NCAPOP National Clincal Audit & Patient Outcomes Programme

NDA National Database Audit

NDISA National Diabetes Inpatient Safety Audit

NHS The publicly funded healthcare system of the United Kingdom,

the National Health Service.

a The number of angels per unit area

A The area of the needle point

c Speed of light in a vacuum inertial frame

h Planck constant

LMI Linear Matrix Inequalities

N The number of angels per needle point

## Contents

1	Introduction	1
	1.1 Clinical Overview	1
	1.2 Background	1
	1.3 Aims and Objectives	3
	1.4 Report Structure	3
	1.4.1 Dissertation Length	4
<b>2</b>	Literature Survey & Review	5
3	Dataset	6
4	Methodology and Implementation	8
5	Main Results and Findings	9
	5.1 Interpretation Analysis and Evaluation	9
6	Legal, Social, Ethical and Professional Issues	9
7	Others	9
	7.1 Maths	9
	7.2 Glossary and acronyms	10
	7.3 Figures	10
	7.4 Table	11
8	More Others	12
	8.1 What is calibration?	12
	8.2 Numerical methods for calibration	12
9	Legal, Social, Ethical and Professional Issues	13
10	Conclusion	14
Re	eferences	15
$\mathbf{A}$	Appendix	17
	A.1. Detect	17

# List of Figures

1	This is the caption for the figure	10
1		
$^2$	This is the caption for the figure which is not even present	10
3	Another caption	11
4	Raw dataset	17
5	Dataset with cleaned features (this is in addition to the fields of the raw	
	dataset)	18
$\mathbf{List}$	of Tables	
1	Random data for a table	11

### 1 Introduction

#### 1.1 Clinical Overview

Hypoglycaemia (also known as a "hypoglycaemic episode" or a "hypo" for short) is the condition that occurs when the human body's blood glucose (sugar) level drops below the normal healthy range of 4.0 to 6.0 mmol/L. While it can affect anyone, it is most common in diabetic individuals who are prescribed drugs like insulin or metformin to inhibit glucose. Hypoglycaemic events are relatively simple and straightforward to resolve, but they need to be treated immediately to avoid serious damage to the brain and heart as a result of loss of consciousness or arrhythmias. High-sugar consumables are generally effective in correcting mild cases and are commonly recommended for immediate treatment, but severe cases of hypoglycaemia such as when the person is unconscious or having a seizure can only be resolved with an urgent, immediate glucagon injection to prevent them from deteriorating into a coma (or in rare cases, even leading to death).

To underscore how and why this matters, diabetes is one of the most significant and expensive long-term health conditions faced by the NHS, with recent figures from Diabetes UK suggesting that over 5.8 million people in the UK are living with diabetes, regardless of a formal diagnosis. It is estimated to cost the NHS over £10.7 billion a year, approximately 10% of its entire annual budget, which could go up to £18 billion by 2035 [1]. A stark finding is that almost 60% of this cost (around £6.2 billion) is spent on treating the largely preventable complications of diabetes, such as heart attacks, strokes, blindness, and so on, including hypoglycaemia [2]. Hypoglycaemic instances make up a major component of these preventable costs, mainly accounting for the emergency, ambulance, and acute care expenses associated with diabetes. The Local Impact of Hypoglycaemia Tool (LIHT) suggests that hypoglycaemia can cost up to £2,195 per episode, possibly increasing substantially with a longer stay in hospital [3], and it is estimated that there are up to 100,000 ambulance callouts annually according to the Diabetes Research and Wellness Foundation (DRWF) [4]. DRWF's study hinted that 1 in 10 individuals that experience a severe hypo (meaning requiring medical intervention or resuscitation) have considerable chances of another one within a fortnight.

#### 1.2 Background

After introspective analysis supported by information from the National Diabetes Inpatient Safety Audit (NDISA) it has been recognized that severe hypoglycaemia and recurrent severe hypoglycaemia have been occurring relatively frequently across GSTT medical facilities. The NDISA forms part of the National Diabetes Audit (NDA), and it maintains that "The prevalence of diabetes continues to increase. In England

1.2 Background 2

between 2017-18 and 2021-22 prevalence of type 1 diabetes went up from 248,240 to 270,935 and the prevalence of type 2 and other diabetes from 2,952,695 to 3,336,980", as of 2022 [5].

GSTT administers upwards of 500,000 point-of-care glucose tests (POCT) annually, in addition to kidney function and glycated haemoglobin (HbA1c) tests as well. The Trust also possesses blood glucose / ketone data with additional linked data including demographics, dates of admission and discharge, patient as well as family history and current or previous medications. They have two major kinds of patient records, inpatient records for patients that have to stay over the course of one or multiple nights (for example, in case of surgeries or for long term care), and outpatient records where the patient doesn't require overnight stay. The Trust manages all of this data through their electronic health record management system called Epic, and facilitates patient access to their own records through the MyChart web application.

Hypoglycaemia is a frequent complication amongst inpatients having complex health conditions, especially within those in intensive care settings that have been / are critically ill due to advanced diseases or comorbidities, or in patients following major surgical interventions. The Trust is undertaking proactive measures to identify and mitigate the risk of hypoglycaemic episodes at an early stage, to support better planning, reduce healthcare costs, efficiently allocate hospital resources and also schedule operations optimally. The ideal way to assess risk would along the lines of developing tools to predict individualized risk scores for inpatients after considering all relevant factors. However, this presents a herculean task due to the sheer volume and complexity of factors involved, compounded by the challenges of producing reliable results even within small populations — such as those in remote areas — while also adhering to legal and governmental regulations:

- 1. Weighing up the risk of hypoglycaemia depends upon numerous aspects such as lifestyle, renal function, recent food intake, blood glucose history and current medication to name a few, making this a highly complicated modelling problem. In addition to this, patients differ widely in age, comorbidities, ethnic factors and even insulin sensitivity. This variability makes it a formidable challenge to develop a model that is generalizable, dependable and unbiased.
- 2. Any such analytical tool in the vicinity of patient healthcare requires medical evaluation and approval, validation trials, governance oversight as well as ethical considerations. Even a good model may fail if it does not fit the clinical workflow. Initial skepticism towards AI, the effort required to train staff, defining clear responsibilities and limits of liability, and rehearing procedures or plans of action for every possible scenario will all produce appreciable organizational inertia.

Successfully implementing even a small-scale solution, within GSTT to begin with, would be a significant strategic breakthrough that serves as a foundational model which other

NHS trusts or institutions could adapt and build upon. This positions this research initiative which is a Knowledge Exchange Project (KEP) with Guy's & St.Thomas' NHS Foundation Trust, an indispensable constituent of London's healthcare system, as a valuable and worthwhile research endeavour.

### 1.3 Aims and Objectives

This research project has the following objectives:

- To extract insights from provided dataset for the given time period and population. GSTT has expressed a strong interest towards gaining a deeper understanding of their inpatient population. The dataset they have provided includes demographic details, length of hospital stay, and ward information in addition to the main clinically relevant variables such as glycated haemoglobin levels, renal function measurements, patient age and so on. This enables a comprehensive, multifaceted analysis. The knowledge gained from this study, such as identifying which hospital wards have more vulnerable or at-risk patients, will be used to enhance staff training, in turn improving both future admissions routines as well as post-discharge support for patients. Every observation, regardless of scale, holds potential to refine hospital processes and operating procedures.
- To identify the main influencing / contributing factors for hypoglycaemia and develop a weighted risk score to predict episodes (Variant 4 KEP). The Trust is establishing and implementing measures to "pre-assess" inpatients to evaluate their risk of a hypoglycaemic episode, which will allow medical professionals to design protocols and policies to prevent episodes from occurring as well as take early remediative action as soon as possible to resolve an episode should it occur. I aim to find data-backed values for the key features responsible for hypoglycaemia, through statistical tests and machine learning algorithms, in order to create a risk score. This risk score can then be applied in hospital to determine the best course of action based on the patient's reason for being admitted.

#### 1.4 Report Structure

Section 2 contains a comprehensive, detailed review of similar research carried out by other universities, teaching hospitals and medical facilities including references to relevant medical literature. I have compared and contrasted datasets used, approaches taken and results obtained.

Section 3 delves deeper into the dataset provided by GSTT, elaborating on the raw features provided and those that were derived from them for analysis.

Section 4 (Methodology & Implementation) outlines the statistical and mathematical theory behind the concepts used for analysis, ranging from machine learning algorithms to hypothesis testing methods.

Sections 5 (Main Results) onwards discuss the main research executed within the project and deliberates on the results achieved

Section 6 (Ethical Professional Legal Social issues)

Section 7 (Conclusion and Applicability)

### 1.4.1 Dissertation Length

This dissertation comprises a total of wordcount XXXX words excluding references and appendices.

## 2 Literature Survey & Review

Various kinds of different prediction models have already been devised and developed for predicting hypoglycemia. Yi Wu and others have systematically compared, and evaluated the applicability of models in clinical practice in a paper in Biological Research for Nursing[1] where it was found that the major predictors were age, HbA1c, history of hypoglycemia, and insulin use. Lin Yang, Zhiguang Zhou have carried out similar research in the Frontiers in Public Health journal[2] uncovering risk factors that could possibly lead to hypoglycemic events, after employing various data driven models based on ML techniques such as neural networks, autoregressive / ensemble learning and such.

In silico proof of concept studies like the one from Zecchin[3] have also been researched to investigate how continuous glucose monitoring short-term glucose prediction algorithms could be exploited to recognise the run up to hypoglycemic episodes, allowing the patient to take appropriate countermeasures to mitigate events. They found that there was a significant reduction in both the time spent in a hypoglycemic event as well as the number of hypoglycemic events.

As this is a Knowledge Exchange Project (KEP) with NHS England I have been provided a real world dataset from GSTT. Medical data is difficult to obtain, and it rarely fits a research objective without needing much modification.

### 3 Dataset

Our industry advisor from GSTT has graciously provided a year's worth of data in multiple .xlsx files, which have been combined into one for the purposes of analysis and research for this project. The **raw fields provided** within the data were:

- UniqueID: Unique identifier for the patient and test, which is just a number. Meant to identify same patients (not personally) when considered together with Order Time, Order Date and Age, as same patients can have multiple blood glucose tests during their stay.
- Order Date: The date when the glucose measurement was ordered or taken
- Order Time: The timestamp at which the glucose measurement was ordered or taken
- Inpatient Admission Date: The date at which the patient was admitted into the medical facility
- Discharge Date: The date the patient was discharged from the medical facility
- Length of Stay: The amount of time the patient has spent in the medical facility in days and hours (for eg. "5d 6h")
- Ward: The ward that the glucose measurement was taken in, usually matches the ward that the patient was admitted to
- Last Lab Test Results: The result of the glucose measurement in mmol/L. Most values in this column are of the format "Manual blood glucose: 8.70 mmol/L" or "POCT Glucose Blood Manually: 2.7 mmol/L".
- Age: Age of the patient at the time of measuring blood glucose in years
- Ethnicity: Specific ethnicity of the patient, values ranging from "South American Columbian" to "Black or Black British Nigerian" to "Other" or even missing.
- Gender Identity: Gender of the patient.
- **HbA1c:** Numerical value of HbA1c in mmol/mol.
- HbA1c Date: Date the HbA1c test was done for that patient
- eGFR: Estimated Glomerular Filtration rate, which is a measurement of how well the kidneys are functioning. This is a percentage from 0 to 90, with anything 91 and over displayed by the NHS electronic health record system (Epic) as ">91" because an eGFR of 91 percent and above indicates healthy renal function.
- eGFR Date: The date the eGFR test was conducted.

The variables derived from these features were:

- Age\_Range: Categorical variable to store the age category of the patient based on their age to aid in visualisation. Possible values for this column are: "Young (1 to 25)", "Adult / Middle Aged (26-50)", "Older Adult / Old (51-75)" and "Elderly (76-100)"
- Has\_Hypoglycemia: Binary variable to store whether the patient has hypoglycemia. A glucose measurement of 4mmol/L and below means the patient is hypoglycemic and has 1 in this column, 0 otherwise.
- Glycemia\_Type: Categorical variable to store the type of glycemia based on the patient's glucose measurement. For the purposes of this project, the classes we have been instructed to use are (all units in mmol/L):
  - 1. "Severe Hypoglycemia" for blood glucose values 2.2 and below
  - $2.\,$  "Hypoglycemia"- for blood glucose values from 2.3 to 4 both inclusive
  - 3. "Target Range"- for blood glucose values from 4.1 to 11 both inclusive
  - 4. "Hyperglycemia" for blood glucose values above 11
- eGFR\_Category: Categorical variable that shows how serious the loss of kidney function is, based on the eGFR percentage. The possible values for this column are:
  - 1. "eGFR less than 20 Kidney Failure" for eGFR less than or equal to 20%
  - 2. "eGFR between 20 & 40 Critical Loss of Kidney Function"- for eGFR above 20% but less than or equal to 40%
  - 3. "eGFR between 40 & 60 Significant Loss of Kidney Function"- for eGFR above 40% but less than or equal to 60%
  - 4. "eGFR between 60 & 80 Moderate Loss of Kidney Function"- for eGFR above 60% but less than or equal to 80%
  - 5. "eGFR between 80 & 90 Minor Loss of Kidney Function"- for eGFR above 80% but less than or equal to 90%
  - 6. "eGFR above 90 Normal kidney function"- for eGFR above 90% (data has been processed to only include "91" for this class as healthy eGFR is 91% and above).
- Wider\_Ethnic\_Group: Categorical variable to store the overarching ethnic group based on the one specified in the ethnicity column, as that had a total of 57 unique values. Possible values are: "Unknown or Not Stated", "White", "Mixed", "Asian or Asian British", "Black or Black British" and "Other Ethnic Groups".

Note that columns obtained after cleaning the original data to extract a numerical value (such as blood glucose) have been omitted for brevity.

Please see Appendix A subsection A.1 for screenshots of the dataset(s).

## 4 Methodology and Implementation

It presents and justifies the methodology used to deal with the problem and describes in detail the implementation procedures. The background theory presented in the previous chapter can be recalled to support the proposed implementation. The originality, novelty and contribution are to be demonstrated with the discussion of the strengths and limitations.

The content of "Main results" is in "\contents\introduction.tex"

## 5 Main Results and Findings

The chapter reports the contributions of your work. For example, it could contain the following sub-sections to summarise the contribution of the project such as Theoretical Development, Analysis and Design, Implementation and Experimental Work, Results, Observation and Discussion.

### 5.1 Interpretation Analysis and Evaluation

It summarises the results obtained from the proposed design and methodology. The way to obtain the results should be described in detail. Analysis and evaluation have to be performed. Comparisons should be made. It should justifies if the project aims, objectives, requirements and specifications have been achieved.

## 6 Legal, Social, Ethical and Professional Issues

A chapter gives a reasoned discussion about legal, social ethical and professional issues within the context of your project problem. You should also demonstrate that you are aware of the Code of Conduct & Code of Good Practice issued by the British Computer

Society

(BSC)

(https://www.bcs.org/membership/become-a-member/bcs-code-of-conduct/) for computer science project and Rule of Conduct issued by The Institution of Engineering and Technology (IET)

(https://www.theiet.org/about/governance/rules-of-conduct/) for engineering project. You should have applied their principles, where appropriate, as you carried out your project. You could consider aspects like: the effects of your project on the public well-being, security, software trustworthiness and risks, Intellectual Property and related issues, etc.

#### 7 Others

This section is for demonstration of equations, figures, tables, which is not required for the report.

#### 7.1 Maths

$$\frac{\mathrm{d}S_t}{S_t} = r\mathrm{d}t + \sigma\mathrm{d}W_t, \qquad S_0 > 0, \tag{7.1}$$

The equation  $\sigma = ma$  follows easily [?].

## 7.2 Glossary and acronyms

Latexlinuxs and other Unix operating systems are better then Windows because they support lymformula out of the box [?].

A ref is missing here

### 7.3 Figures

Here is an example [?] of how to insert a picture:

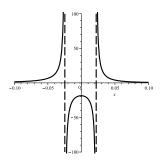


Figure 1: This is the caption for the figure.

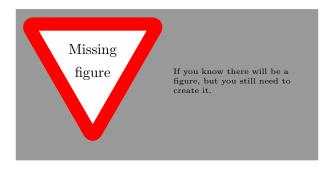


Figure 2: This is the caption for the figure which is not even present.

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or two side-by-side pictures:

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7.4 Table 11

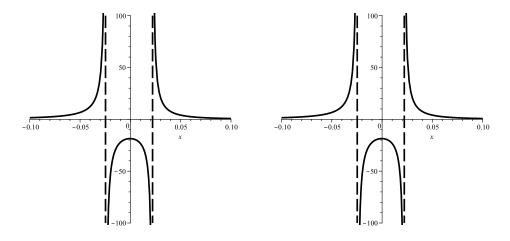


Figure 3: Another caption

#### 7.4 Table

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This needs further explanation

Something	Someother	Thing
Seems	to be	$\operatorname{good}$

Table 1: Random data for a table.

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## 8 More Others

## 8.1 What is calibration?

Here is an example of a matrix[?] in  $A \in \mathcal{M}_n(\mathbb{R})$ :

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \vdots \\ a_{n1} & \dots & \dots & a_{1n}. \end{pmatrix}$$

## 8.2 Numerical methods for calibration

...

## 9 Legal, Social, Ethical and Professional Issues

Research and projects within the medical domain are always inherently sensitive regardless of the kind of data involved or the presence of human participants. This sensitivity is amplified when a highly prominent industry stakeholder such as the NHS is interested, in view of the fact that it oversees public health across all of the UK. Right from the start, I have prioritized regular and transparent communication with our industry advisor through recurring meetings, while upholding implicit confidentiality agreements regarding the nature of the data and the project's specific objectives. All analysis was conducted within the agreed-upon scope. All deliverables were presented in a coherent and actionable format, thereby reflecting my commitment to their distinct requirements and towards fostering a trustworthy working relationship. Being cognizant of my social and ethical responsibility in this undertaking to advance public welfare, I have submitted an application in KCL's Research Ethics Management Application System (REMAS) which should supplement the agreements and principles established at the time of inception of the project, considering that the project is a KEP with industry (NHS England). According to KCL and REMAS guidelines, this project is classed as "Minimal Risk" [ref], in that it involves the study of pre-existing data that is not available to the general public, but is fully anonymous at the point which I as a researcher gain access to it. The industry advisor has kindly provided us the necessary data after complete anonymization, which removes any risk of personal identification. (Still submitted a Full Application Form instead of a minimal risk application) To further support this and in line with the guidelines listed in the General Data Protection Regulation (GDPR) as well as the Data Protection Act (DPA) 2018, the data was both shared with me and only accessed through secure organization / university credentials, meaning that it did not need to be fetched at all through any resource or API calls, eliminating the risk of interception. It was stored locally for on-machine data analysis and modelling through frequently used, open-source Python libraries, without the involvement of any online tools where the data has to be uploaded for research. Efforts have been taken to determine whether the project requires approval from any external entities, for example the Health Research Authority[ref] https://www.hra.nhs.uk/planning-and-improving-research/researchplanning/student-research/. This was found to be not necessary. No recruitment of human participants was in the picture. Every care was taken to prevent any conflicts of interest from occurring, whether around other similar research, intellectual property, project objectives or any other sectors. I have also considered reliability measures to minimize the possibility of any kind of "reverse engineering" that may be carried out on my work. This substantiates that I have displayed special adherence to the British Computer Society (BCS) Code of Conduct and Code of Practise[ref], especially the directives regarding "Public Interest" and "Professional Competence and Integrity".

The content of "Conclusion" is in "\contents\conclusion.tex"

## 10 Conclusion

It is a chapter to sum up the main points and findings of the work; how you achieve the project aims and address the research questions; the contributions and results you have achieved. Future plan and development can be mentioned in this section as well. It is normally in one or two pages.

References 15

## References

[1] Diabetes UK, "Cost of devastating complications highlights urgent need to transform diabetes care in the UK." URL. Accessed: 2025-07-07.

- [2] Diabetes Professional Care, "Almost 60costs are for 'preventable' complications." URL. Accessed: 2025-07-07.
- [3] Hospital Pharmacy Europe, "Data from health economic model shows cost of managing hypoglycaemia." URL. Accessed: 2025-07-07.
- [4] Diabetes Research and Wellness Foundation, "Emergency call-outs for diabetesrelated condition reduced following hypos education campaign." URL. Accessed: 2025-07-07.
- [5] NHS Digital, "National diabetes audit 2021-22, report 1: Care processes and treatment targets, detailed analysis report." National Diabetes Audit 2021-22. Accessed: 2025-07-07.

References 16

## A Appendix

## A.1 Dataset

ď
STH GYNAECOLOGY WARD Manual blood glucose: 12.00 mmol/L 52 Black or Black British - Unspecified
Manual blood glucose: 10.10 mmol/L 59 Black or Black British - African
STH WILLIAM GULL WARD Manual blood glucose: 10.40 mmol/L 88 White - Any other White background Choose not to disclose
GH RICHARD BRIGHT WARD Manual blood glucose: 5.70 mmol/L 64 Not stated/Undefined
Manual blood glucose: 11.90 mmol/L 76 White - British
Manual blood glucose: 7.40 mmol/L 74 Black or Black British - Caribbean
POCT Glucose Blood Manually En: 7. 52 White - English
Manual blood glucose: 4.80 mmol/L 74 White - English
Manual blood glucose: 7.20 mmol/L 81 Not stated/Undefined
Manual blood glucose: 6.50 mmol/L 78 Black or Black British - Caribbean
Manual blood glucose: 9.60 mmol/L 68 Black or Black British - African
Manual blood glucose: 5.70 mmol/L 34 Black or Black British - African
Manual blood glucose: 6.10 mmol/L
Manual blood glucose: 7.50 mmol/L
Manual blood glucose: 6.30 mmol/L
Manual blood glucose: 6.50 mmol/L 69 White - English
Manual blood glucose: 5.10 mmol/L 68 White - British
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Manual blood glucose: 11.10 mmol/L
Manual blood glucose: 13.00 mmol/L
Manual blood glucose: 9.20 mmol/L
Manual blood glucose: 3.40 mmol/L
Manual blood glucose: 4.30 mmol/L
Manual blood glucose: 6.60 mmol/L
POCT Glucose Blood Manually E: 11.
Manual blood glucose: 12.40 mmol/L
Manual blood glucose: 6.20 mmol/L
Manual blood glucose: 11.00 mmol/L 88 White - Any other White background Choose not to disclose
Manual blood glucose: 6.10 mmol/L 59 Black or Black British - African
Manual blood glucose: 5.80 mmol/L 35 Any Other Ethnic Group
Manual blood almost 6 80 mmol/l 00 Block or Block British Caribboan

Figure 4: Raw dataset

A.1 Dataset 18

Age Ethnicity	Gender Identity Las	: HbA1c La	st HBA1C Dt Las	t eGFR EGFRD	late Admit Weight	Last HbA1c Last HBA1C Dt Last eGFR EGFR Date Admit Weight Glucose Value Length of Stay (Ti Age_Range	Age_Range	Has_Hypos Glycemia_Type	eGFR_Category Wider_Ethi	Wider_Ethnic_Group
56				78 05,06,2025	3025	10.1 0 days 03:00:00	10.1 0 days 03:00:00 Older Adult / Old (51-75)	0 Target Range	eGFR between 60 & 80 - Moderate Loss of Kidney F Unknown or Not Stated	or Not Stated
59 Black or Black British - African	Female	43	20,01,2024	58 01,07,2	58 01,07,2025 121 kg	5.2 157 days 18:00:00	5.2 157 days 18:00:00 Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	ack British
88 White - Any other White background Choose not to disclose	Choose not to disclose	23	30,04,2024	59 18,06,2025	3025	10.4 24 days 02:00:00 Elderly (76-100)	Elderly (76-100)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney White	
64 Not stated/Undefined	Male	82	19,12,2024	23 27,06,2025	2025	5.7 1 days 09:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 20 & 40 - Critical Loss of Kidney Fun Unknown or Not Stated	or Not Stated
76 White - British				61 27,06,2025	3025	11.9 42 days 01:00:00 Elderly (76-100)	Elderly (76-100)	0 Hyperglycemia	eGFR between 60 & 80 - Moderate Loss of Kidney F White	
74 Black or Black British - Caribbean		25	22,04,2024	50 30,06,2	30,06,2025 95 kg	7.4 45 days 09:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	ack British
52 White - English			06×		27,06,2025 55 kg	7.5 30 days 18:00:00	7.5 30 days 18:00:00 Older Adult / Old (51-75)	0 Target Range	eGFR above 90 - Normal kidney function White	
74 White - English	Male			32 24,06,2025	3025	4.8 3 days 03:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 20 & 40 - Critical Loss of Kidney Fun White	
81 Not stated/Undefined		61	13,11,2023	82 20,06,2025	3025	7.2 16 days 00:00:00	Elderly (76-100)	0 Target Range	eGFR between 80 & 90 - Minor Loss of Kidney Func Unknown or Not Stated	or Not Stated
78 Black or Black British - Caribbean	Female	39	24,04,2025	5 17,06,2025	3025	6.5 5 days 16:00:00	Elderly (76-100)	0 Target Range	eGFR less than 20 - Kidney Failure Black or Black British	ack British
68 Black or Black British - African	Male	47	03,05,2024	58 06,06,2025	3025	9.6 0 days 16:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	ack British
34 Black or Black British - African	Male		06<	25,06,2025	3025	5.7 8 days 17:00:00	Adult / Middle Aged (26-50)	0 Target Range	eGFR above 90 - Normal kidney function Black or Black British	ack British
72 Black or Black British - Caribbean	Male	43	17,04,2024	60 06,06,2025	3025	6.1 1 days 10:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	ack British
52 Black or Black British - Unspecified	Female			77 06,06,2025	3025	7.5 1 days 09:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 60 & 80 - Moderate Loss of Kidney FBlack or Black British	ack British
72 Black or Black British - Caribbean	Male	43	17,04,2024	60 06,06,2025	2025	6.3 1 days 10:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	ack British
69 White - English			>90	05,06,2025	2025	6.5 1 days 01:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR above 90 - Normal kidney function	
68 White - British		47	12,04,2024 >90		30,06,2025 75 kg	5.1 70 days 09:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR above 90 - Normal kidney function	
50 Black or Black British - Any other Black Female	k Female	109	10,06,2025 >90	10,06,2025	3025	10.8 4 days 13:00:00	Adult / Middle Aged (26-50)	0 Target Range	eGFR above 90 - Normal kidney function Black or Black British	ack British
74 Black or Black British - Caribbean		25	22,04,2024	50 30,06,2	30,06,2025 95 kg	6 45 days 09:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	lack British
65 White - English	Male			70 30,06,2025	2025	11.1 6 days 21:00:00	Older Adult / Old (51-75)	0 Hyperglycemia	eGFR between 60 & 80 - Moderate Loss of Kidney f White	
59 White - Any other White background	Male				26,06,2025 62.3 kg	13 61 days 17:00:00	13 61 days 17:00:00 Older Adult / Old (51-75)	0 Hyperglycemia	eGFR between 60 & 80 - Moderate Loss of Kidney F White	
41 Black or Black British - Caribbean				63 06,06,2025	3025	9.2 0 days 14:00:00	Adult / Middle Aged (26-50)	0 Target Range	eGFR between 60 & 80 - Moderate Loss of Kidney FBlack or Black British	ack British
35 Any Other Ethnic Group	Male				29,06,2025 77 kg	3.4 77 days 17:00:00	3.4 77 days 17:00:00 Adult / Middle Aged (26-50)	1 Hypoglycemia	eGFR between 80 & 90 - Minor Loss of Kidney Fund Other Ethnic Groups	nic Groups
35 Any Other Ethnic Group	Male			81 29,06,2	29,06,2025 77 kg	4.3 77 days 17:00:00	Adult / Middle Aged (26-50)	0 Target Range	eGFR between 80 & 90 - Minor Loss of Kidney Fund Other Ethnic Groups	nic Groups
53 Black or Black British - Caribbean	Female	40	08,12,2023 >90	05,06,2025	2025	6.6 1 days 13:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR above 90 - Normal kidney function Black or Black British	lack British
68 Black or Black British - African	Male	47	03,05,2024	58 06,06,2025	3025	0 days 16:00:00	Older Adult / Old (51-75)		eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	lack British
68 Black or Black British - African	Male	47	03,05,2024	58 06,06,2025	3025	11.4 0 days 16:00:00	Older Adult / Old (51-75)	0 Hyperglycemia	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	ack British
69 White - English			>90	05,06,2025	2025	12.4 1 days 01:00:00	Older Adult / Old (51-75)	0 Hyperglycemia	eGFR above 90 - Normal kidney function White	
82 Not stated/Undefined				53 01,07,2	01,07,2025 67.2 kg	6.2 40 days 19:00:00 Elderly (76-100)	Elderly (76-100)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Unknown or Not Stated	or Not Stated
88 White - Any other White background Choose not to disclose	Choose not to disclose	23	30,04,2024	59 18,06,2025	2025	11 24 days 02:00:00 Elderly (76-100)	Elderly (76-100)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney White	
59 Black or Black British - African	Female	43	20,01,2024	58 01,07,2	58 01,07,2025 121 kg	6.1 157 days 18:00:00	6.1 157 days 18:00:00 Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Black or Black British	lack British
72 Asian or Asian British - Arab	Male	8	13,05,2024	59 26,06,2025	2025	9.2 4 days 09:00:00	Older Adult / Old (51-75)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney Asian or Asian British	sian British
35 Any Other Ethnic Group	Male			81 29,06,2	29,06,2025 77 kg	5.8 77 days 17:00:00	5.8 77 days 17:00:00 Adult / Middle Aged (26-50)	0 Target Range	eGFR between 80 & 90 - Minor Loss of Kidney Func Other Ethnic Groups	nic Groups
99 Black or Black British - Caribbean	Female			26 16,06,2025	2025		Elderly (76-100)	0 Target Range	eGFR between 20 & 40 - Critical Loss of Kidney Fun Black or Black British	lack British
76 White - British				61 27,06,2025	2025	7.3 42 days 01:00:00	Elderly (76-100)	0 Target Range	eGFR between 60 & 80 - Moderate Loss of Kidney FWhite	
88 White - British	Female	61	22,01,2024	49 10,06,2025	2025	4.8 6 days 21:00:00	Elderly (76-100)	0 Target Range	eGFR between 40 & 60 - Significant Loss of Kidney White	
71 Milita Deltich			OBS	107 06 2025	3005	6.2 3 days 08:00:00	Older Adult / Old (51-75)	0 Target Range	Mhita Mormal kidney function	

Figure 5: Dataset with cleaned features (this is in addition to the fields of the raw dataset)