

Article

Determining risk of diabetes and cardiovascular diseases occurrence utilizing data via smart devices

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1. Introduction

The introduction should [1], [2], [3], [4], [5], [6], [7], [8], [9], [10], [11], [12], [13], [14], [15], [16] briefly place the study in a broad context and highlight why it is important. It should define the purpose of the work and its significance. The current state of the research field should be reviewed carefully and key publications cited. Please highlight controversial and diverging hypotheses when necessary. Finally, briefly mention the main aim of the work and highlight the principal conclusions. As far as possible, please keep the introduction comprehensible to scientists outside your particular field of research. Citing a journal paper. Now citing a book reference or other reference types. Please use the command for the following MDPI journals, which use author–date citation: Administrative Sciences, Arts, Econometrics, Economies, Genealogy, Humanities, IJFS, Journal of Intelligence, Journalism and Media, JRFM, Languages, Laws, Religions, Risks, Social Sciences, Literature.

2. Related Work

As mentioned before, the applications of Statistical Analysis and Machine Learning in healthcare and more specifically in diabetes condition have demonstrated a steady rise in the last two decades, since the development of corresponding programming frameworks

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have enabled the easy storage, collection, processing, analysis of the massively available data quantity and employment of statistical and Machine Learning models [17–19]. Regarding diabetes research field, the literature deals with the identification of diabetic people, early or long term (2-10 years) prognosis and diabetes complications prediction or identification. Our literature review is focused on relatively new research articles or systematic reviews which are related with the context of our article e.g prediction of diabetes mellitus or prediabetes utilizing demographic, anthropometric, biometric, laboratory, nutritional, medical history, etc. data as input features. The first mathematical approaches over diabetes issue consisted of statistical risk scores exploiting questionnaires filled by waves from the participants. Some of the famed risk scores are Leicester Risk Assessment Score[20] developed by Leicester University and FINDRISC [21] developed by University of Helsinki. The former utilizing a Logistic Regression model, take into account age, ethnicity, sex, first degree family history of diabetes, antihypertensive therapy or history of hypertension, waist circumference and BMI to predict current impaired glucose regulation or diabetes mellitus, achieving an AUC metric of 72% and the latter -also exploited Logistic Regression-uses gender, age, BMI, use of blood pressure medication, history of high blood glucose, physical activity, daily consumption of vegetables, fruits or berries and family history of diabetes to predict a 10-year development achieving an AUC metric of 86%. There are also numerous researches that deal with deep learning and more specifically with image recognition for the classification of diabetic retinopathy, which is a typical complication and very well studied in the research field, using images from eye bulb as input [2,12]. We can observe at a first glance two variances of diabetes studies. The Leicester Risk aims to identify the current health condition, while FINDRISC tries to predict a long term prevalence. Another diabetes complications studies utilizing Machine Learning and Deep Learning include neuropathy and nephropathy [2,12]. Apart from classification problems there are also regression methods which are exploited for the prediction of Fasting Plasma Glucose or HbA1c levels, such biomarkers that are the best indicators of abnormal glucose regulation and consequently diabetes mellitus presence [2,3,12]. Delving into our main purpose whi

3. Materials and Methods

Materials and Methods should be described with sufficient details to allow others to replicate and build on published results. Please note that publication of your manuscript implicates that you must make all materials, data, computer code, and protocols associated with the publication available to readers. Please disclose at the submission stage any restrictions on the availability of materials or information. New methods and protocols should be described in detail while well-established methods can be briefly described and appropriately cited.

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This is an example of a quote.

4. Results

This section may be divided by subheadings. It should provide a concise and precise description of the experimental results, their interpretation as well as the experimental conclusions that can be drawn.

4.1. Subsection

4.1.1. Subsubsection

Bulleted lists look like this:

- First bullet;
- Second bullet;
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Numbered lists can be added as follows:

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All figures and tables should be cited in the main text as Figure 1, Table 1, Table 2, etc.



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Title 1	Title 2	Title 3
Entry 1	Data	Data
Entry 2	Data	Data

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Title 1	Title 2	Title 3	Title 4
Entry 1	Data	Data	Data
Entry 2	Data	Data	Data ¹

¹ This is a table footnote.

Text.
Text.

4.3. Formatting of Mathematical Components

This is the example 1 of equation:

$$a = 1, \quad (1)$$

the text following an equation need not be a new paragraph. Please punctuate equations as regular text.

This is the example 2 of equation:

$$a = b + c + d + e + f + g + h + i + j + k + l + m + n + o + p + q + r + s + t + u + v + w + x + y + z \quad (2)$$



Figure 2. This is a wide figure.

Please punctuate equations as regular text. Theorem-type environments (including propositions, lemmas, corollaries etc.) can be formatted as follows:

Theorem 1. *Example text of a theorem.*

The text continues here. Proofs must be formatted as follows:

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7. Patents

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Abbreviations

The following abbreviations are used in this manuscript:

MDPI	Multidisciplinary Digital Publishing Institute
DOAJ	Directory of open access journals
TLA	Three letter acronym
LD	Linear dichroism

Appendix A

Appendix A.1

The appendix is an optional section that can contain details and data supplemental to the main text—for example, explanations of experimental details that would disrupt the flow of the main text but nonetheless remain crucial to understanding and reproducing the research shown; figures of replicates for experiments of which representative data are shown in the main text can be added here if brief, or as Supplementary Data. Mathematical proofs of results not central to the paper can be added as an appendix.

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Entry 1	Data	Data
Entry 2	Data	Data

Appendix B

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