

Table SI. [Models and model settings]

In model settings (*hyperparameters*), numbers and names in parentheses in bold are the optimal parameters chosen. Parameters without values are the default for that model. For the random forests model, the max depths were 18 when training on all data, 12 for demography only, and 10 for age and sex only.

Model	Description	Key Model Settings	References for Implementation
Logistic Regression	Logistic regression is a linear model widely used for binary classification tasks. It estimates the probability that a given input belongs to a particular class by applying a logistic function to a linear combination of input features.	C (regularization strength (0.01)), Penalty (L1 or L2), Solver (e.g., ' liblinear ', 'saga')	Scikit-learn: LogisticRegression
Random Forests	Random forests are ensemble models that create multiple decision trees during training and output the mode of the classes for classification tasks. This method is robust to overfitting and can handle high-dimensional data well.	Number of Trees (n_estimators (1000)), Max Depth (max_depth (18, 12, 10)), Min Samples Split (min_samples_split), Min Samples Leaf (min_samples_leaf)	Scikit-learn: RandomForestClassifier
Gradient Boosting Machines (GBM)	Gradient boosting machines build models sequentially by training each new model to correct the errors made by the previous ones. This method is powerful and often achieves high predictive performance, especially on structured data.	Number of Trees (n_estimators (1500)), Learning Rate (0.1), Max Depth (max_depth (1)), Subsample	Scikit-learn: GradientBoostingClassifier
Multi Layer Perceptron (MLP)	The multi-layer perceptron (MLP) is a type of feedforward neural network	Number of Hidden Units (1), Number of nodes (16), Activation Function	Scikit-learn: MLPClassifier

	capable of learning non-linear relationships in the data. In our study, we used an MLP with a single hidden layer to capture complex patterns that simpler models might miss.	(e.g., ' relu ', 'tanh'), Learning Rate (0.005), Alpha (L2 Penalty), Epochs (300), Batch size (1000)	
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Table SII. [TRIPOD+AI guidelines]



Version: 11-January-2024

Section/Topic	Item	Development / evaluation ¹	Checklist item	Reported on page
TITLE				
<i>Title</i>	1	D;E	Identify the study as developing or evaluating the performance of a multivariable prediction model, the target population, and the outcome to be predicted	Title page
ABSTRACT				
<i>Abstract</i>	2	D;E	See TRIPOD+AI for Abstracts checklist	Page 2
INTRODUCTION				
<i>Background</i>	3a	D;E	Explain the healthcare context (including whether diagnostic or prognostic) and rationale for developing or evaluating the prediction model, including references to existing models	Page 3, 4
	3b	D;E	Describe the target population and the intended purpose of the prediction model in the context of the care pathway, including its intended users (e.g., healthcare professionals, patients, public)	Page 3, 4
	3c	D;E	Describe any known health inequalities between sociodemographic groups	Page 7, 10
<i>Objectives</i>	4	D;E	Specify the study objectives, including whether the study describes the development or validation of a prediction model (or both)	Page 4
METHODS				
<i>Data</i>	5a	D;E	Describe the sources of data separately for the development and evaluation datasets (e.g., randomised trial, cohort, routine care or registry data), the rationale for using these data, and representativeness of the data	Page 4, 5
	5b	D;E	Specify the dates of the collected participant data, including start and end of participant accrual; and, if applicable, end of follow-up	Page 4, 5
<i>Participants</i>	6a	D;E	Specify key elements of the study setting (e.g., primary care, secondary care, general population) including the number and location of centres	Page 4, 5
	6b	D;E	Describe the eligibility criteria for study participants	Page 4-6
	6c	D;E	Give details of any treatments received, and how they were handled during model development or evaluation, if relevant	N / A
<i>Data preparation</i>	7	D;E	Describe any data pre-processing and quality checking, including whether this was similar across relevant sociodemographic groups	Page 5
<i>Outcome</i>	8a	D;E	Clearly define the outcome that is being predicted and the time horizon, including how and when assessed, the rationale for choosing this outcome, and whether the method of outcome assessment is consistent across sociodemographic groups	Page 5
	8b	D;E	If outcome assessment requires subjective interpretation, describe the qualifications and demographic characteristics of the outcome assessors	N / A
	8c	D;E	Report any actions to blind assessment of the outcome to be predicted	N / A
<i>Predictors</i>	9a	D	Describe the choice of initial predictors (e.g., literature, previous models, all available predictors) and any pre-selection of predictors before model building	Page 5, 8
	9b	D;E	Clearly define all predictors, including how and when they were measured (and any actions to blind assessment of predictors for the outcome and other predictors)	Page 4, 5
	9c	D;E	If predictor measurement requires subjective interpretation, describe the qualifications and demographic characteristics of the predictor assessors	N / A
<i>Sample size</i>	10	D;E	Explain how the study size was arrived at (separately for development and evaluation), and justify that the study size was sufficient to answer the research question. Include details of any sample size calculation	Page 4, 5, 7
<i>Missing data</i>	11	D;E	Describe how missing data were handled. Provide reasons for omitting any data	Page 5, 6
<i>Analytical methods</i>	12a	D	Describe how the data were used (e.g., for development and evaluation of model performance) in the analysis, including whether the data were partitioned, considering any sample size requirements	Page 6
	12b	D	Depending on the type of model, describe how predictors were handled in the analyses (functional form, rescaling, transformation, or any standardisation)	Page 5, 6
	12c	D	Specify the type of model, rationale ² , all model-building steps, including any hyperparameter tuning, and method for internal validation	Page 6
	12d	D;E	Describe if and how any heterogeneity in estimates of model parameter values and model performance was handled and quantified across clusters (e.g., hospitals, countries). See TRIPOD-Cluster for additional considerations ³	N / A
	12e	D;E	Specify all measures and plots used (and their rationale) to evaluate model performance (e.g., discrimination, calibration, clinical utility) and, if relevant, to compare multiple models	Page 7, 14
	12f	E	Describe any model updating (e.g., recalibration) arising from the model evaluation, either overall or for particular sociodemographic groups or settings	N / A
	12g	E	For model evaluation, describe how the model predictions were calculated (e.g., formula, code, object, application programming interface)	Page 7
<i>Class imbalance</i>	13	D;E	If class imbalance methods were used, state why and how this was done, and any subsequent methods to recalibrate the model or the model predictions	Page 6
<i>Fairness</i>	14	D;E	Describe any approaches that were used to address model fairness and their rationale	Page 7
<i>Model output</i>	15	D	Specify the output of the prediction model (e.g., probabilities, classification). Provide details and rationale for any classification and how the thresholds were identified	Page 7

¹ D=items relevant only to the development of a prediction model; E=items relating solely to the evaluation of a prediction model; D;E=items applicable to both the development and evaluation of a prediction model

² Separately for all model building approaches.

³ TRIPOD-Cluster is a checklist of reporting recommendations for studies developing or validating models that explicitly account for clustering or explore heterogeneity in model performance (eg, at different hospitals or centres). Debray et al, BMJ 2023; 380: e071018 [DOI: 10.1136/bmj-2022-071018]

Training versus evaluation	16	D;E	Identify any differences between the development and evaluation data in healthcare setting, eligibility criteria, outcome, and predictors	Page 10,11
Ethical approval	17	D;E	Name the institutional research board or ethics committee that approved the study and describe the participant-informed consent or the ethics committee waiver of informed consent	Page 11
OPEN SCIENCE				
Funding	18a	D;E	Give the source of funding and the role of the funders for the present study	Page 11
Conflicts of interest	18b	D;E	Declare any conflicts of interest and financial disclosures for all authors	Page 11
Protocol	18c	D;E	Indicate where the study protocol can be accessed or state that a protocol was not prepared	N / A
Registration	18d	D;E	Provide registration information for the study, including register name and registration number, or state that the study was not registered	Page 11
Data sharing	18e	D;E	Provide details of the availability of the study data	N / A
Code sharing	18f	D;E	Provide details of the availability of the analytical code ⁴	N / A
PATIENT & PUBLIC INVOLVEMENT				
Patient & Public Involvement	19	D;E	Provide details of any patient and public involvement during the design, conduct, reporting, interpretation, or dissemination of the study or state no involvement.	N / A
RESULTS				
Participants	20a	D;E	Describe the flow of participants through the study, including the number of participants with and without the outcome and, if applicable, a summary of the follow-up time. A diagram may be helpful.	Page 4-5, 8
	20b	D;E	Report the characteristics overall and, where applicable, for each data source or setting, including the key dates, key predictors (including demographics), treatments received, sample size, number of outcome events, follow-up time, and amount of missing data. A table may be helpful. Report any differences across key demographic groups.	Page 5, 8
	20c	E	For model evaluation, show a comparison with the development data of the distribution of important predictors (demographics, predictors, and outcome).	Page 8
Model development	21	D;E	Specify the number of participants and outcome events in each analysis (e.g., for model development, hyperparameter tuning, model evaluation)	Page 4, 6, 8
Model specification	22	D	Provide details of the full prediction model (e.g., formula, code, object, application programming interface) to allow predictions in new individuals and to enable third-party evaluation and implementation, including any restrictions to access or re-use (e.g., freely available, proprietary) ⁵	N / A
Model performance	23a	D;E	Report model performance estimates with confidence intervals, including for any key subgroups (e.g., sociodemographic). Consider plots to aid presentation.	Page 8
	23b	D;E	If examined, report results of any heterogeneity in model performance across clusters. See TRIPOD Cluster for additional details ³ .	Page 8
Model updating	24	E	Report the results from any model updating, including the updated model and subsequent performance	N / A
DISCUSSION				
Interpretation	25	D;E	Give an overall interpretation of the main results, including issues of fairness in the context of the objectives and previous studies	Page 9-10
Limitations	26	D;E	Discuss any limitations of the study (such as a non-representative sample, sample size, overfitting, missing data) and their effects on any biases, statistical uncertainty, and generalizability	Page 10
Usability of the model in the context of current care	27a	D	Describe how poor quality or unavailable input data (e.g., predictor values) should be assessed and handled when implementing the prediction model	Page 5, 6
	27b	D	Specify whether users will be required to interact in the handling of the input data or use of the model, and what level of expertise is required of users	N / A
	27c	D;E	Discuss any next steps for future research, with a specific view to applicability and generalizability of the model	Page 9-10

From: Collins GS, Moons KGM, Dhiman P, et al. *BMJ* 2024;385:e078378. doi:10.1136/bmj-2023-078378

⁴ This relates to the analysis code, for example, any data cleaning, feature engineering, model building, evaluation.

⁵ This relates to the code to implement the model to get estimates of risk for a new individual.