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Last updated by author(s): Jun 21, 2024

Reporting Summary

Nature Portfolio wishes to improve the reproducibility of the work that we publish. This form provides structure for consistency and transparency in reporting. For further information on Nature Portfolio policies, see our <u>Editorial Policies</u> and the <u>Editorial Policy Checklist</u>.

Statistics

For all statistical analyses, confirm that the following items are present in the figure legend, table legend, main text, or Methods section.

n/a	Cor	nfirmed
	\boxtimes	The exact sample size (n) for each experimental group/condition, given as a discrete number and unit of measurement
	\boxtimes	A statement on whether measurements were taken from distinct samples or whether the same sample was measured repeatedly
	\boxtimes	The statistical test(s) used AND whether they are one- or two-sided Only common tests should be described solely by name; describe more complex techniques in the Methods section.
	\boxtimes	A description of all covariates tested
	\boxtimes	A description of any assumptions or corrections, such as tests of normality and adjustment for multiple comparisons
	\boxtimes	A full description of the statistical parameters including central tendency (e.g. means) or other basic estimates (e.g. regression coefficient) AND variation (e.g. standard deviation) or associated estimates of uncertainty (e.g. confidence intervals)
	\boxtimes	For null hypothesis testing, the test statistic (e.g. <i>F</i> , <i>t</i> , <i>r</i>) with confidence intervals, effect sizes, degrees of freedom and <i>P</i> value noted Give <i>P</i> values as exact values whenever suitable.
\boxtimes		For Bayesian analysis, information on the choice of priors and Markov chain Monte Carlo settings
\boxtimes		For hierarchical and complex designs, identification of the appropriate level for tests and full reporting of outcomes
	\boxtimes	Estimates of effect sizes (e.g. Cohen's d, Pearson's r), indicating how they were calculated
		Our web collection on <u>statistics for biologists</u> contains articles on many of the points above.

Software and code

Policy information about availability of computer code

Data collectionWe used REDCap (version 11.1.3) to generate the questionnaire for clinicians. All clinicians reviewed MRIs using 3D Slicer (version 4.10.2) and
logged their findings in REDCap (version 11.1.3).Data analysisOur software development utilized Python (version 3.11.7) and the models were developed using PyTorch (version 2.1.0). We used several
other Python libraries to support data analysis, including pandas (version 1.5.3), scipy (version 1.10.1), tensorboardX (version 2.6.2),
torchvision (version 0.15) and scikit-learn (version 1.2.2). Figures were prepared using Canva and Adobe Illustrator. Python scripts as well as
help files along with information on the study population are made available on GitHub https://github.com/vkola-lab/nmed2024.

For manuscripts utilizing custom algorithms or software that are central to the research but not yet described in published literature, software must be made available to editors and reviewers. We strongly encourage code deposition in a community repository (e.g. GitHub). See the Nature Portfolio guidelines for submitting code & software for further information.

Data

Policy information about availability of data

All manuscripts must include a data availability statement. This statement should provide the following information, where applicable:

- Accession codes, unique identifiers, or web links for publicly available datasets
- A description of any restrictions on data availability
- For clinical datasets or third party data, please ensure that the statement adheres to our policy

This study includes data from the following nine cohorts: National Alzheimer's Coordinating Center (NACC), Alzheimer's Disease Neuroimaging Initiative (ADNI), Frontotemporal Lobar Degeneration Neuroimaging Initiative (NIFD), Parkinson's Progression Marker Initiative (PPMI), Australian Imaging, Biomarker and Lifestyle Flagship Study of Ageing (AIBL), Open Access Series of Imaging Studies (OASIS), 4 Repeat Tauopathy Neuroimaging Initiative (4RTNI), Lewy Body Dementia Center for Excellence at Stanford University (LBDSU), and Framingham Heart Study (FHS). Data from ADNI, AIBL, NIFD, PPMI and 4RTNI can be downloaded from the LONI website at https://ida.loni.usc.edu. NACC and OASIS data can be downloaded at https://naccdata.org and https://sites.wustl.edu/oasisbrains/, respectively. Finally, data from FHS https://www.framinghamheartstudy.org and LBDSU https://med.stanford.edu/poston-lab/LBD.html can be obtained upon request, subject to institutional approval. We used the Montreal Neuroimaging Institute MNI152 template for image processing purposes, and the template can be downloaded at http://www.bic.mni.mcgill.ca/ServicesAtlases/ICBM152NLin2009.

Human research participants

Policy information about studies involving human research participants and Sex and Gender in Research.

Reporting on sex and gender	The data was obtained from existing cohorts. We reported sex-related information on all the cohorts whenever available. We also performed sex-specific subgroup analysis.
Population characteristics	Participants in the study were 51,269 individuals with a mean age of 73. They included 22,349 males and 28,920 females. The ethnic composition was 40,335 White, 5,840 Black or African American, 1,285 Asian, 276 American Indian or Alaskan Native, 38 Native Hawaiian or Pacific Islander and 1,430 Multiracial. All participants were screened for cognitive impairment, with 19,849 classified as having normal cognition, 9,357 as having mild cognitive impairment (MCI) and 22,063 as having dementia. More details on the diagnostic information can be found in the Methods section.
Recruitment	We did not recruit any participants for this study.
Ethics oversight	The data collection for the Framingham Heart Study and the Lewy Body Dementia Center for Excellence at Stanford University was approved by the respective institutional review boards.

Note that full information on the approval of the study protocol must also be provided in the manuscript.

Field-specific reporting

Please select the one below that is the best fit for your research. If you are not sure, read the appropriate sections before making your selection.

Life sciences

Behavioural & social sciences

Ecological, evolutionary & environmental sciences

For a reference copy of the document with all sections, see nature.com/documents/nr-reporting-summary-flat.pdf

Life sciences study design

All studies must disclose on these points even when the disclosure is negative.

Sample size	We obtained data from all the nine cohorts. No sample size calculation was carried out. We considered all subjects satisfying the inclusion and exclusion criterion described in the manuscript. We included 38,319 participants for training and 12,950 participants for testing the model. More details on the study population can be found in the manuscript.
Data exclusions	We excluded data from the cohorts if the diagnosis information (normal cognition (NC), mild cognitive impairment (MCI), or dementia (DE)) was not available.
Replication	The results can be replicated by following the methods described in the manuscript or by running the code available in our GitHub repository. The data from ADNI, NACC, AIBL, NIFD, PPMI, OASIS, and 4RTNI are open access. Additional data requests are required to access the data from FHS and LBDSU for replicating the findings from our study.
Randomization	When building the deep learning model, the cases were shuffled using a consistent random seed and were split into train, validation and testing sets using stratified sampling at person level.
Blinding	In the comparison of clinicians versus deep learning model performance, clinicians were blinded to the documented clinical diagnoses of the cases presented.

Reporting for specific materials, systems and methods

Methods

 \square

 \boxtimes

n/a Involved in the study

Flow cytometry

MRI-based neuroimaging

ChIP-seq

We require information from authors about some types of materials, experimental systems and methods used in many studies. Here, indicate whether each material, system or method listed is relevant to your study. If you are not sure if a list item applies to your research, read the appropriate section before selecting a response.

Materials & experimental systems

n/a	Involved in the study
\boxtimes	Antibodies
\boxtimes	Eukaryotic cell lines
\boxtimes	Palaeontology and archaeology
\boxtimes	Animals and other organisms
\boxtimes	Clinical data
\boxtimes	Dual use research of concern

Magnetic resonance imaging

Experimental design

Design type	Resting state structural MRI
Design specifications	We used multiple scan sequences including T1-weighted, T2-weighted, FLAIR and SWI scans whenever available. Detailed descriptions of the scan protocols and design specifications can be obtained from the respective websites of the study cohorts.
Behavioral performance measures	Not applicable
Acquisition	
Imaging type(s)	Structural
Field strength	1.5 or 3 Tesla
Sequence & imaging parameters	T1-weighted, T2-weighted, FLAIR, SWI sequences
Area of acquisition	Whole brain
Diffusion MRI Used	Not used
Preprocessing	
Preprocessing software d T ir	he collected imaging data were stored in the NIFTI file format, categorized by participant and the date of their visit. The MRI cans underwent a singular pre-processing step, which involved skull stripping using SynthStrip, a computational tool esigned for extracting brain voxels from various image types. MRI scans were linearly registered based on the MNI152 atlas o ensure the purity of the dataset, we excluded calibration, localizer, and 2D scans from the downloaded data before itiating model training.
Normalization	ee previous response
Normalization template	INI152
Noise and artifact removal	II MRI scans were normalized to the range [0,1] to increase the homogeneity of the data.
Volume censoring	o volume censoring was used in this study.

Statistical modeling & inference

Model type and settings	Our model employs the transformer architecture to process diverse diagnostic data, including demographics, medical history, neuroimaging, functional assessments, and neuropsychological test scores. Each data type is first transformed into a fixed-length vector using a tailored approach, creating the initial input layer for the transformer. The transformer then synthesizes these vector inputs, interpreting and converting them into a coherent series of diagnostic predictions, effectively leveraging the complex interplay of varied health-related parameters.
Effect(s) tested	Task- and stimulus-related effects were not tested in this study.

Specify type of analysis: 🛛 Whole brain 🔄 ROI-based 🔄 Both				
Statistic type for inference (See <u>Eklund et al. 2016</u>)	We used Shapley analysis to perform feature importance analysis.			
Correction	We applied the Kruskal-Wallis H-test for independent samples and subsequently conducted post-hoc Dunn's testing with Bonferroni correction to evaluate the relationship between clinical dementia rating scores and the model-predicted probabilities.			
Models & analysis				
Eunctional and/or effective connectivity				
Graph analysis				
Multivariate modeling or predictive analysis				

Multivariate modeling and predictive analysis (We summarized our model results using area under receiver operating characteristic curves (AUROC) and precision-recall curves (AUPR). Also, model accuracy, sensitivity, specificity, F1-score and Matthew's correlation coefficient values were reported.