nature portfolio

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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 Editorial Policies and the Editorial Policy Checklist.

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FUI	all statistical analyses, commit that the following items are present in the figure regend, table regend, main text, or Methods section.
n/a	Confirmed
	The exact sample size (n) for each experimental group/condition, given as a discrete number and unit of measurement
	A statement on whether measurements were taken from distinct samples or whether the same sample was measured repeatedly
	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)
	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 <i>Give P values as exact values whenever suitable.</i>
\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 <i>d</i> , Pearson's <i>r</i>), indicating how they were calculated
	Our web collection on statistics for hiologists contains articles on many of the points above

Software and code

Policy information about availability of computer code

Data collection

No software was used for data collection as data was obtained from public datasets.

Data analysis

The code is made publicly available under https://github.com/albarqounilab/FedDis. The following packages are used: scikit_image==0.17.2 matplotlib==3.3.2 numpy = 1.19.2wandb==0.10.30 monai==0.6.0 plotly==5.1.0 PyYAML==5.4.1

skimage==0.0 torch==1.9.0 Python >= 3.6

scikit_learn==0.24.2

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

Most of the datasets used in this study are publicly available and can be downloaded after signing a Data Usage Agreement. The OASIS dataset is available at https://www.oasis-brains.org; the ADNI-P datasets are available at http://adni.loni.usc.edu/data-samples/access-data/; the MSLUB dataset was available at http://lit.fe.uni-lj.si/tools.php?lang=eng; the MSISBI dataset is available at https://smart-stats-tools.org/lesion-challenge-2015; the WMH dataset is available at https://wmh.isi.uu.nl; and the BRATS 2018 dataset is available at https://www.med.upenn.edu/sbia/brats2018/data.html. For the prospective cohort, KRI, MSKRI, and GBKRI, upon reasonable request and signing of data transfer agreements and pending approval by our IRB and data protection officer, data can be shared.

Field-specific reporting			
X Life sciences	ne below that is the best fit for your research. If you are not sure, read the appropriate sections before making your selection. Behavioural & social sciences		
Life scier	nces study design		
All studies must dis	close on these points even when the disclosure is negative.		
Sample size	All experimental design and data details are provided in the "Methods" section		
Data exclusions	All experimental design and data details are provided in the "Methods" section		
Replication	All experimental design and data details are provided in the "Methods" section		
Randomization	Randomization was not performed as this is not applicable to our Image analysis work where we don't study treatment effects.		
Blinding	Randomization was not performed as this is not applicable to our work as well		
We require information	g for specific materials, systems and methods on from authors about some types of materials, experimental systems and methods used in many studies. Here, indicate whether each material, red 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 & exp	perimental systems Methods		
n/a Involved in th			
Antibodies			
Eukaryotic			
Palaeontology and archaeology MRI-based neuroimaging Animals and other organisms			
Human research participants			
Clinical dat	a		
Dual use re	search of concern		

Human research participants

Policy information about studies involving human research participants

Population characteristics

All demographic details are provided in Figure 1

Recruitment

For KRI, MSKRI, and GBKRI, all patients were part of in-house observational cohorts, some of which were prospective (MSKRI; with patient consent), while the others were retrospective (without patient consent).

Ethics oversight

For all patients, our local IRB approved the use of imaging data for research purposes after anonymization. As several patients were part of retrospective cohorts without explicit patient consent, these data cannot be shared as mandated by our IRB. For the prospective cohort, upon reasonable request and signing of data transfer agreements and pending approval by our IRB and data protection officer, data can be shared.

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

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Experimental design		
Design type	Structural MRI (FLAIR)	
Design specifications	All experimental design and data details are provided in the "Methods" section	
Behavioral performance measur	es not applicable	
Acquisition		
Imaging type(s)	Structural MRI	
Field strength	ЗТ	
Sequence & imaging parameters	OASIS-3: Resolution: 5x1x1, TR/TE/T1: Diverse ADNI-S: Resolution: 5x0.9x0.9, TR/TE/T1: Diverse ADNI-P: Resolution: 5x0.9x0.9, TR/TE/T1: 9/0.09/2.5 KRI: Resolution: 1.5x0.9x0.9, TR/TE/T1: 10/0.14/2.75 MSLUB: Resolution: 0.8x0.47x0.47, TR/TE/T1: 5/0.392/1.8 MSISBI: Resolution: 2.2x0.82x0.82, TR/TE/T1: 11/0.068/2.8 MSKRI: Resolution: 1.5x0.9x0.9, TR/TE/T1: 10/0.14/2.75 GBKRI: Resolution: 1.5x0.9x0.9, TR/TE/T1: 5/0.395/1.8 BRATS: Resolution: 1x1x1, TR/TE/T1: Diverse WMH: Resolution: Diverse, TR/TE/T1: Diverse	
Area of acquisition	uisition whole-brain	
Diffusion MRI Used	Not used Not used	
Preprocessing		
Preprocessing software	All scans have been registered to the SRI24 atlas template space to ensure all data share the same volume size and orientation. Subsequently, the scans have been skull-stripped with ROBEX.	
Normalization	The brain MRI volumes were linearly normalized to the [0,1] range.	
Normalization template	Describe the template used for normalization/transformation, specifying subject space or group standardized space (e.g. original Talairach, MNI305, ICBM152) OR indicate that the data were not normalized.	
Noise and artifact removal	No.	
Volume censoring		
Statistical modeling & infere	ence	
Model type and settings	deep learning methods were for processing the imaging data	
Effect(s) tested	ed the ability to detect pathologies without providing annotations	
Specify type of analysis: Whole brain ROI-based Both		
Statistic type for inference (See <u>Eklund et al. 2016</u>)	pixel-wise reconctruction	
Correction	No correction was needed	
Models & analysis n/a Involved in the study		

Multivariate modeling and predictive analysis Deep auto-encoding models were used in this manuscript to learn the normative prior of healthy brain

images. This was used later on to reconstruct a pseudo healthy version of a given input data. The residual is then used to detect and segment anomalies (pathologies) in Brain MRI Imaging. DICE Score was used to evaluate the predicted segmented pathology.