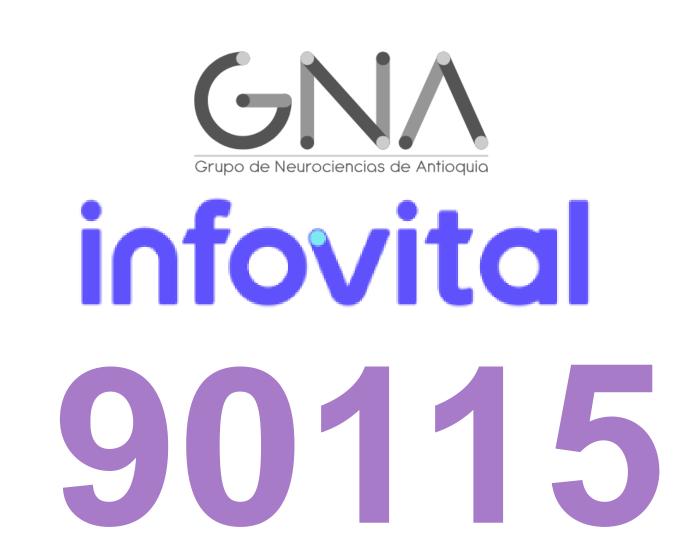
Improving diagnostic accuracy: development of an artificial intelligence model to aid in the diagnosis of Alzheimer's disease dementia.



Valeria Guerra Espinosa, MD¹, Juan Carlos Arbelaez Maestre, BA¹, Daniela Zabaleta, BA¹, Mateo Ruiz Espinosa, BA^{1,2}, Francisco Lopera, MD³, David Fernando Aguillon, MD PhD⁴ and Sofia Guerra Espinosa¹, (1) Infovital, Envigado, Colombia, (2) Universidad de Antioquia, Medellín, Colombia, (3) Grupo de Neurociencias de Antioquia, Facultad de Medicina, Universidad de Antioquia, Medellín, Colombia, (4) Grupo de Neurociencias, Universidad de Antioquia, Medellín, Colombia.

Early diagnosis of Alzheimer's disease is crucial to improve patient outcomes. Using the CRIPS-DM methodology, an IA model was developed that achieved high accuracy in diagnosing Alzheimer's. This model reduces the need for extensive neuropsychological testing, making it a costeffective tool for health professionals. However, it's currently adapted to the population of Medellin, Colombia, and requires further validation for wider use. This tool has the potential to significantly improve Alzheimer's diagnosis and support clinical decision making.

INTRODUCTION

Alzheimer's Disease Dementia (ADD)

is a neuropsychiatric disorder that

significantly and progressively impairs

cognitive domains, primarily affecting

memory, language and visuospatial

skills. ADD accounts for 60% to 70%

of dementia cases globally and its

prevalence in people older than 60

Although several organizations like

American

Aging

Association

clear and precise criteria for ADD,

there are several factors that hinder

their application such as the clinical

heterogeneity of ADD, its similarities

comorbidites, as well as the costs and

resources (neuroimaging and genetic

testing) needed to establish a

differential diagnosis, especially in

countries

and

Health Organization,

the

Psychiatric

and

neurodegenerative

the presence of

National

establish

such as

the

years old equals 4.02%.

Rationale

Association

Alzheimer's

Institute

World

other

diseases and

Background **CRISP-DM** methodology

1. Project understanding: Mining objective

METHODS

- Type of analysis
- Type of learning
- Analytical task
- 2. Data understanding:
- Data cycle
- Data dictionary
- Quality rules

3. Data preparation:

- Exploratory data analysis
- Outlier and null value cleaning
- Correlation analysis
- Class balancing
- **Transformations**

4. Modelling:

- Training with 70% of the data
- Models trained: decision trees, Knearest neighbors, neural networks, support vector machines, random forests,
- Hyperparameter selection: cross validation using genetic algorithms.

5. Evaluation

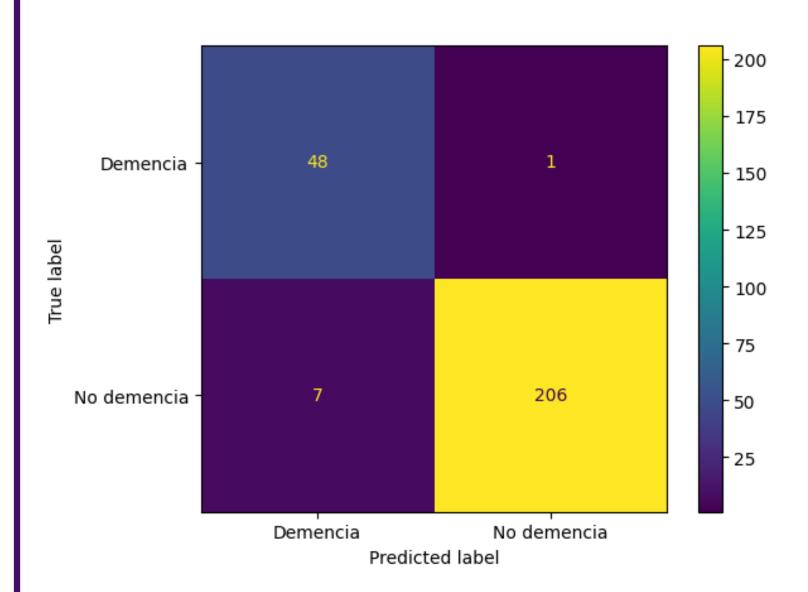
Metrics used: sensitivity, specificity, f1 score, AUC-ROC, balanced accuracy.

UI design to use the model in a real world setting.

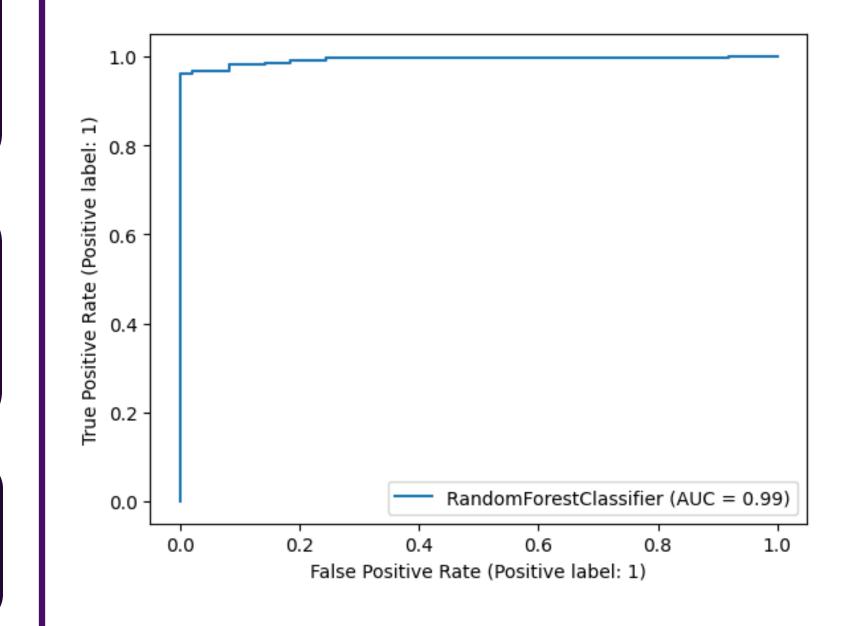
RESULTS

A total of 147 ML models were trained to classify people with and without ADD. Using a random forest (RF) algorithm and 33 features we achieved a sensitivity of 98%, a specificity of 100% and an area the receiver operating characteristic curve (AUC-ROC) of 0,988.

Confusion matrix of the model with the best performance



ROC of the model with the best performance



Discussion

CONCLUSIONS

We trained and evaluated a variety of ML models, highlighting the accuracy of a RF algorithm together with the use of GAs, to distinguish people with and without ADD. Among the 33 variables used are the results of the following neuropsychological tests: The Lawton Instrumental Activities of Daily Living Scale, Clinical Dementia Rating, Fast, Barthel, KATZ, MMSE, EDG, MLP, NPI, ROCF, TMT, FAS and Raven. We deployed the model with the best performance metrics through a graphical user interface in order to promote its future use in clinical settings. Although these results are promising, the ongoing need to evaluate and re-train the model, as well as the ethical considerations of its implementation.

Limitations

Since our model only takes into sociodemographic data, account medical records neuropsychological tests results, its predictions are only valid from a symptomatic standpoint and don't replace the need for neuroimaging or genetic testing in order to provide a differential diagnosis of Furthermore, our model is based towards the population included in this research: 870 patients from Medellin, Colombia. Therefore, due cultural and social, environmental factors over the clinical presentation of ADD, the model is not intended to be used in other countries or even in Colombian regions which vary significantly from Medellin.

REFERENCES

- Han, J., Pei, J., & Tong, H. (2023). Data mining: Concepts (Fourth edition).
- Fiest, K. M., Roberts, J. I., Maxwell, C. J., Hogan, D. B., Smith, E. E., Frolkis, A., Cohen, A., Kirk, A. Pearson, D., Pringsheim, T., Venegas-Torres, A., & Jetté, N. (2016). The Prevalence and Incidence of Dementia Due to Alzheimer's Disease: A Systematic Review and Meta-Analysis. Canadian Journal of Neurological Sciences / Journal Canadien Des Neurologiques, 43(S1), S51-S82. https://doi.org/10.1017/cjn.2016.36
- Sacuiu, S. F. (2016). Dementias. Handbook of Clinical Neurology (Vol. 138, pp. 123-151). Elsevier. https://doi.org/10.1016/B978-0-12-802973-2.00008-
- SPSS Inc., NCR Corporation, & Daimler AG. (2000). 1.0: Step-by-step data mining
- Tay, L. X., Ong, S. C., Tay, L. J., Ng, T., & Parumasivam, T. (2024). Economic Burden of Alzheimer's Disease: A Systematic Review. Value in Regional Issues, https://doi.org/10.1016/j.vhri.2023.09.008
- World Health Organization. (2004). ICD-10: International statistical classification of diseases and related health problems: Tenth revision. ICD-10.

https://apps.who.int/iris/handle/10665/42980

CONTACT



Association International Conference

The views and opinions expressed do not necessarily reflect the Alzheimer's Association.

Objective

Colombia.

developing

Develop a computerized diagnostic tool based on Artificial Intelligence (AI) to strengthen the capacity of healthcare professionals during the diagnosis of ADD.

- using random stratified sampling
- XGBoost and stacking.

6. Deployment