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USE OF METAHEURISTCS FOR MACHINE LEARNING IN THE CONTEXT OF ALS

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- Strong patient heterogeneity leads to unreliable prognosis.
- Machine learning methods based on large patients datasets have successfully identified underlying correlations in patient data as in the Pro-Act challenge [1].
- Limited performance of machine learning models because of too many irrelevant features leads to poorly usable results to predict disease progression.



Method	RMSE	R²-adjusted	Nb features
No selection	3.64	77.84	131
Metaheuristic	3.61	78.29	83

- Feature selection significantly reduces the number of features required to design the model while increasing its predictive efficiency.
- Overall our model is close to 3.61 on average compared to the actual value.
- The selected features explain 78% of the variation



Method

- The data consist of 2983 patients from 2 different datasets: Clinical trials from Pro-Act [1] and Exhonit [2].
- The data available across the datasets are age, sex, site of onset, ALSFRS, weight, height, time since first symptom, etc. both at T0 (first visit) and at T3 (next visit 3 months later) with a total of 131 features to forecast the evolution of the disease over the next few months (patients' ALSFRS)
- An autoregressive model allows each forecast of ALSFRS in a patient to feed the value back into the model to forecast its new value for a patient 3 months later.
- Feature selection consists in choosing a part of the features considered as the most relevant in the construction of the model to improve the

in ALSFRS among patients.

Figure 2. Comparison of forecast with reality in 3 patients



performance of the learning algorithm used.

Figure 1. Illustration of feature selection



- An estimate of the evolution of ALSFRS in a patient is plotted against the patient's characteristics.
- The model works well when a patient has slow or moderate disease progression. However, the model fails to forecast a recovery or a very strong progression.

- Implementation: Clinicians enter patient stats to obtain a forecast of disease progression using a web interface.
- Additional data will provide a more relevant model.
- The model can be used regardless of where the patient is in the progression of the disease.

References

- The evaluation of all possible combinations leads to a combinatorial explosion due to the overly high number of features.
- Metaheuristics are computational methods for solving complex optimisation problems and finding a solution close to the optimal solution.
- An improved version of differential evolution [3] was developed to find the best possible feature subset.
- [1] N et al Atassi. The pro-act database: design, initial analyses, and predictive features. *Neurology*, 83.19:1719–1725, 2014.
- [2] V et al Meininger. Pentoxifylline in als: a double-blind, randomized, multicenter, placebo-controlled trial. *Neurology*, 66.1:88–92, 2006.
- [3] T et al. Anani. Experimental comparison of metaheuristics for feature selection in machine learning in the medical context. pages 194–205, 2022.
- [4] Ali Furkan Kalay. Uq beamerposter template. Overleaf Poster Templates, 2021.