

# Tabular Data: Deep Learning is Not All You Need

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## Tabular Data

- Images or text are not the most frequent data
- Tabular data – a mix of numeric, symbolic, and textual data
- Challenges of tabular data:
  - Mixed features
  - Sparse data
  - Less data
  - Different datasets types
  - Prior knowledge (feature engineering)

## Compare Different Models

- Deep models
  - TabNet<sup>1</sup>
  - NODE<sup>2</sup>
  - DNF-Net<sup>3</sup>
  - 1D-CNN
- XGBoost<sup>4</sup>
- Ensemble of models

## Datasets and optimization

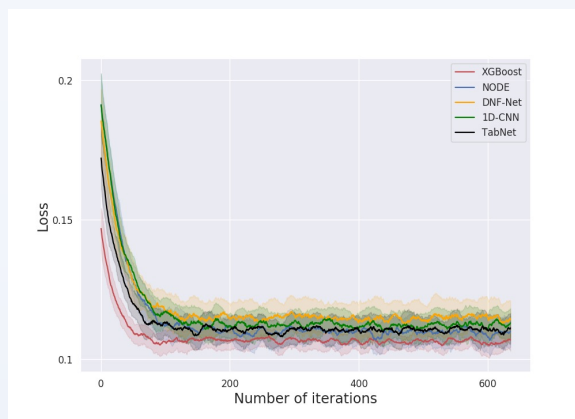
- No common benchmark
- Three datasets from each paper
- Two unrelated datasets
- Bayesian Hyper-parameter search
- 1000 – 10000 runs

## Results

Name	Average Relative Performance (%)
XGBoost	3.34
NODE	14.21
DNF-Net	11.96
TabNet	10.51
1D-CNN	7.56
Simple Ensemble	3.15
Deep Ensemble w/o XGBoost	6.91
Deep Ensemble w XGBoost	<b>2.32</b>

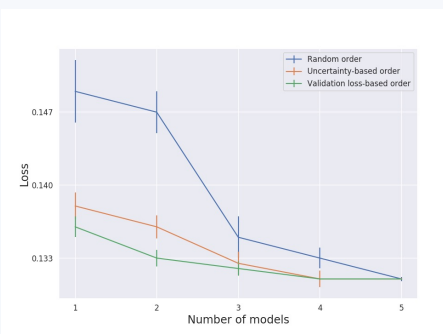
XGBoost had a much better performance than the deep models, their ensemble performed slightly better (lower is better)

## Hyper – Parameter Optimization



It is easier to optimize XGBoost

## Selecting Subset of Models in the Ensemble



Getting good results with only 3 models

## Summary

- On datasets that did not appear in their original papers, deep models were weaker
- XGBoost had better accuracy than the deep models
- Ensemble of deep models with XGBoost performed better
- XGBoost converged more quickly to good performance
- In an ensemble, the order of selecting models was important

## References

1. Arik, S. O., & Pfister, T. (2021). TabNet: Attentive Interpretable Tabular Learning. *Proceedings of the AAAI Conference on Artificial Intelligence*, 35(8), 6679-6687
2. Popov, Sergei, Stanislav Morozov, and Artem Babenko. "Neural oblivious decision ensembles for deep learning on tabular data." *ICLR, 2019*
3. Abutbul, Ami, et al. "DNF-Net: A Neural Architecture for Tabular Data." *ICLR, 2021*
4. Chen, Tianqi, and Carlos Guestrin. "Xgboost: A scalable tree boosting system." *Proceedings of the 22nd acm sigkdd international conference on knowledge discovery and data mining*. 2016.