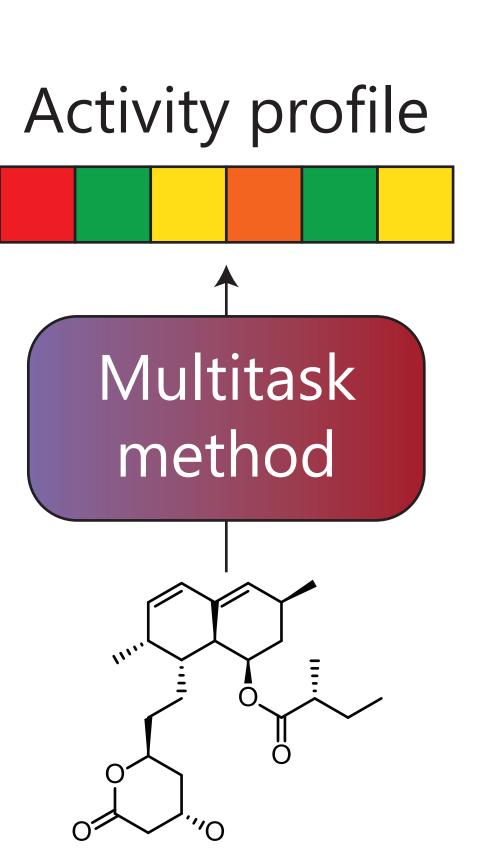


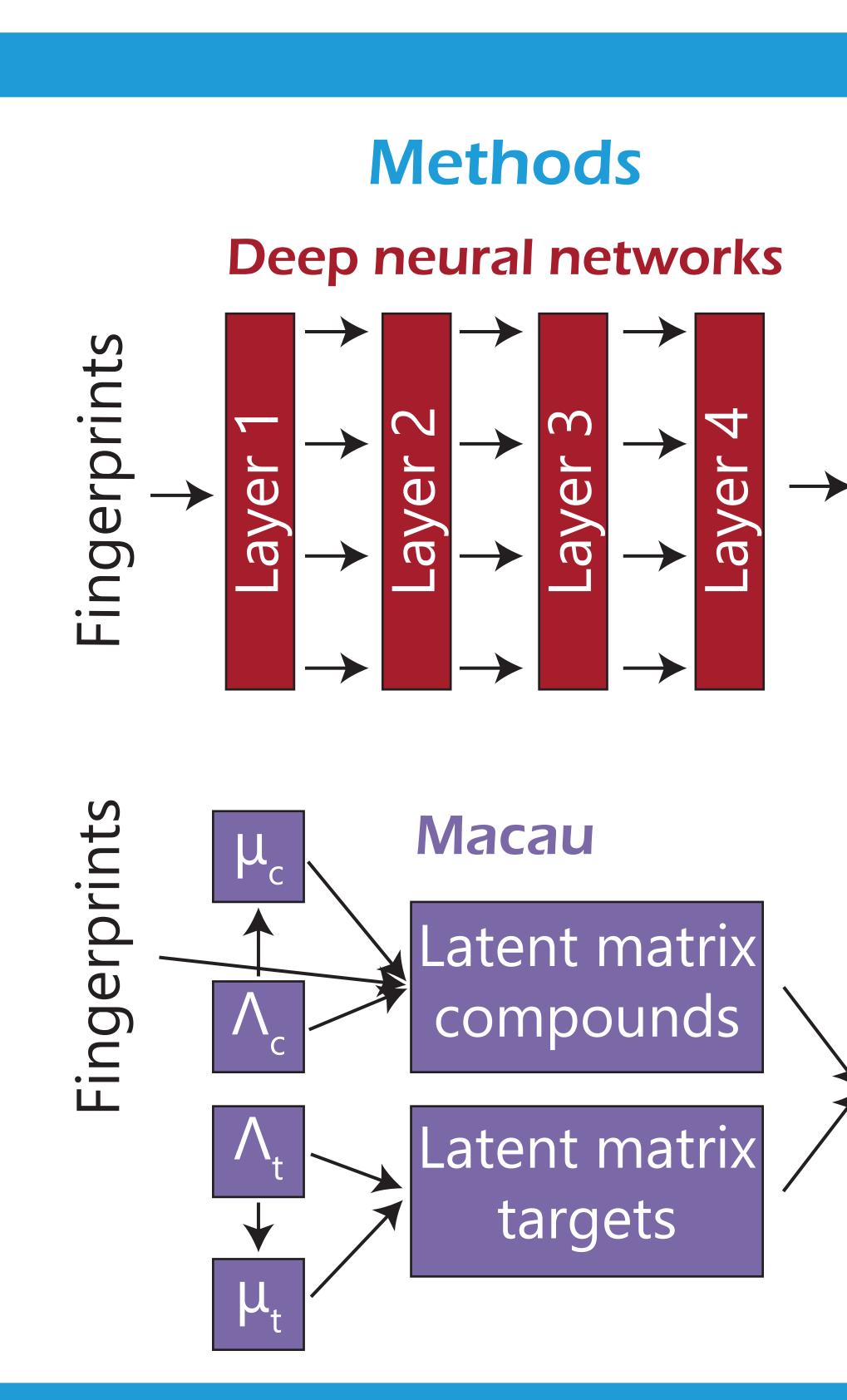


Introduction

Multitask machine learning techniques are able to predict several outputs with a single model. These techniques have become popular in the chemoinformatics field, thanks to deep neural networks (DNNs).¹ However, DNNs are not the only multitask method available. A novel method based on matrix factorization called Macau has also been applied to chemoinformatics problems.² In this work, we test these two methods on different multitarget datasets.

Sheffield.

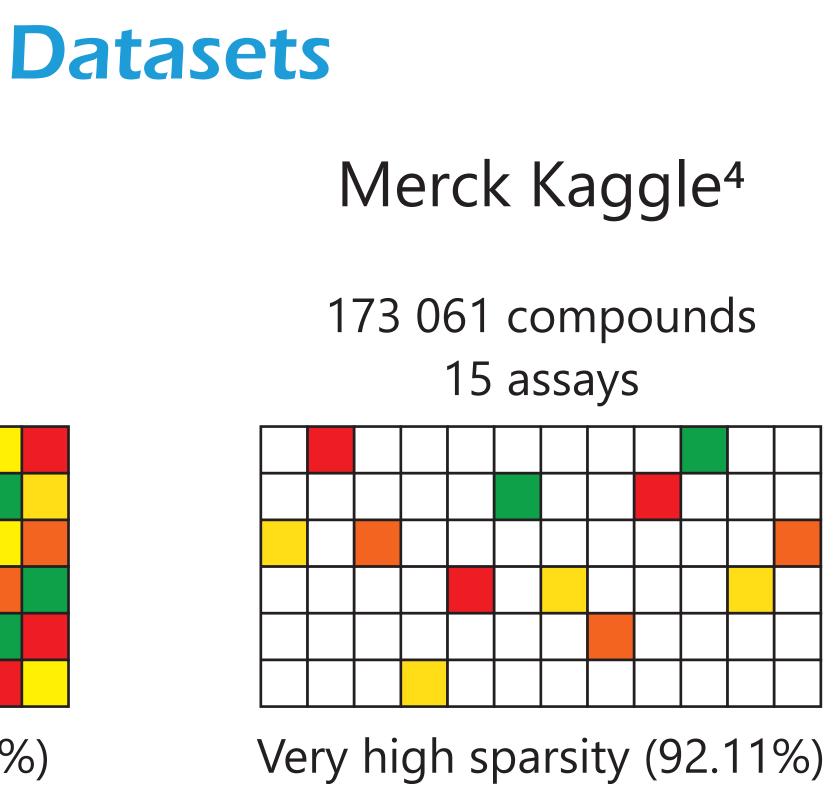




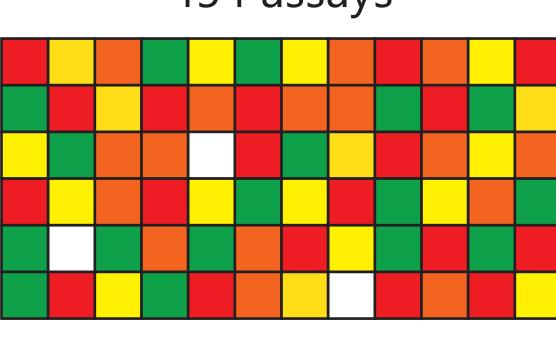
The University Comparison of multitask prediction methods for chemical data

Antonio de la Vega de León, Valerie J. Gillet

University of Sheffield, Regent Court, 211 Portobello, S1 4DP Sheffield, United Kingdom

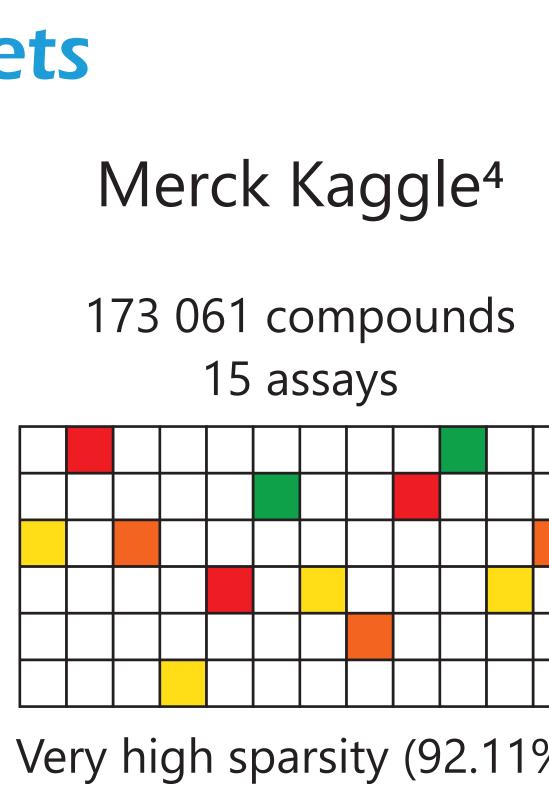


PKIS³ 367 compounds 454 assays

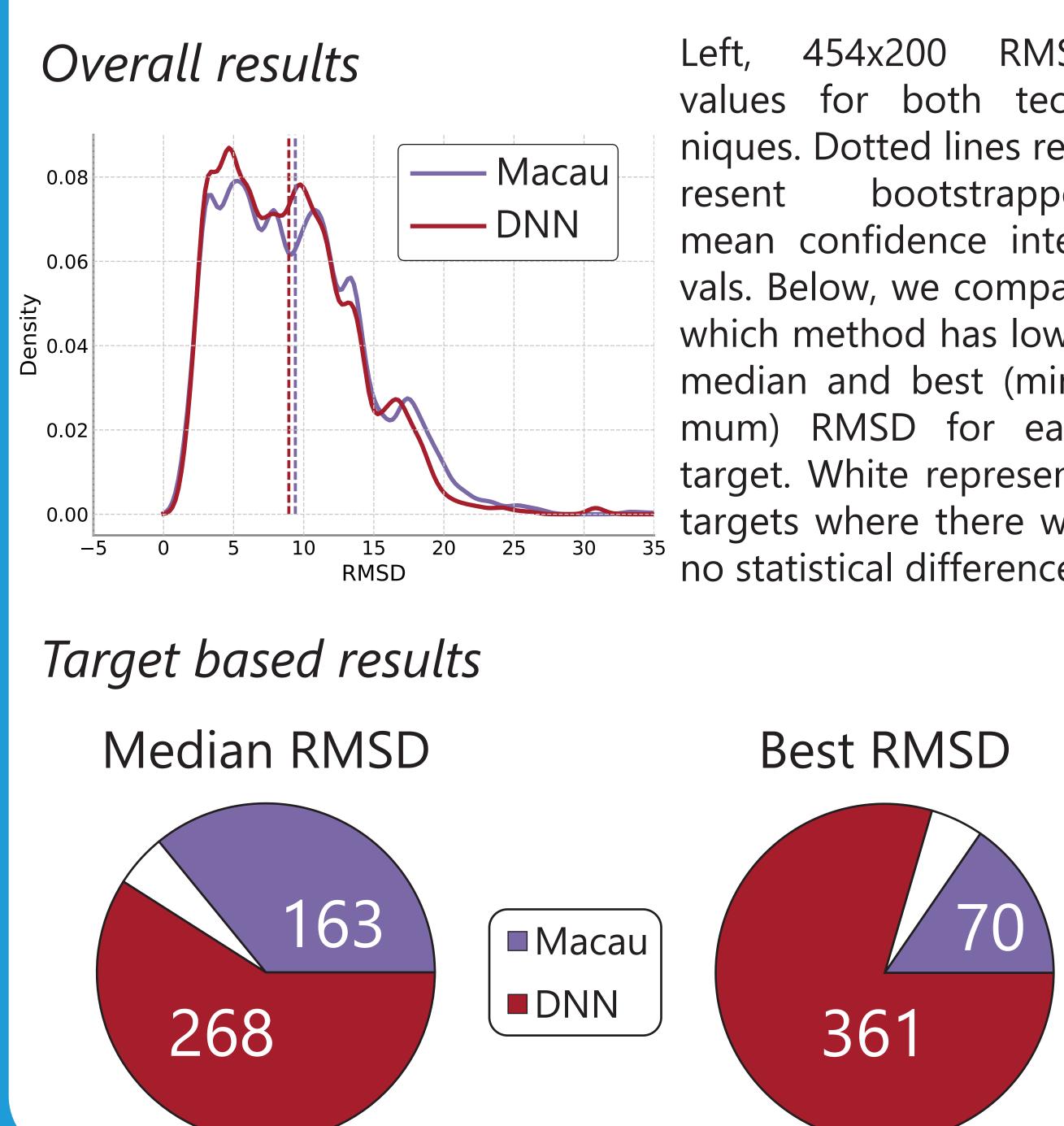


Very low sparsity (0.05%)

200 models trained Performance as RMSD



PKIS Results





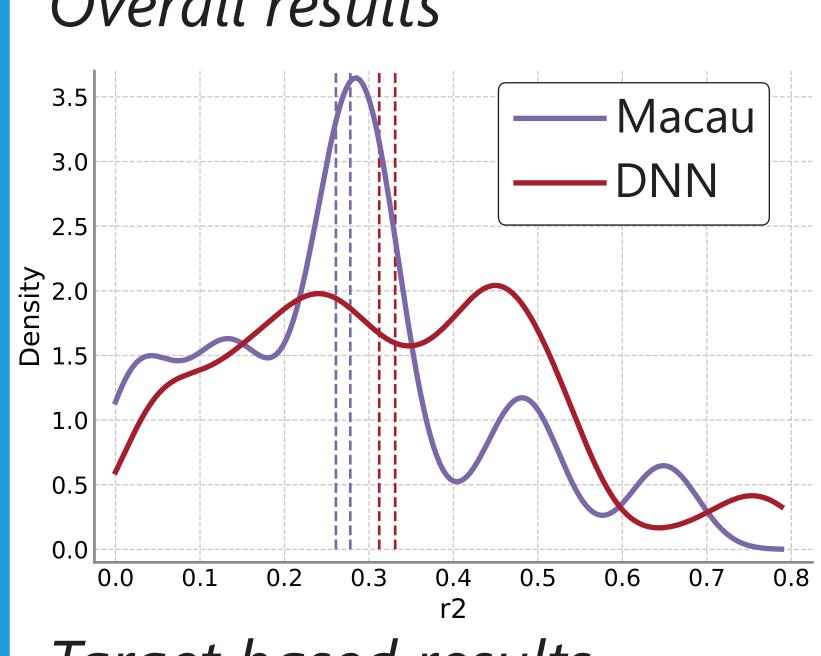


100 models trained Performance as r²

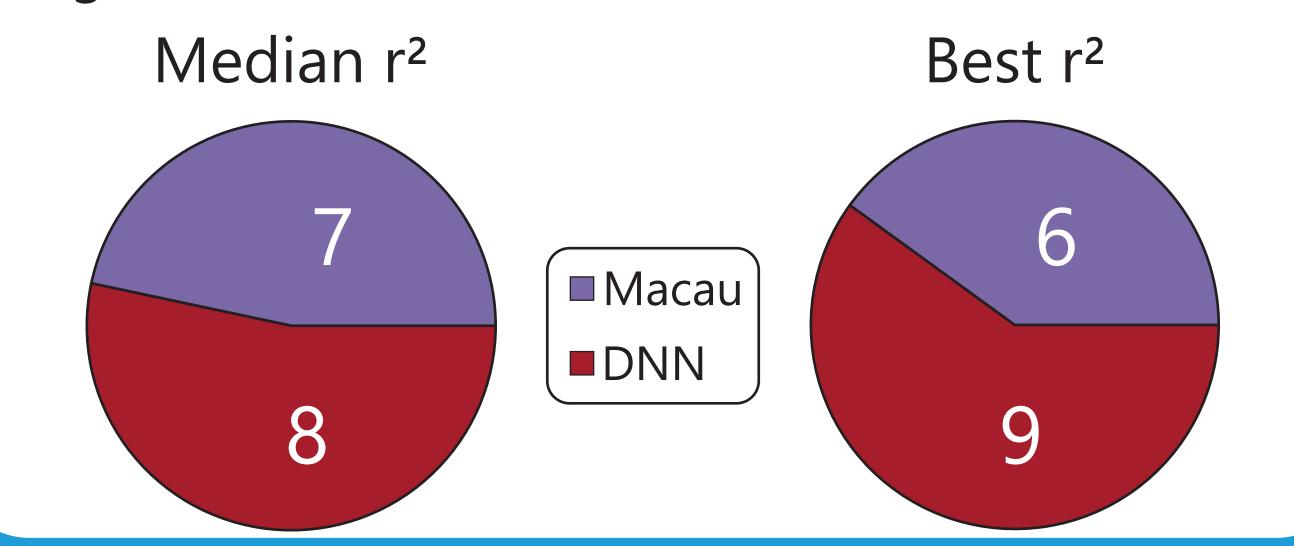
RMSD values for both techniques. Dotted lines repbootstrapped mean confidence intervals. Below, we compare which method has lower median and best (minimum) RMSD for each target. White represents targets where there was no statistical difference.

Merck Kaggle Results

Overall results



Target based results Median r²



Multitask prediction methods are very promising in chemoinformatics. Although the spotlight has been focused on deep neural networks, other methods such as Macau provide comparable performance. In addition, Macau does not require GPU resources to train models in adequate time.

1) LeCun et al. Nature 521, 436-444, 2015 2) Simm et al. arXiv 1509.04610, 2015

The research leading to these results has received funding from the European Union's Seventh Framework Programme (FP7/2007-2013) under grant agreement n°612347.

Left, 15x100 r² values for both techniques. Dotted bootlines represent confistrapped mean dence intervals. Below, which compare we higher method has median and best (maximum) r² for each target.

Conclusions

References

3) https://www.ebi.ac.uk/chembldb/extra/PKIS 4) Ma et al. JCIM 55, 263-274, 2015

Funding



