

Imputation of protein activity data using deep learning

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Neural network algorithm to



Merge simulations, physical laws, and experimental data

Reduce the need for expensive experimental development

Accelerate drugs and materials discovery

Generic with experimentally proven applications in materials discovery

A black box





Train on complete data





Digitize handwriting





Train on fragmented data





Predict on fragmented data

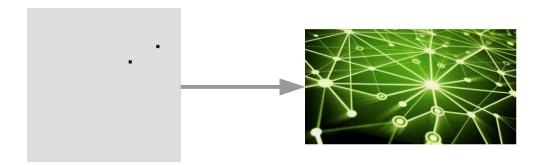




Enhancing drug discovery



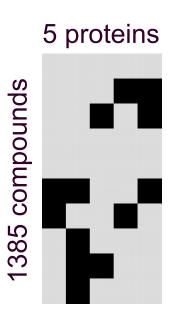
ChEMBL dataset just 0.1% complete



Data available for Adrenergic receptors



5 proteins with 1731 compounds, dataset 37% complete

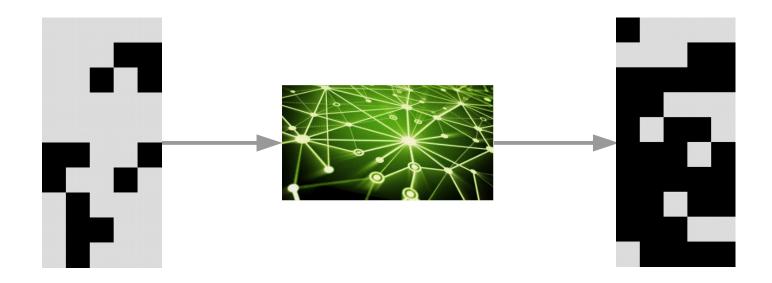




Impute the Adrenergic receptors dataset



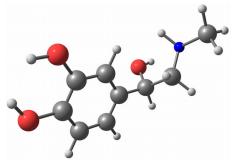
Original dataset 37% complete, filled 65% of entries

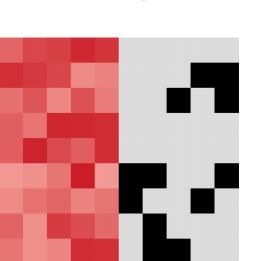




Additional descriptors for Adrenergic receptors







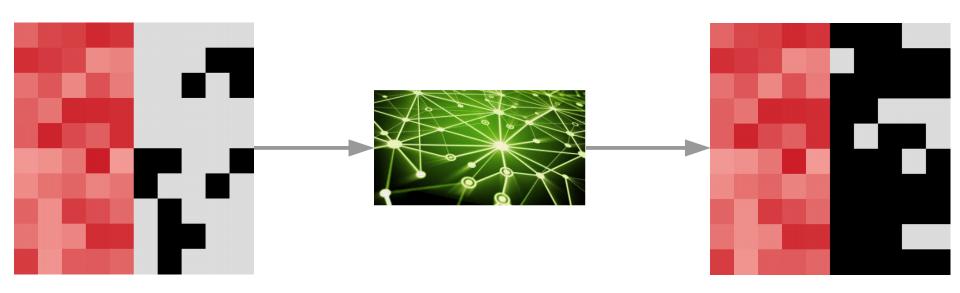




Improved predictions



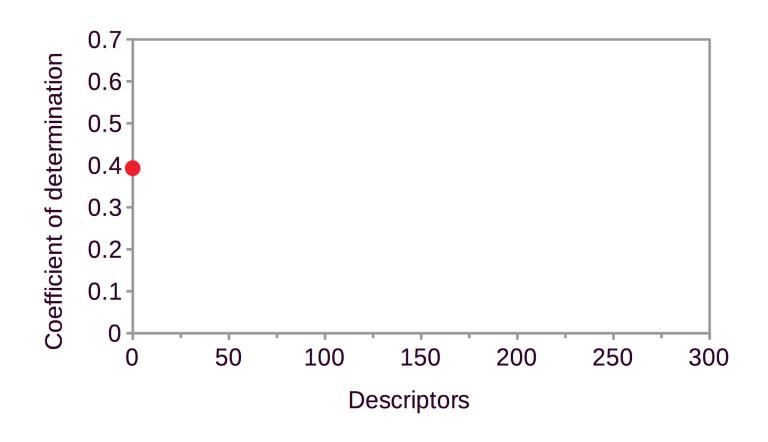
Include structural information to fill to 82%





Predictions from just activities

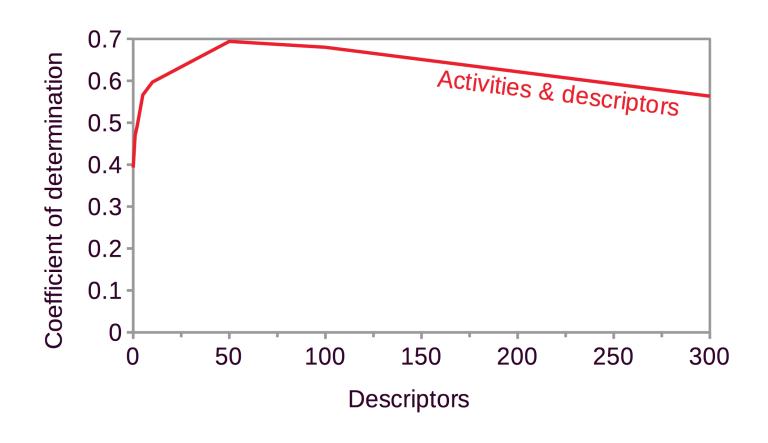






Predictions from activities and descriptors

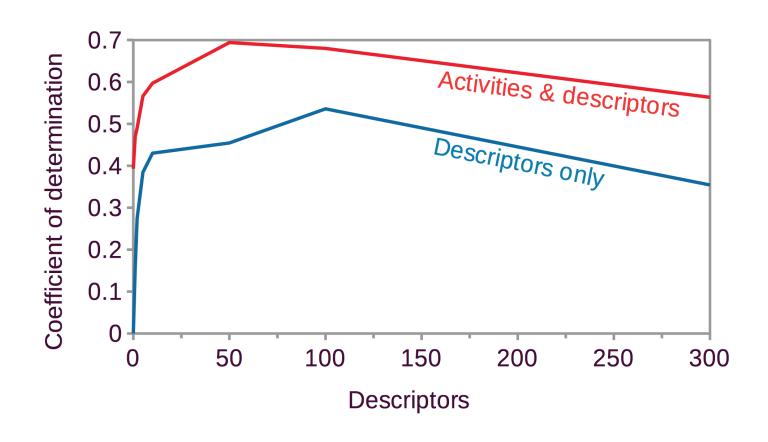






Predictions with just descriptors

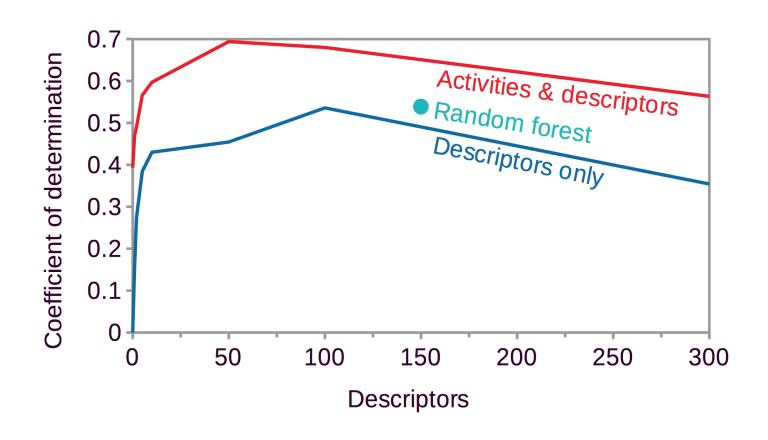






Comparison to random forest







Future prospects



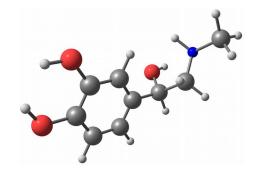
Enhance ChEMBL dataset from 0.1% to 20% complete



Materials designed



Drug discovery



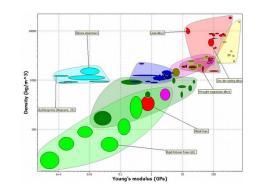
e-therapeutics

3D printed alloy designed from 10 data entries





Found errors in materials databases





Even more materials designed



Battery design with DFT and experimental data



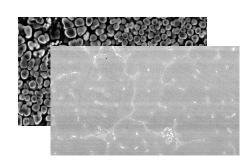


Designing lubricants with DFT and experimental data





Nickel and molybdenum alloys





Summary



Apply deep learning to high-value fragmented data

Merge experiments and simulations into holistic design tool

Experimentally **proven** applications in materials design, founded start-up **intellegens**

Scientists establish all possible **SOURCES** of information