

Imputation of protein activity data using deep learning

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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



5 proteins with 1731 compounds, dataset 37% complete







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









Enhance ChEMBL dataset from 0.1% to 20% complete



Materials designed



Drug discovery

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







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 **SOUICES** of information