

2 Higher costs, longer timelines, and lower success rates

1 Research productivity in pharmaceutical industry **decreased** in recent decades



3 Rise of de novo drug design



6 Proposes new approach for optimizing for chemical properties collectively using **Transfer Learning**



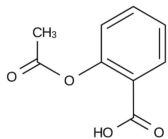
5 Optimizing for **specific chemical properties** is hard

4 Proposes new approach for generating **new chemical compounds** using **Recurrent Neural Networks**

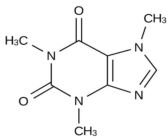
Introduction

Multiobjective de novo drug design with recurrent neural networks and nondominated sorting.

7 Molecules are represented using the **SMILES** string notation



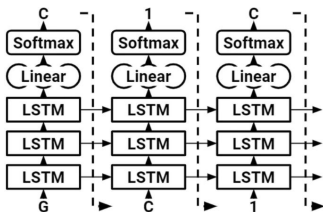
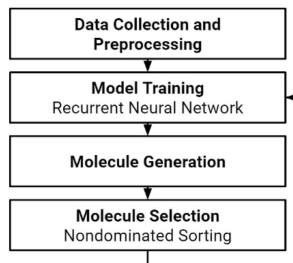
Acetylsalicylic Acid (Aspirin)
CC(=O)Oc1ccccc1C(=O)O



Epinephrine
CNC[C@H](O)c1ccc(O)c(O)c1

9 Generates molecules and then selects the **best ones based on nondominated sorting** then train with the newly generated

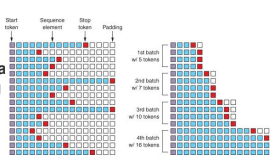
10 This process **simulates the traditional design-synthesis-test cycle** far more rapidly.



8 The network used composed of **three stacked LSTM layers**, each of size 1024, regularized with a 0.2 dropout ratio.
21 million trainable parameters.
Sequence length = 75 time steps

19 Further exploration of **ML in drug discovery** offers enormous **potential** to reduce the **cost and time** needed for the development of drugs.

18 Additional data preprocessing techniques (encodings, paddings) to increase efficiency and accuracy



17 **Further Improvements**

Large quantities of data are not required to train the model



This cycle of de novo drug design provides **scalable generation of molecules** with **multiobjective optimization**.

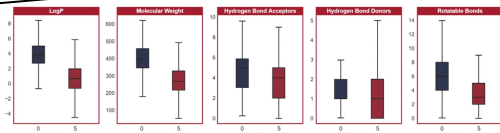
Unrealistic or inferior molecules with one good property can go through the selection

16 **Limitations**

Conclusion

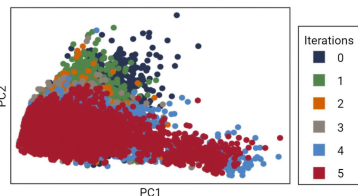
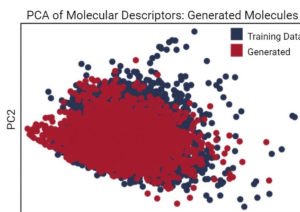


15 **Advantages**



14 After retraining the model with selected molecules, the molecules properties were visualized again.

13 Molecules were evaluated based on five chemical properties then their properties were visualized using **PCA**



11 One million characters were sampled from the LSTM model, yielding **19,722 molecules**, none of which were in the original training data.

12 77% were valid and 6,295 were duplicates. Filtering left **9,415 unique, novel, and valid molecules**.

