

# Computer-Aided Drug Discovery

Seminar #2 - Butler Summer Trainee 2022

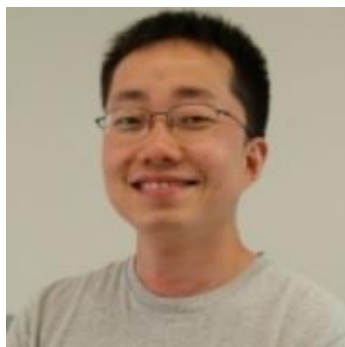
Pinyi Lu

June 30, 2022

# Pinyi Lu

Senior Data Scientist, Cancer Data Science Initiatives, FNLCR

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## My Experience

- 16+ years research experience on translational informatics
- Collaboration with pharmaceutical companies and healthcare systems
- Recent works include:
  - JDACS4C Pilot 1 – Predictive Modeling for Pre-Clinical Screening
  - AI-based drug discovery and design of anticancer drugs that target CENP-E

## Education

- Post-doctoral Appointment, University of Texas Health Science Center
- Ph.D., Computational Biology and Chemoinformatics, Virginia Tech
- M.S., Computer Science, Virginia Tech

## Keywords

- CADD and Machine Learning
- Precision Medicine and Translational Research
- Badminton and Pickleball

JDACS4C: Joint Design of  
Advanced Computing  
Solutions for Cancer;  
CADD: Computer-Aided  
Drug Discovery

# Seminar #2 by Dr. Pinyi Lu

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**Date** - Thu June 30, 2022, 11:00 am – 12:00 pm ET

**Meeting Link** - [https://teams.microsoft.com/l/meetup-join/19%3ameeting\\_YjRiZGU5NzMtOTdlZC00OTQ5LTk2YWEtNzAzNmIwMmUwZWl5%40thread.v2/0?context=%7b%22Tid%22%3a%2214b77578-9773-42d5-8507-251ca2dc2b06%22%2c%22Oid%22%3a%222e31b4fb-2c54-4995-be48-173a24bc5b84%22%7d](https://teams.microsoft.com/l/meetup-join/19%3ameeting_YjRiZGU5NzMtOTdlZC00OTQ5LTk2YWEtNzAzNmIwMmUwZWl5%40thread.v2/0?context=%7b%22Tid%22%3a%2214b77578-9773-42d5-8507-251ca2dc2b06%22%2c%22Oid%22%3a%222e31b4fb-2c54-4995-be48-173a24bc5b84%22%7d)

**TITLE** - Computer-Aided Drug Discovery

**PRESENTER** – Pinyi Lu, Ph.D., Senior Data Scientist, FNLCR

**ABSTRACT** - Computer-aided drug discovery (CADD) has become an important part of the drug discovery process due to the reduced cost of computational methods and the increased availability of three-dimensional structural information. In this seminar, an introduction to CADD will be addressed. I will compare structure-based and ligand-based modeling, focusing primarily on molecular docking and quantitative structure-activity relationship modeling. In addition, I will present an example of application of computational methods in drug discovery and highlight some considerations in the application of CADD.

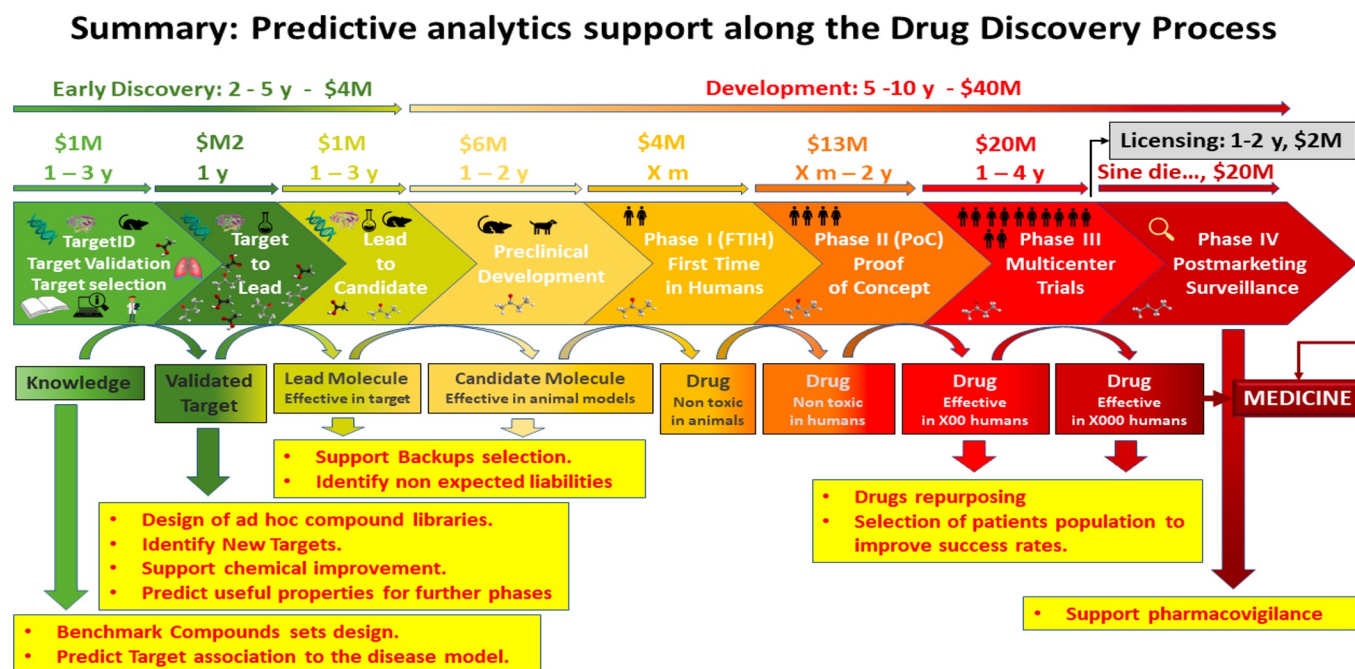
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Part I

# Introduction to Computer-Aided Drug Discovery

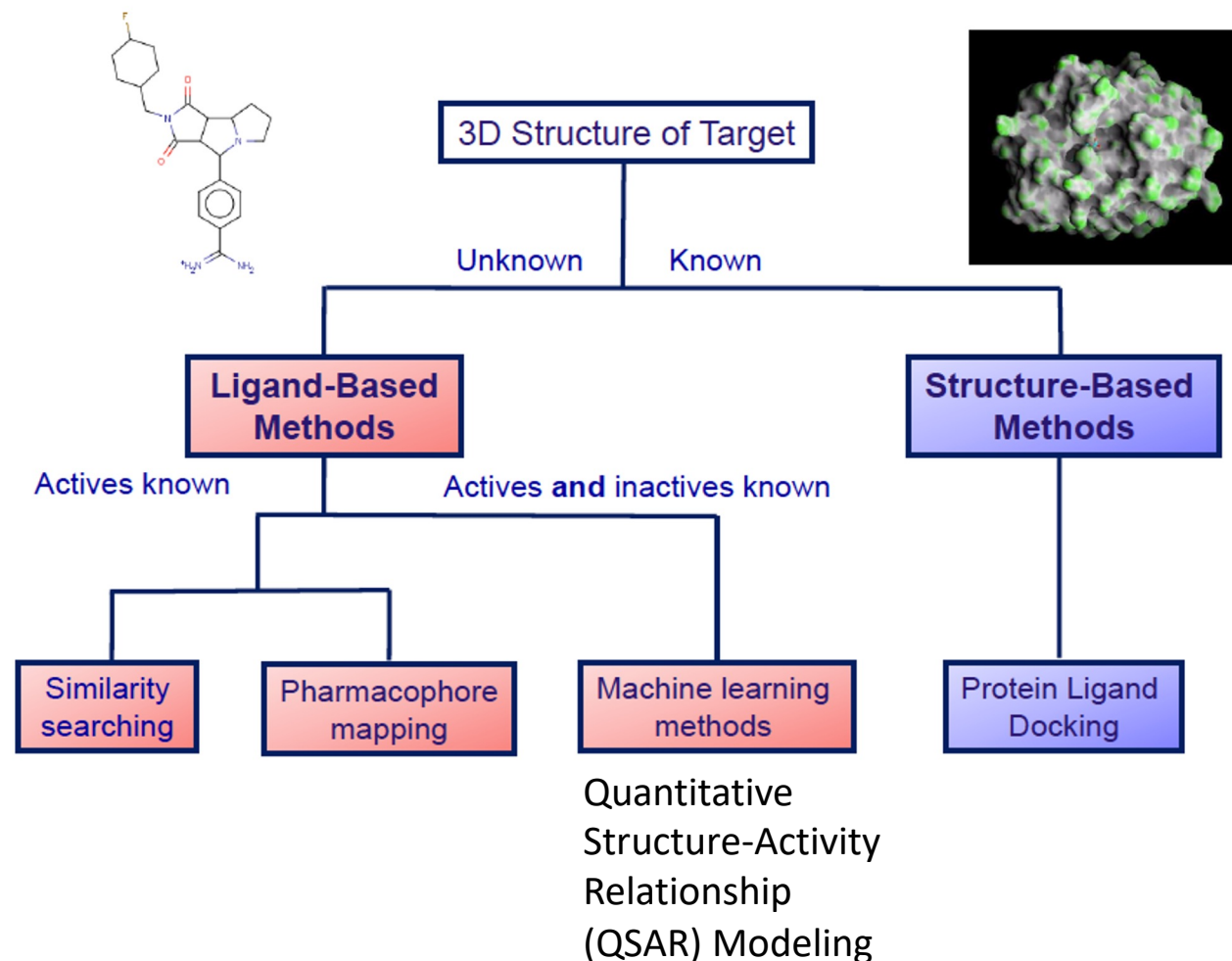
# The Drug Discovery Challenge

- “Drug like” chemical matters, estimated between  $10^{22}$  and  $10^{60}$  unique molecules, evaluating them is a time-consuming and expensive process
- Computer-aided drug discovery, able to save cost, as well as generate a complete traceable and reproducible process



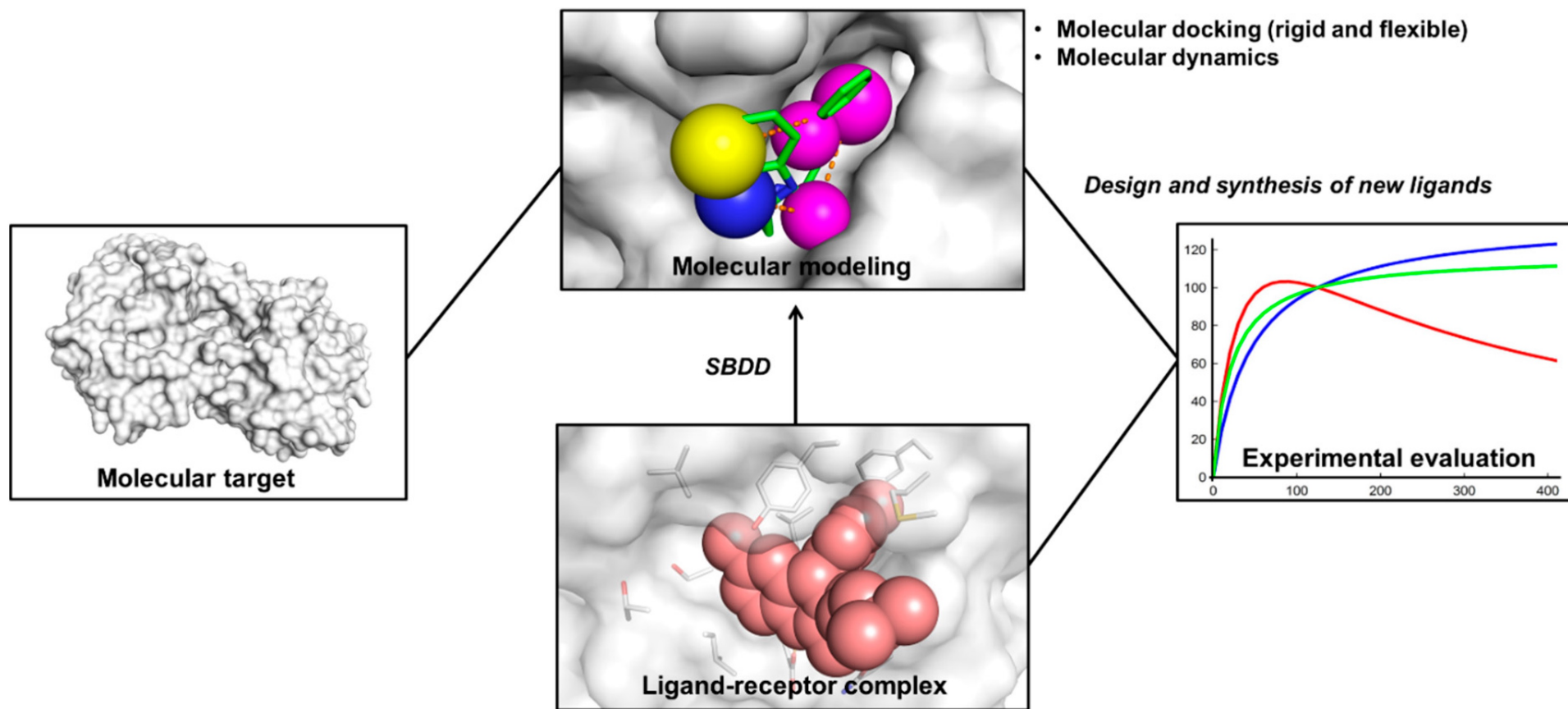
Source: Zhao L. et al. Drug Discov Today (2020); <https://doctortarget.com/machine-learning-applied-drug-discovery/>

# Computer-Aided Drug Discovery



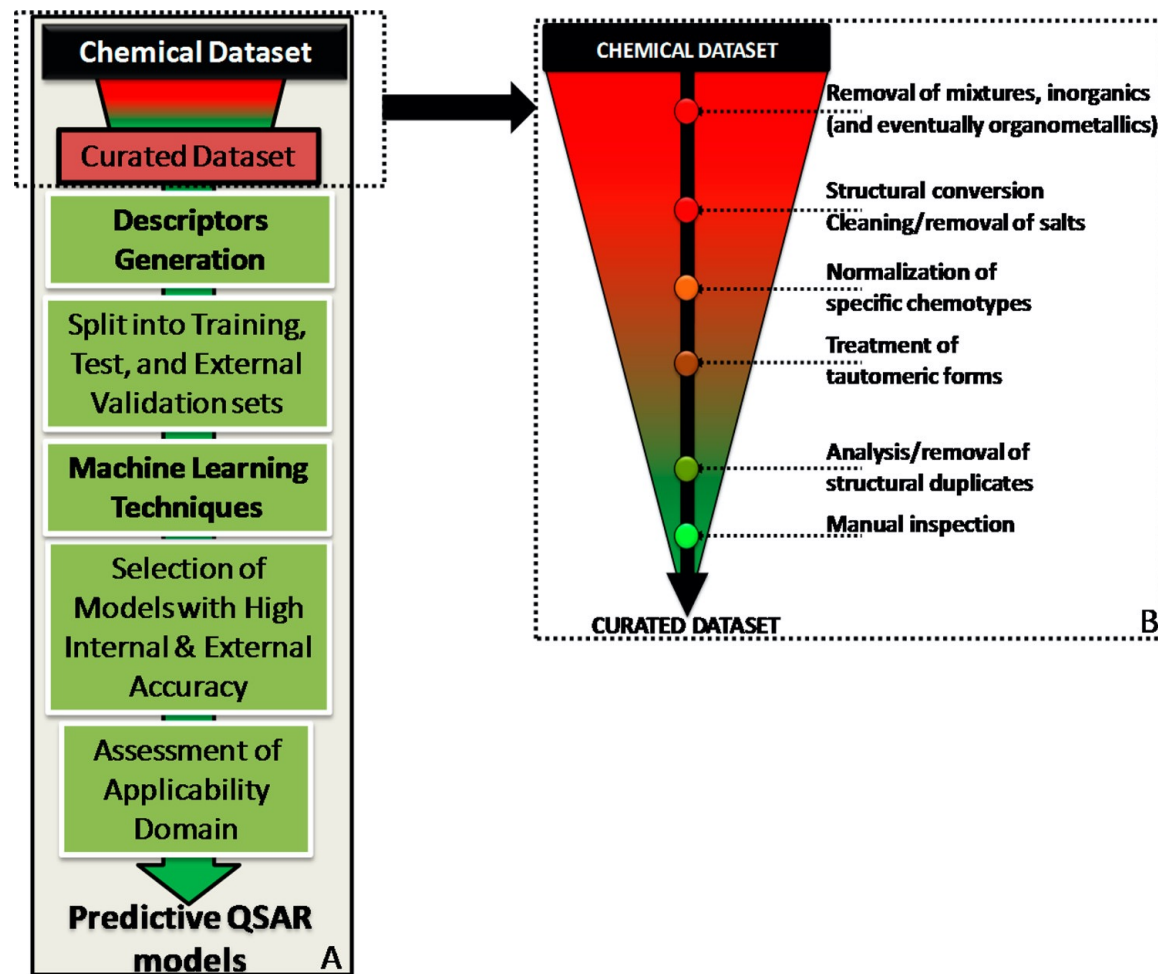
Source: Gillet V. Ligand-Based and Structure-Based Virtual Screening (2013)

# Structure-based Modeling – Molecular Docking



Source: Ferreira L. et al. Molecules (2015)

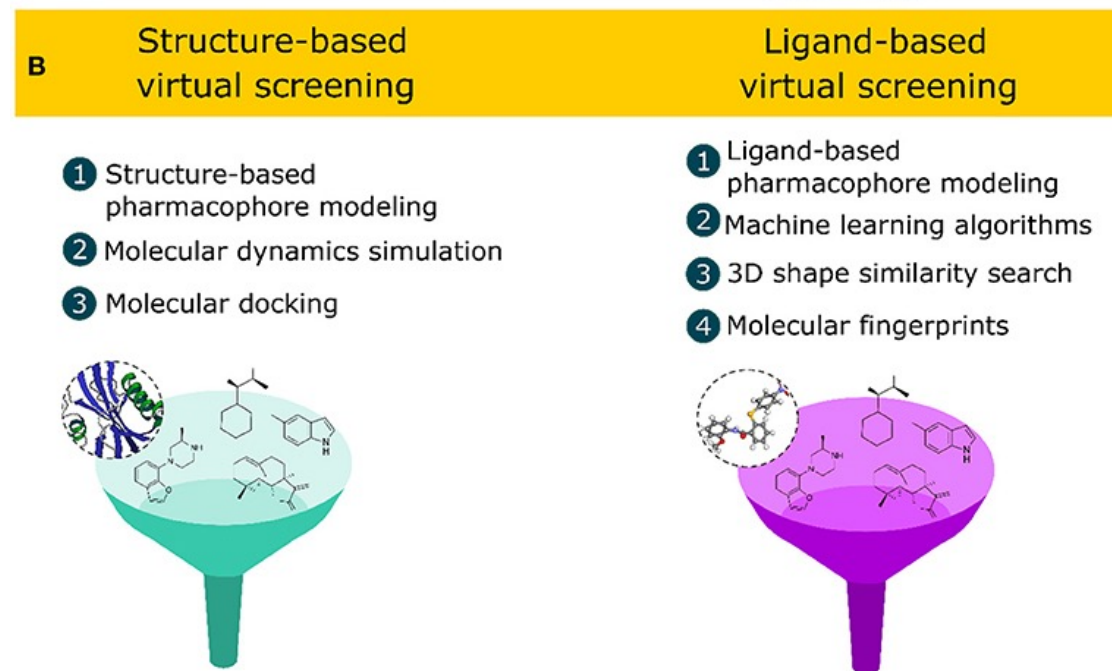
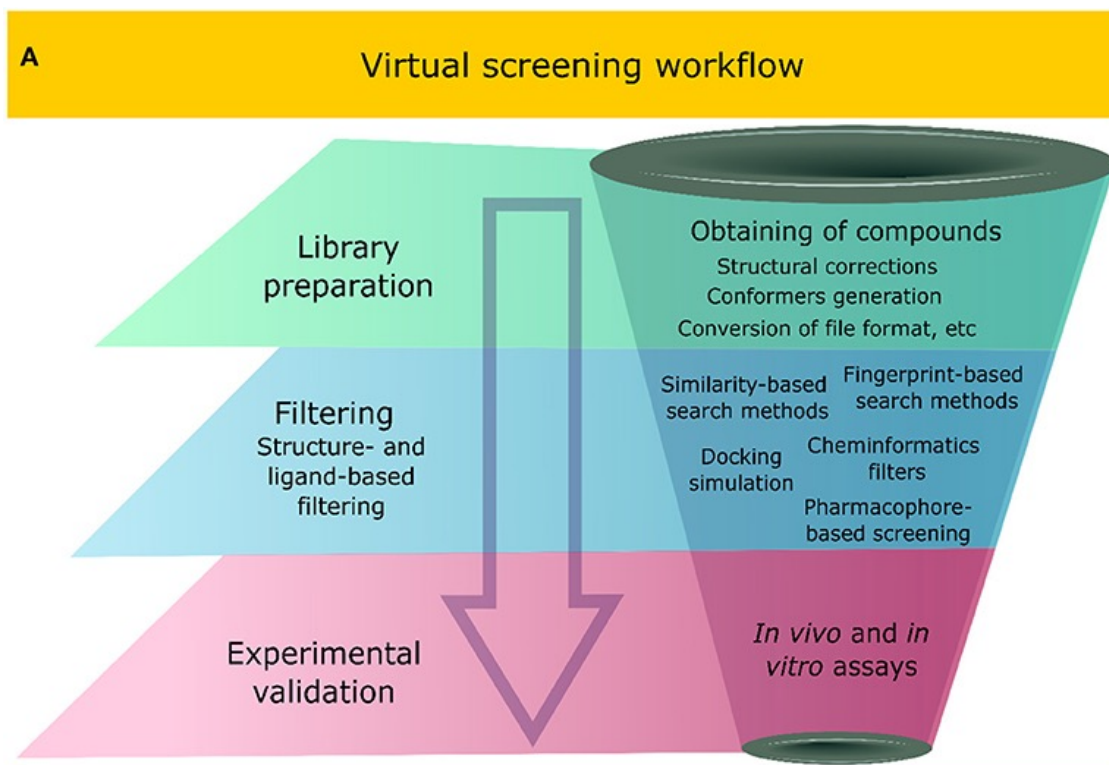
# Ligand-based Modeling – QSAR



Source: Cherkasov A. et al. J Med Chem (2014)



# Virtual Screening



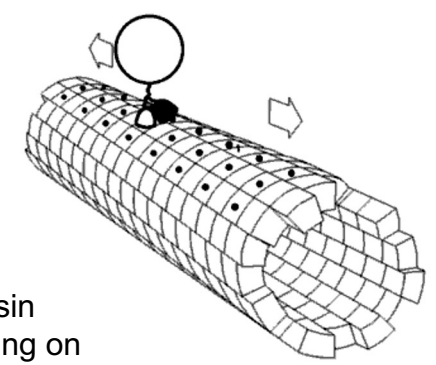
Source: Santana K. et al. Front Chem (2021)

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## Part II

# Example of Application of CADD

# Centromere-associated protein-E



Kinesin  
walking on  
a microtubule

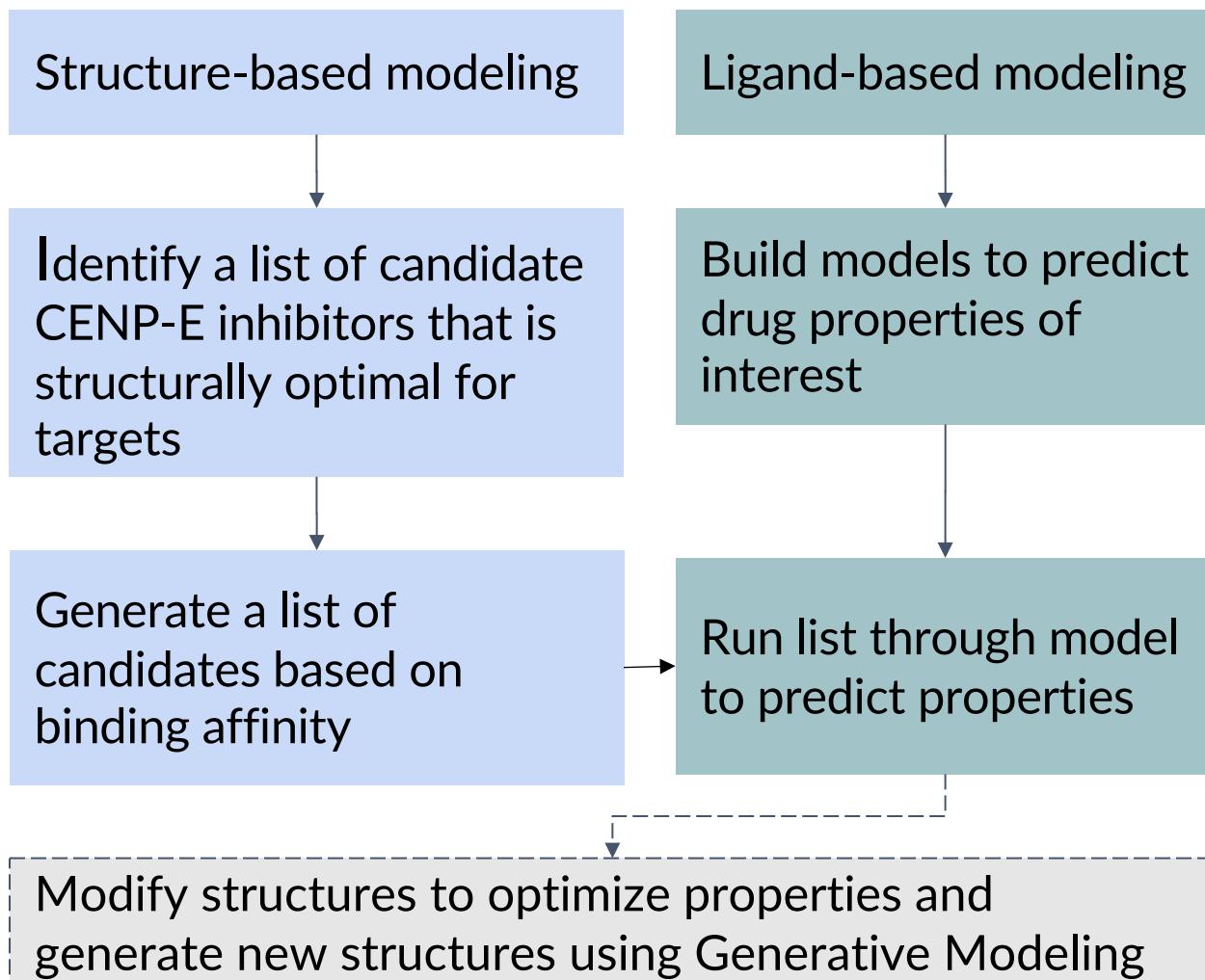
- Centromere-associated protein-E (CENP-E) is a mitotic spindle motor protein and inhibition of CENP-E is promising for cancer therapies
- CENP-E inhibitors could activate innate immune pathway and has potential to induce the immunological conversion from cold to hot in cancer cells
- None went beyond phase I clinical trials
- Not enough bioactivity data



# Methodology Overview

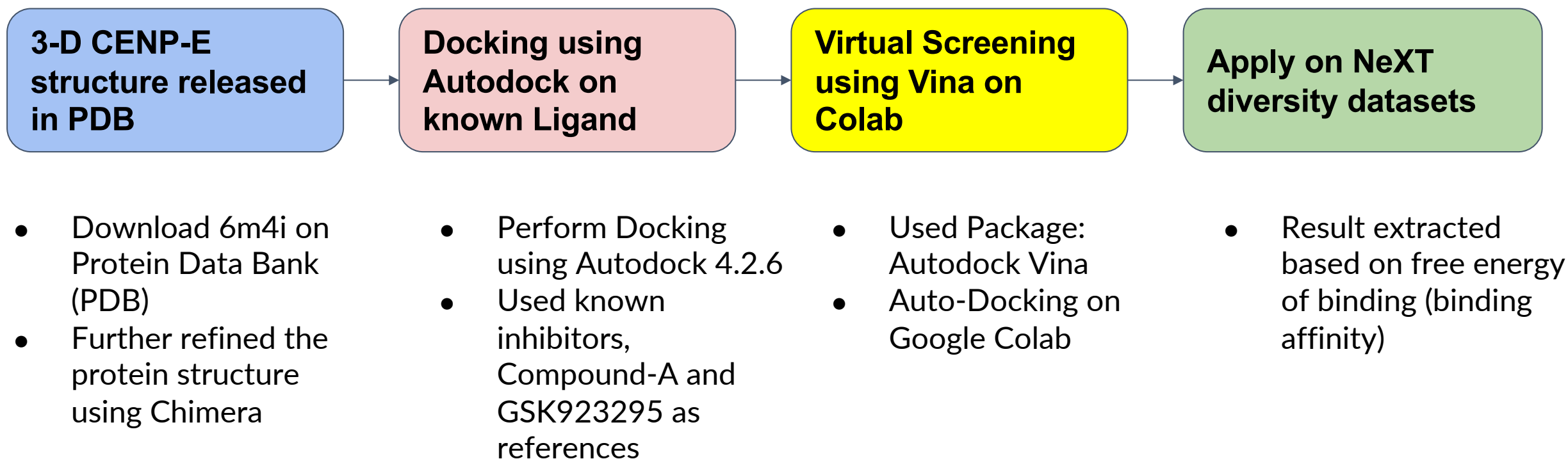
## Objective

Develop a novel workflow integrating structure-based and ligand-based modeling approaches and apply this workflow to aid the discovery and design of novel CENP-E inhibitors

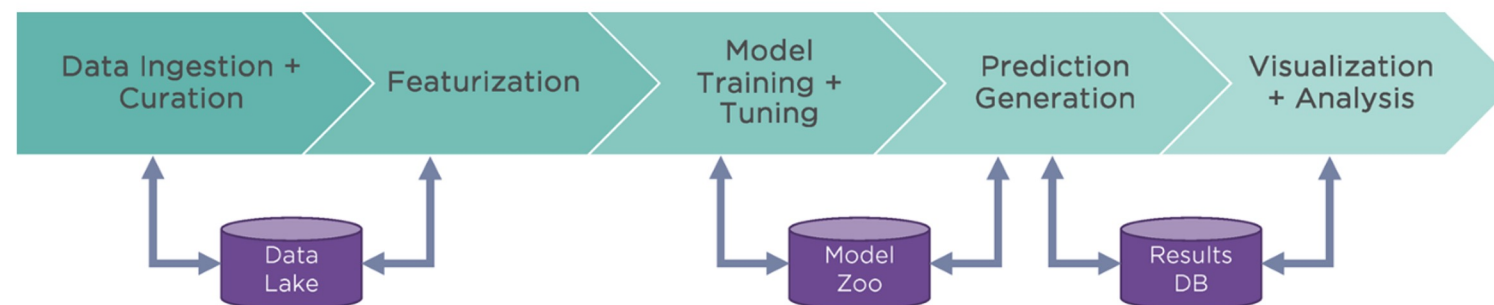


# Methodology – Structure-based modeling on CENP-E

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# Methodology – Ligand-based modeling: AMPL



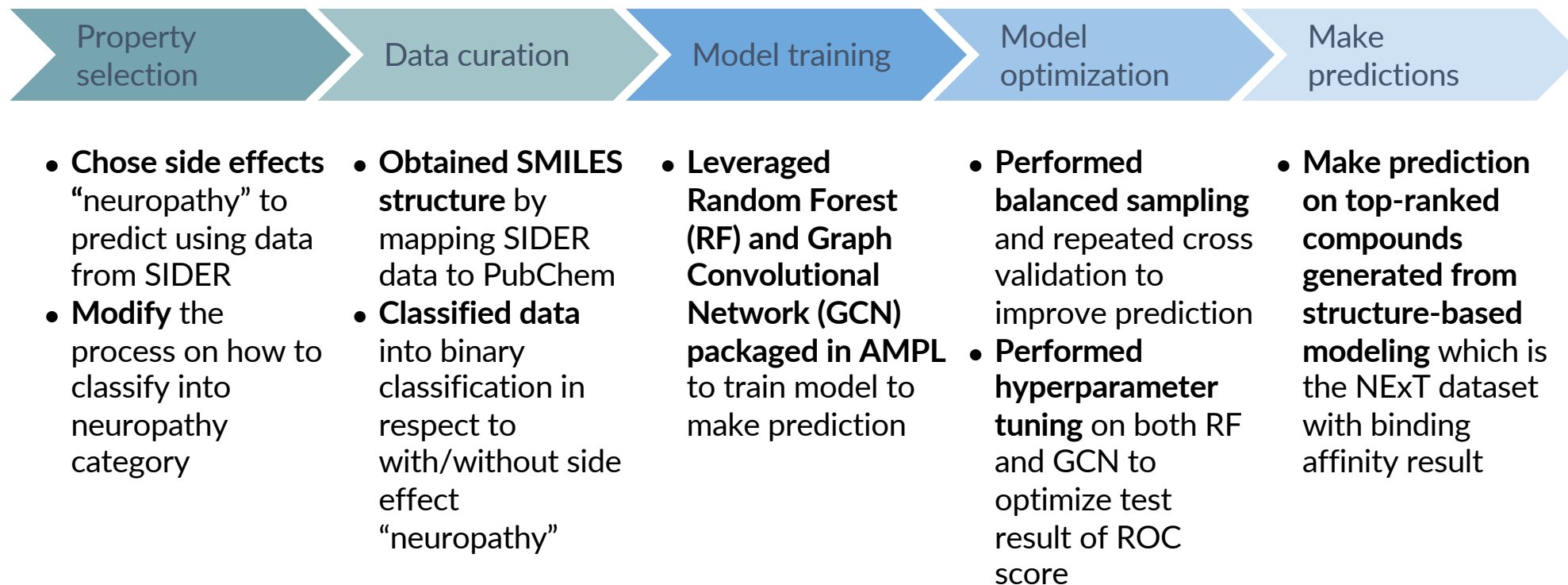
**ATOM Modeling Pipeline (AMPL)**, extends the functionality of the open source library DeepChem and supports an array of ML and molecular featurization

- AMPL supports: Random Forest, XGBoost, Fully Connected Neural Network, Graph Convolutional Neural Network
- Featurizers: Extended connectivity fingerprints (ECFP); graph convolution latent vectors; Mordred open source package; Commercial software package Molecular Operating Environment (MOE)

Source: Minnich AJ et al. J Chem Inf Model (2020)

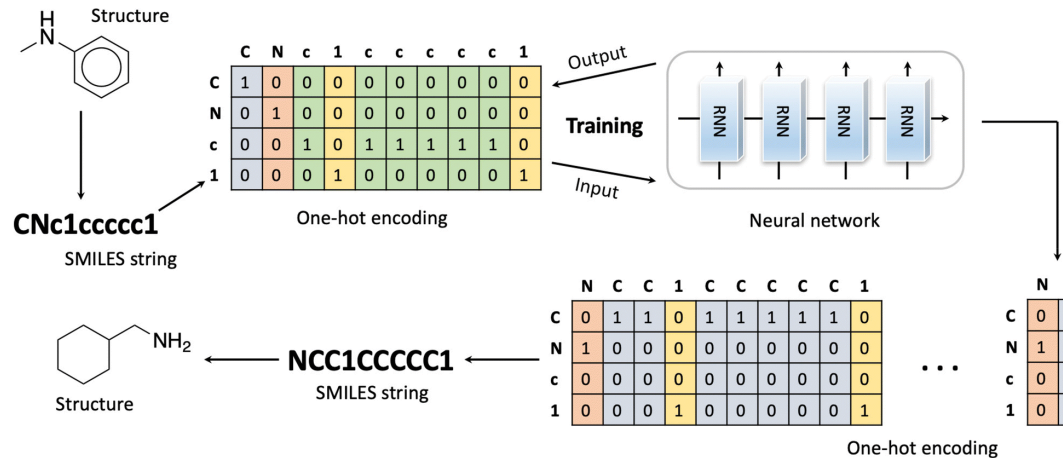
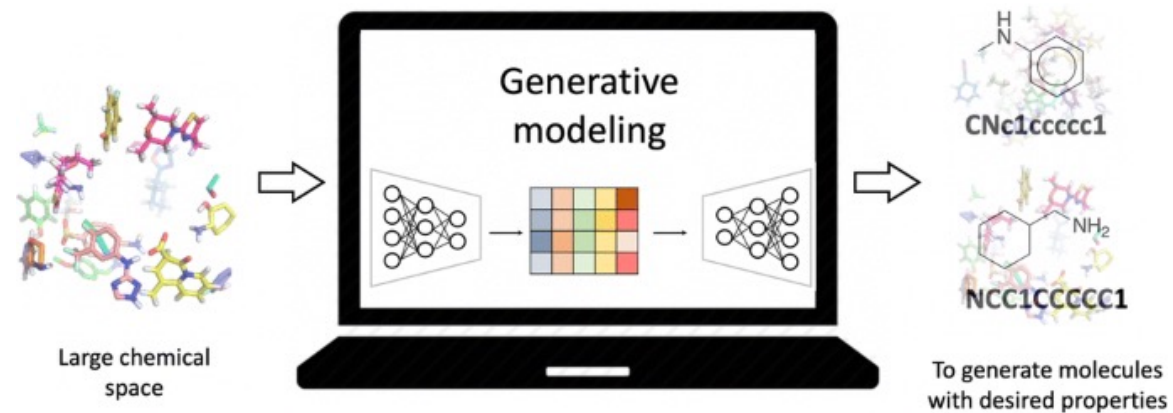
# Methodology – Prediction on side effects

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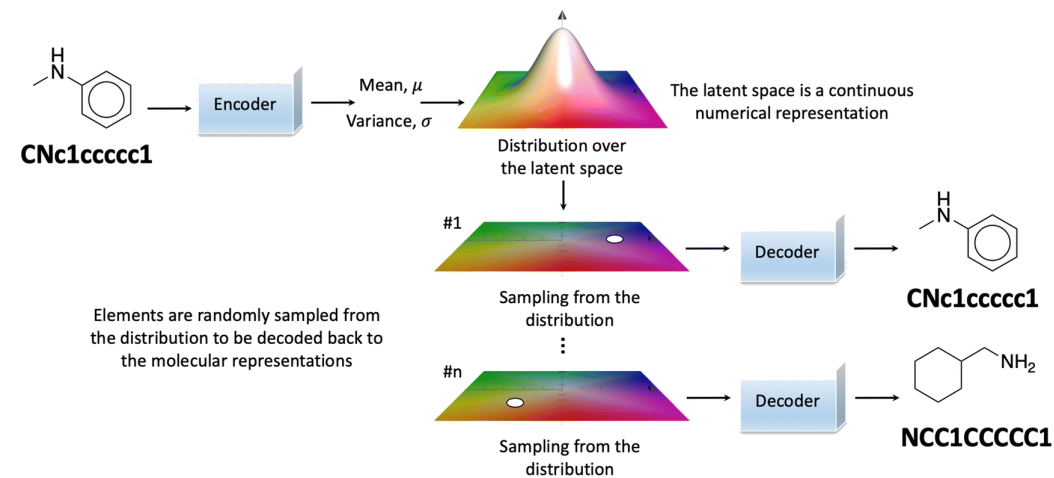




# Methodology – Generative Modeling



Recurrent Neural Network (RNN)



Variational Autoencoders (VAE)

Source: Bian Y et al. J Mol Model (2021)



# Conclusion

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- Identified ligands which have high binding affinity with CENP-E and are favorable in terms of “Neuropathy” side effect
- Performed structure-based and ligand-based modeling to predict free energy of binding and side effect of potential CENP-E ligands
- Performed hyperparameter optimization to improve AMPL model performance
- Applied generative modeling to generate new structures of ligands

# Acknowledgements

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- Columbia University
- Frederick National Laboratory for Cancer Research
- Accelerating Therapeutics for Opportunities in Medicine (ATOM) Consortium

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Thank You!