# Multi-Target Drug Design Using Cheminformatics Approaches



Sudeep Roy, Ph.D.

**Assistant Professor-Cheminformatics(SysBio)** 

**Department of Biomedical Engineering** 

**Brno University of Technology-Brno** 

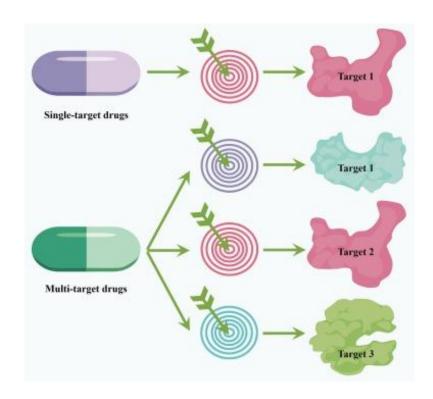
### Introduction

**Definition:** Multi-target Drug Design (MTDD) aims to develop single molecules acting on multiple biological targets.

**Importance:** Suitable for multifactorial diseases like cancer, neurodegeneration.

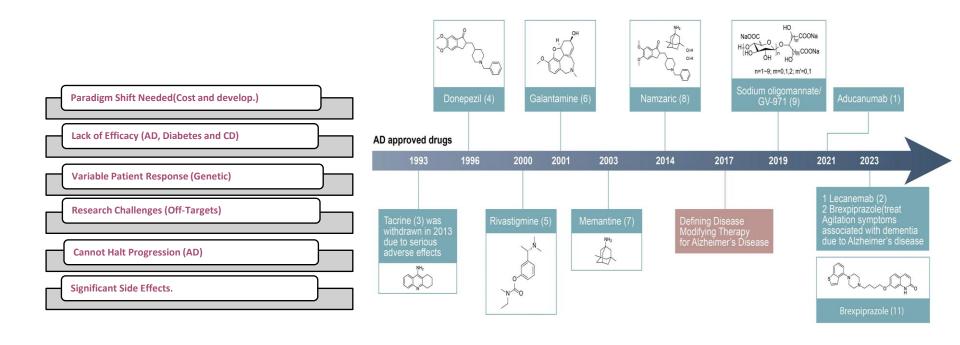
#### **Cheminformatics role:**

Integrates chemical, biological, and pharmacological data.



Source: https://doi.org/10.1016/j.pharmthera.2023.108550

# Limitations of Single-Target Drugs and Recent Therapies



Source: https://www.nature.com/articles/s41392-024-01911-3

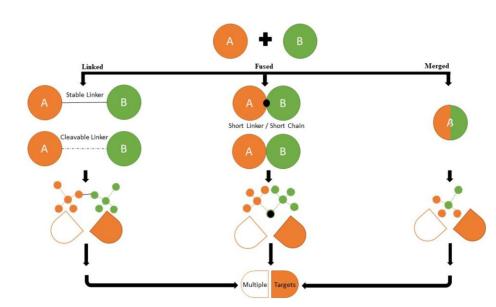
## Advantages of MTDD

Synergistic effects-(Lower Doses)

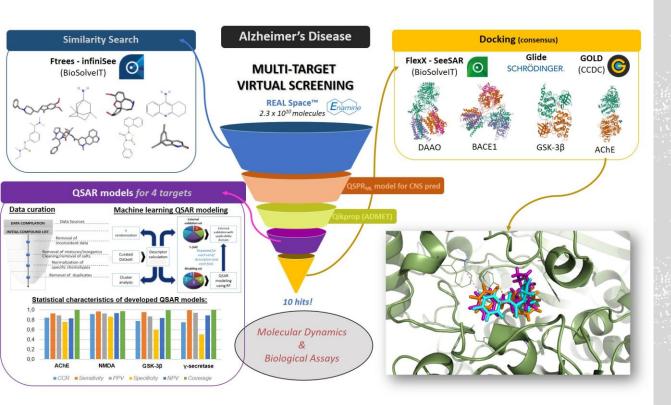
Lower likelihood of resistance-single point mutation (Cancer and Infectious Diseases)

Reduced pill burden (Drug-Drug Interaction)

Better efficacy in complex diseases (Heterogenous Diseases or Diverse patient population)



Source: https://doi.org/10.1002/prp2.70131

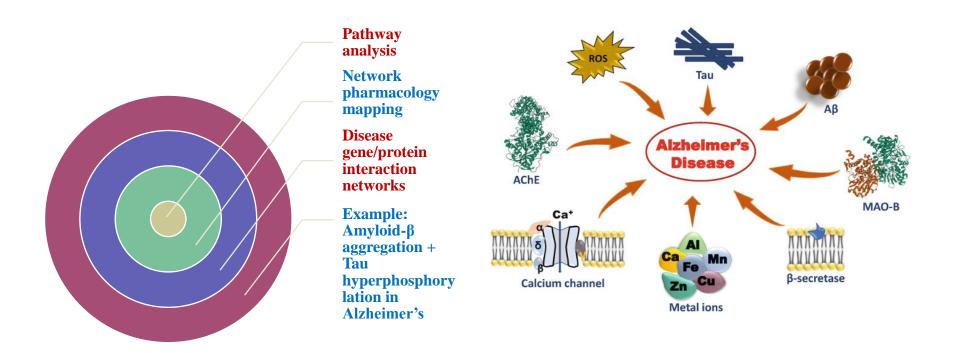


<u>Source: https://www.biosolveit.de/scientific-</u>challenge/project/?projectId=16376243929290

### MTDD Workflow

- 1. Target Identification (System Biology, Network Pharmacology, Disease mapping).
- 2. Ligand Design / Selection (SBDD, Scaffold Hopping, Fragment based Methods)
- 3. Virtual Screening (SBVS, LBVS, Optimal Pharmacological Profiles)
- 4. ADMET Prediction (Exclude Unfavourable pharmacokinetics or high toxicity risk compounds)
- 5. Molecular Dynamics Simulation (Assess Binding stability and potential off-target effects)
- 6. In vitro / In vivo Validation (Synthesis, Animal Models)

## Target Selection Strategies

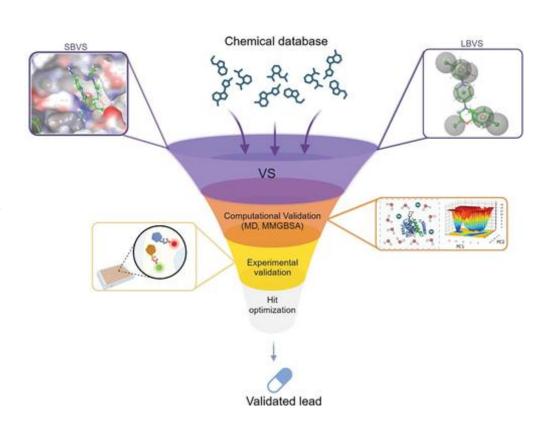


<u>Source: https://link.springer.com/chapter/10.1007/978-981-99-6038-5\_13</u>

## Virtual Screening in MTDD

Ligandbased screening Structurebased screening

Multi-target docking workflows Filtering ligands with affinity for multiple targets



#### Source:

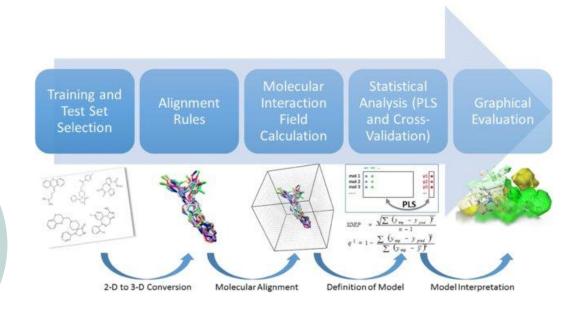
https://www.tandfonline.com/doi/full/10.1080/17460441. 2025.2458666?scroll=top&needAccess=true

## QSAR & Machine Learning

Building predictive models for activity

Integration of chemical descriptors and bioactivity data

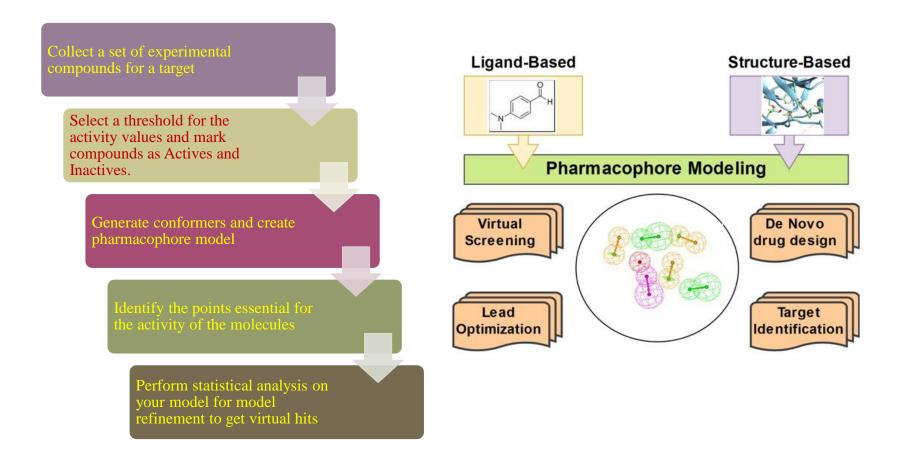
Multitask learning algorithms



#### **Source:**

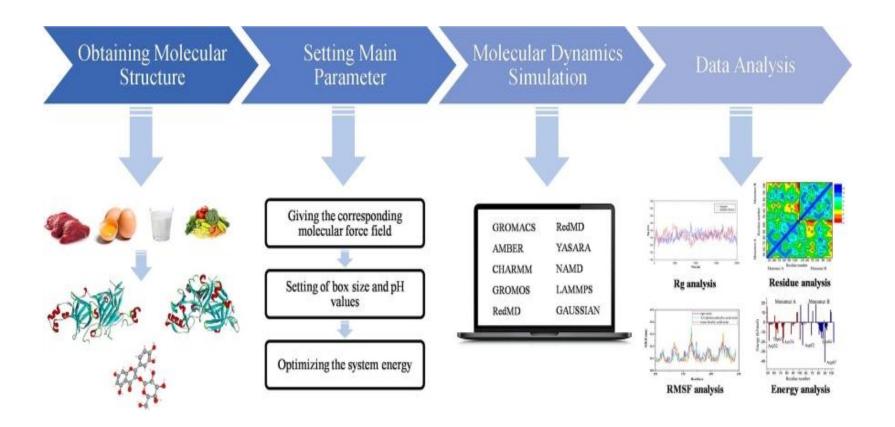
 $\frac{https://link.springer.com/article/10.1007/s10822-019-00231-x}{}$ 

## Pharmacophore Modeling



Source: https://www.profacgen.com/pharmacophore-modeling.htm

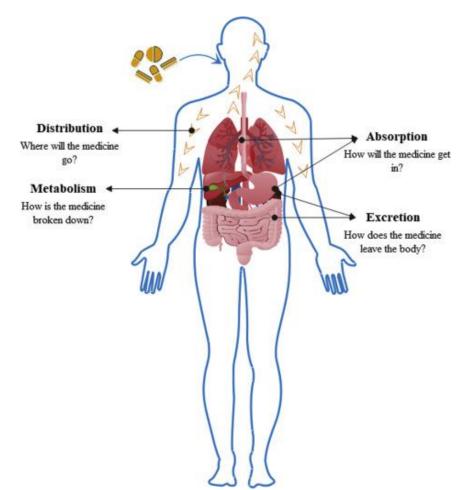
## Molecular Dynamics in MTDD

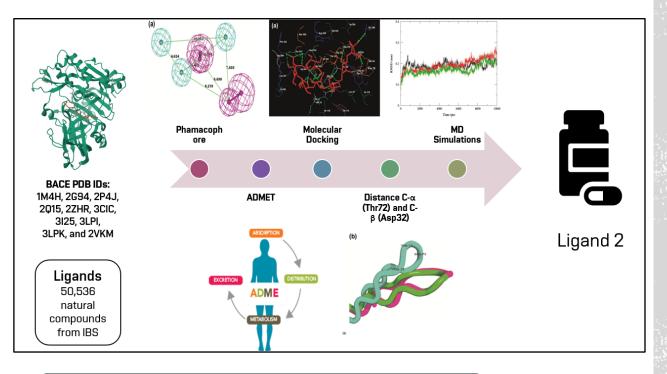


 $\underline{Source: https://www.creative-biostructure.com/molecular-dynamics-}\\ \underline{simulations.html?srsltid=AfmBOopF4wPKdEYd8c0oh\_KQmUK2x2-PABlXthn7ZDTEbpcMD0PNd6wU}$ 

## **ADMET & Toxicity Prediction**

- In silico evaluation of:
  - Absorption
  - Distribution
  - Metabolism
  - Excretion
  - Toxicity
- Filtering out problematic compounds early





#### Purpose

• Identify natural compounds as potential BACE1 inhibitors using *in-silico* approaches.

### **Target Proteins / Pathways**

- BACE1 (β-secretase 1)
- Amyloid-β formation pathway in AD.

#### **Goals Achieved**

- Virtual screening of natural compounds
- 3D–QSAR pharmacophore model developed
- Molecular dynamics confirmed stability
- Promising BACE1 inhibitor leads identified

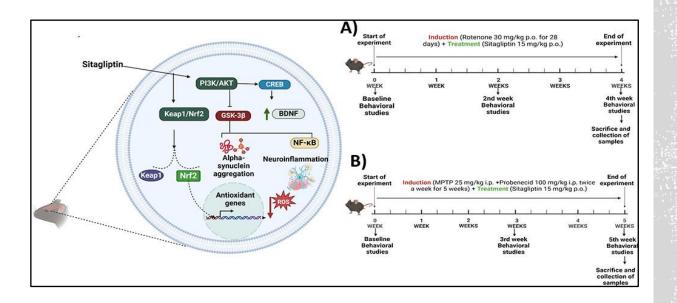
Case Study 1: Molecular docking based virtual screening of natural compounds as potential BACE1 inhibitors: 3D – QSAR pharmacophore mapping and molecular dynamics analysis

#### Source:

https://www.tandfonline.com/doi/full/10.1080/07391102.20 15.1022603

- Large flexible binding sites like BACE1.
- Catalytic ASPDyad (D32 and D228) and Flexible Flap(V67-E77). Imp Residue: T72.
- 10ns Run.
- BBB, Hepatotoxicity,
  PPB, solubility and
  mutagenicity. Scaffolds
  safe for CNS delievery.
- 3 Oligosaccharides Hits.
- Reduction of amyloid levels in AD





Purpose

• Investigate the neuroprotective role of situation in PD mouse models.

Target Proteins / Pathways

- PI3K/AKT → neuronal survival
- Nrf2 → antioxidant defense
- DPP-4 inhibition → upstream trigger

Goals Achieved

- Upregulated PI3K/AKT & Nrf2
- Reduced oxidative stress & neuronal degeneration
- Neuroprotection demonstrated
- Supports repurposing for PD therapy

Case Study 2: Upregulation of the PI3K/AKT and Nrf2 Pathways by the DPP-4 Inhibitor Sitagliptin Renders Neuroprotection in Chemically Induced Parkinson's Disease Mouse Models

Source:

https://pubs.acs.org/doi/10.102 1/acschemneuro.5c00112

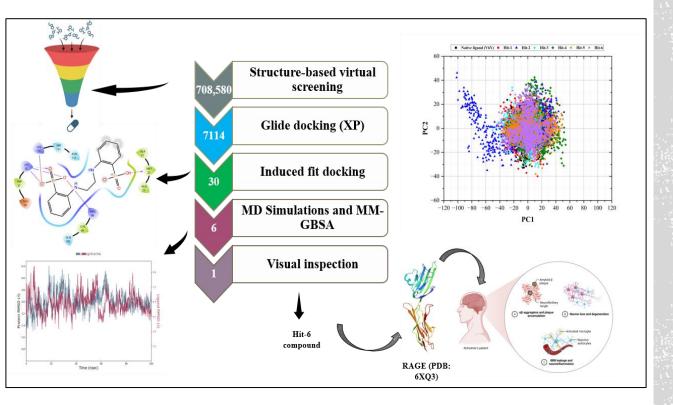
Sita, Lina, Alo, Vilda, saxa-DPP4 Inhibitors

Sita alleviated motor and cognitive deficits.

Upregulated: PI3K, AKT, Nrf2(promotes cellular survival, growth and antioxidant), CREB and BDNF(neuronal survival factors).

Downregulated: GSK-3ß (Neuro and Ototoxicity-ENT), NF-Kb(Cancer, Diabetes, atherosclerosis and auto-immune disorders) and alphasynuclein (PD).

Restored neuronal density and integrity



### **Purpose**

• Identify RAGE inhibitors via in-silico approaches to combat Alzheimer's disease.

### Target Protein/ Pathway

• RAGE—block receptor to inhibit amyloid-β and inflammatory signaling.

### **Goals Achieved**

- Screened over 700,000 compounds.
- Identified Hit-6 as a potential inhibitor of RAGE

Case Study 3: Identifying RAGE inhibitors as potential therapeutics for Alzheimer's disease via integrated insilico approaches

#### Source:

https://www.nature.com/articles/s41598-025-01271-0

**RAGE:** Receptor for advanced glycation end products.

RAGE: Cancer, Diabetes, Kidney Failure, Cardiovascular disease, COPD, Rheumatoid Arthritis, Systemic Lupus, IBS, Asthma and neurological disorders(PD, AD...).

Enamine, TTD, FDA library, ChemDiv, InterBioScreen, Zinc.

Lys39, Lys52, Arg98, Asn112 and Lys110 (HotSpots).





### **Contact**



BioSys\_BUT

Email:

roy@vut.cz