

Lecture 10:

# Drug Discovery and Molecular ML

- AI-powered drug discovery
  - Success stories
- Pipeline transformation

Introduction to Biomedical Datascience

Lecture 10:

# Drug Discovery and Molecular ML

AI-powered drug discovery

Success stories

Pipeline transformation

Introduction to Biomedical Datascience

# Lecture Contents

**Part 1:** Drug Discovery Pipeline

**Part 2:** Molecular Machine Learning

**Part 3:** Practical Applications

## Part 1/3:

# Drug Discovery Pipeline

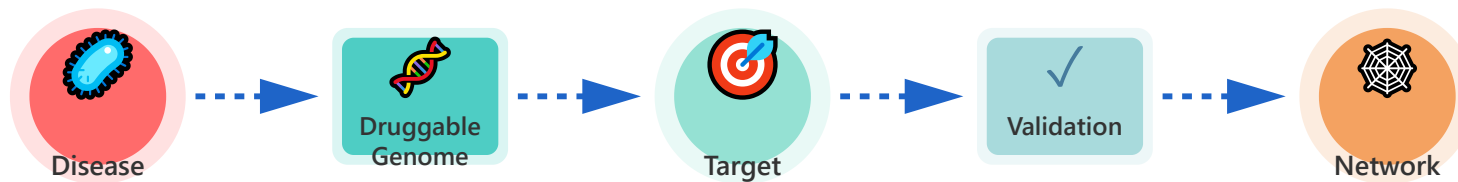
- Traditional vs AI-enhanced approaches
- Time and cost savings
- Success rate improvements

## Part 1/3

# Drug Discovery Pipeline

- Traditional vs AI-enhanced
  - Time and cost savings
- Success rate improvements

# Target Identification



## Disease mechanisms

Understanding biological pathways

## Druggable genome

Identifying targetable proteins

## Target validation

Confirming therapeutic relevance

## Genetic evidence

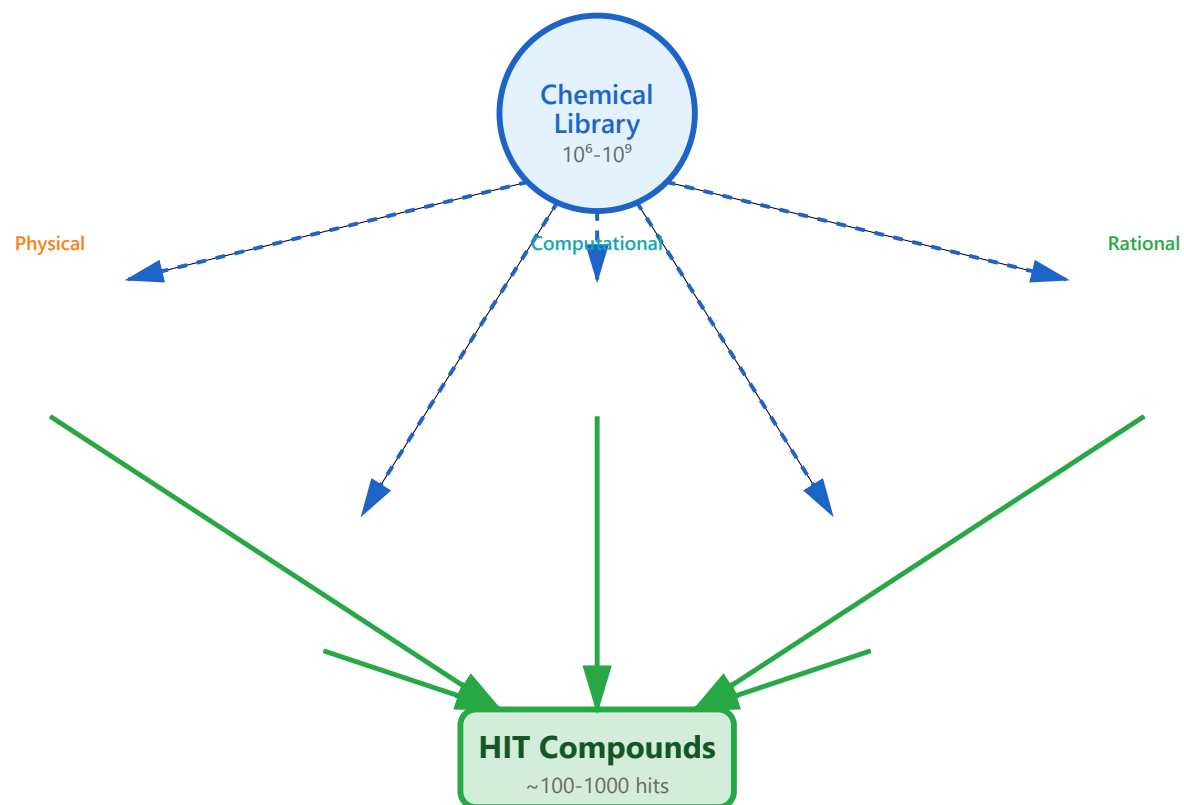
Human genetics support

## Network approaches

Systems biology integration

# Lead Discovery

aries  
al space

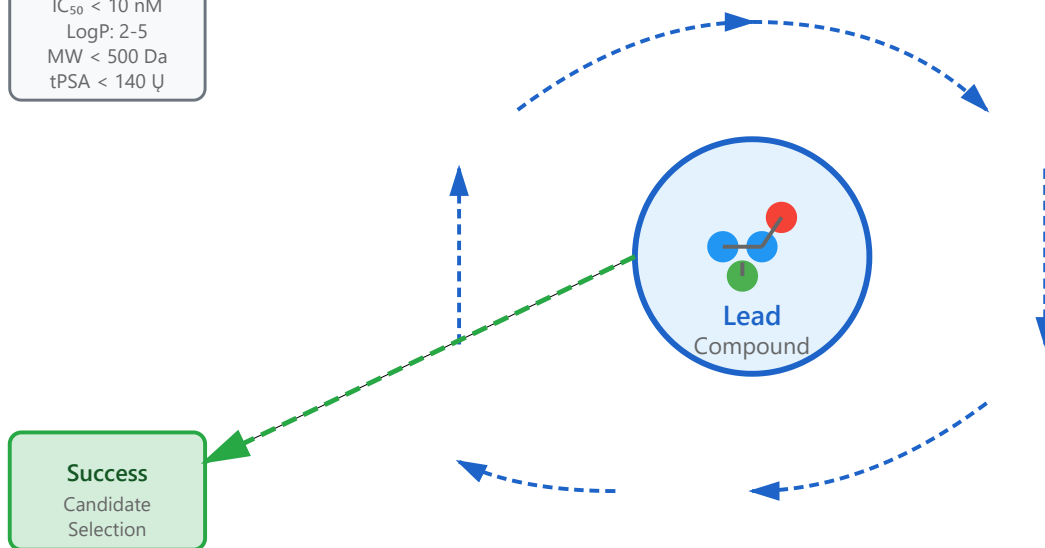


# Lead Optimization

Technical  
ability

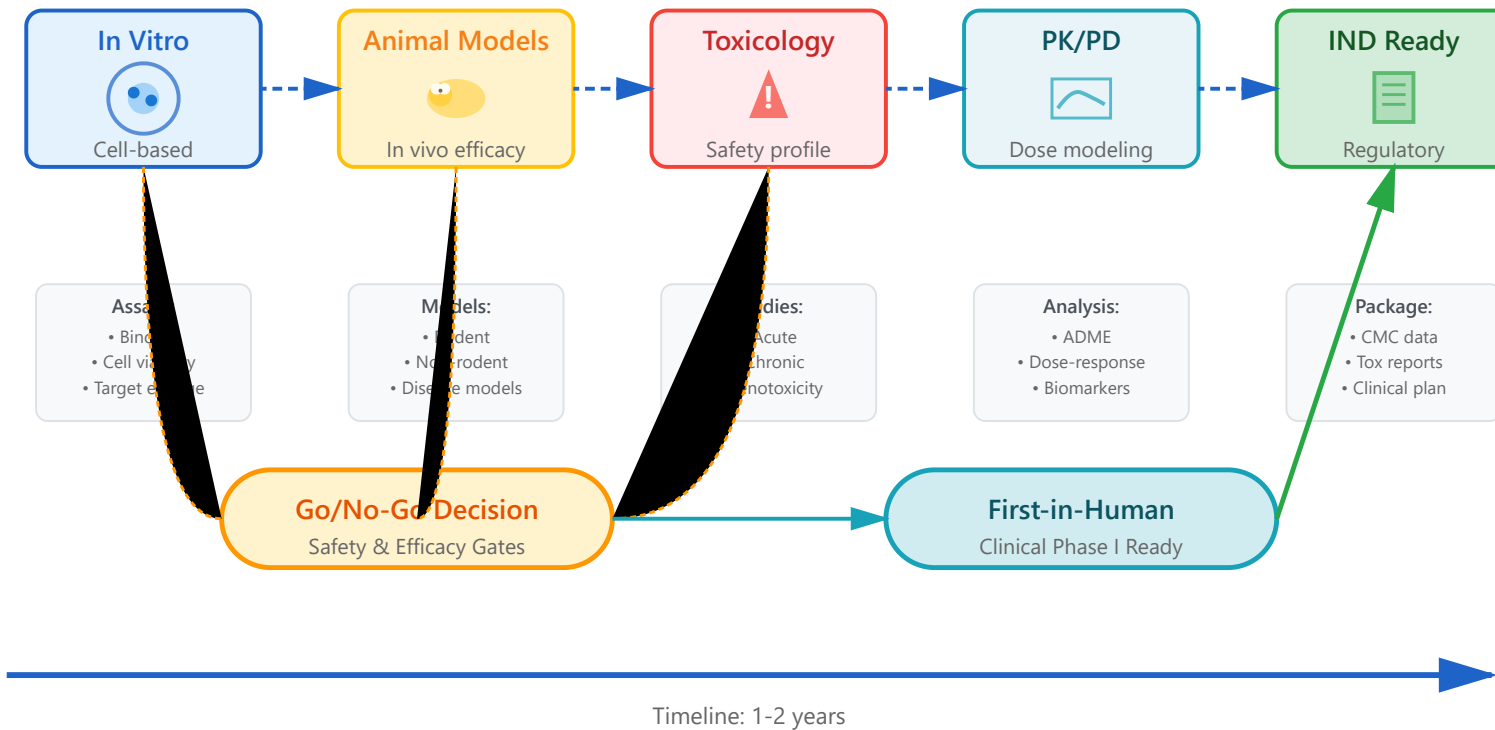
**Metrics**  
IC<sub>50</sub> < 10 nM  
LogP: 2-5  
MW < 500 Da  
tPSA < 140 Å

**Iterations**  
**3-5**  
cycles typical

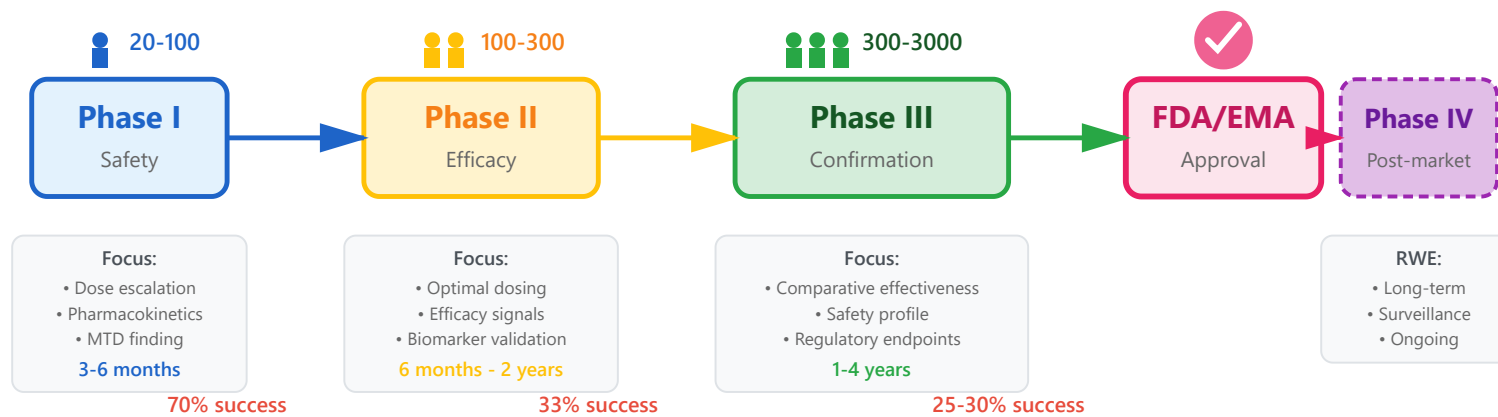




# Preclinical Studies



# Clinical Trials



## Biomarker Strategies

- Patient selection
- Response monitoring
- Surrogate endpoints
- Precision medicine

## Adaptive Trials

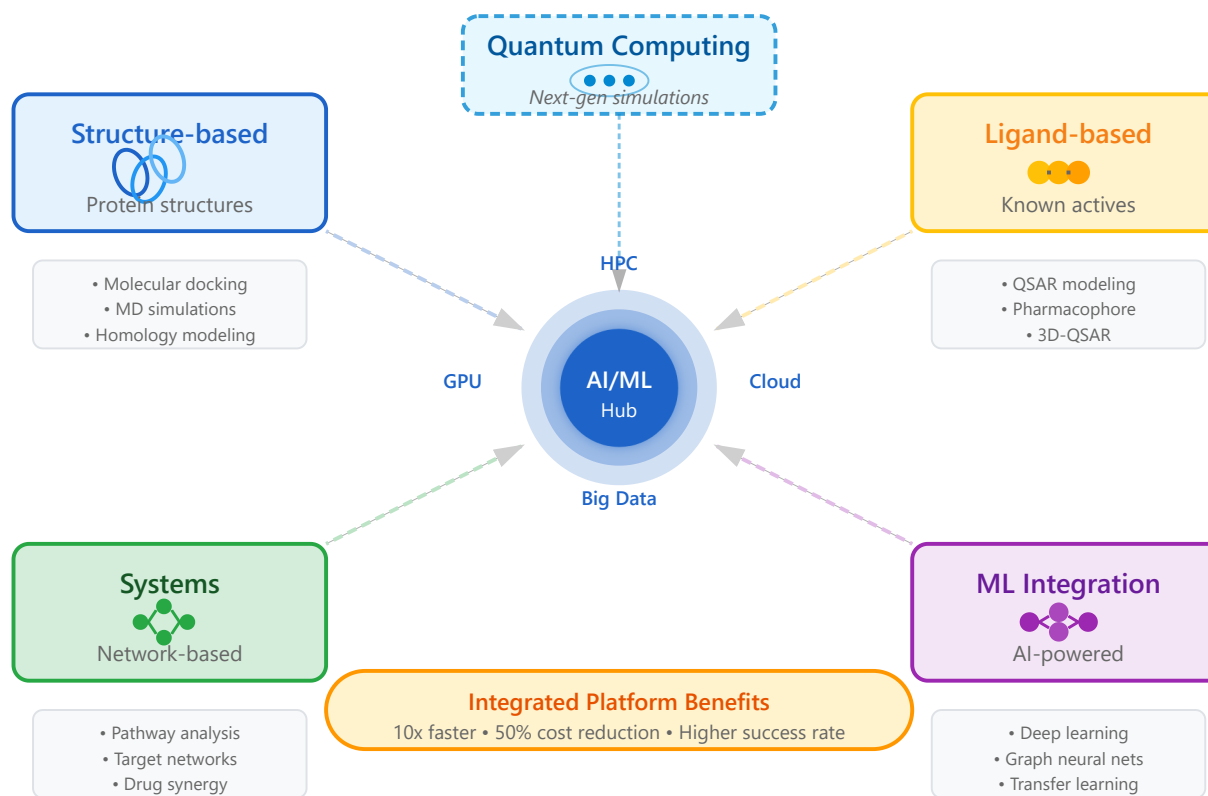
- Dose finding
- Sample size re-estimation
- Seamless phase transition
- Bayesian approaches

## Real-World Evidence

- EHR data mining
- Claims analysis
- Registry studies
- Digital biomarkers

Total: 10-15 years, \$1-3 billion

# Computational Approaches



**Part 2/3:**

# **Molecular Machine Learning**

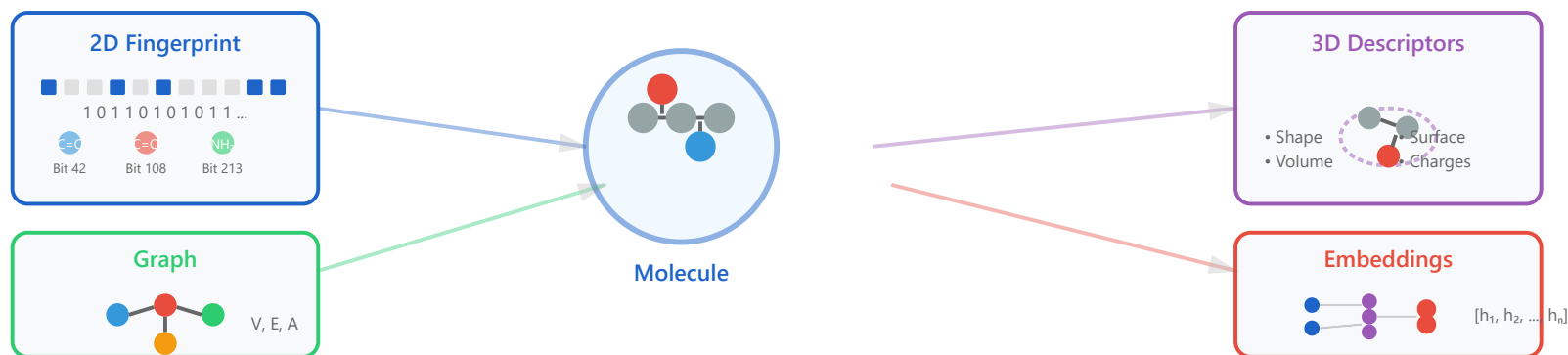
- Representation learning
- Property prediction
- Generative models

**Part 2/3**

# **Molecular ML**

- Representation learning
  - Property prediction
  - Generative models

# Molecular Representations



## 2D fingerprints

Binary feature vectors

## Graph representations

Molecular graph structures

## Multi-view learning

Combining multiple representations

## 3D descriptors

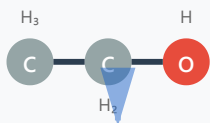
Geometric and conformational features

## Learned embeddings

Deep learning representations

# SMILES Notation

Ethanol



CCO

SMILES String

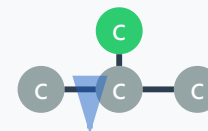
Benzene



c1ccccc1

Ring Closure

Branched



CC(C)C

Branching ()

## Syntax rules

String-based molecular encoding

## Canonical SMILES

Unique molecular representation

## SMARTS patterns

Substructure search patterns

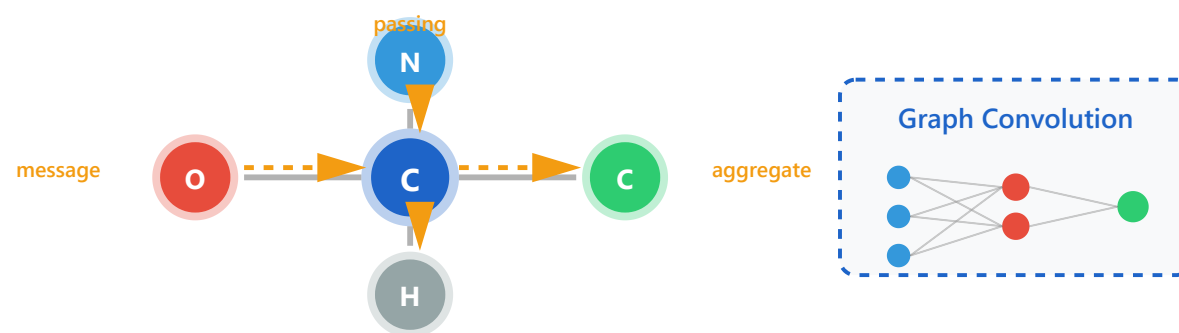
## Tokenization

Breaking into meaningful units

## Augmentation strategies

Data augmentation techniques

# Graph Neural Networks



● Atoms (Nodes) — Bonds (Edges) — Message Flow

## Molecular graphs

Atoms as nodes, bonds as edges

## Message passing

Information flow between atoms

## Graph convolutions

Feature aggregation operations

## Attention mechanisms

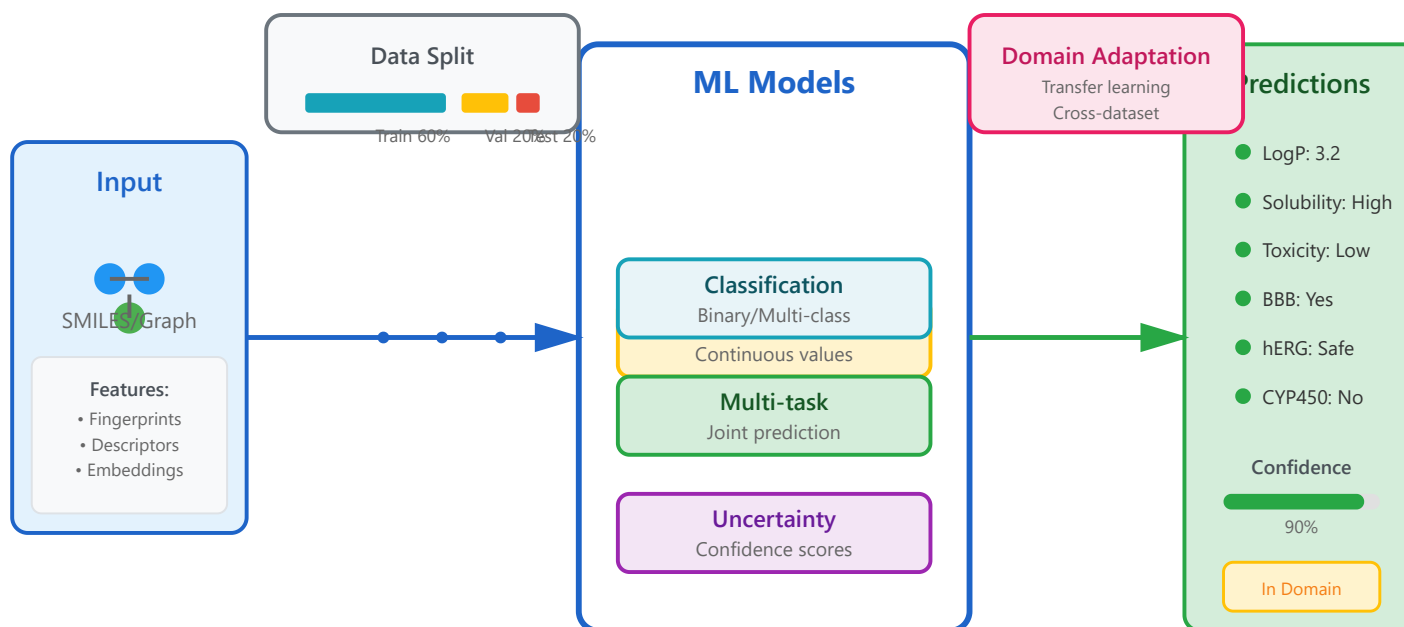
Weighted information aggregation

## Pooling strategies

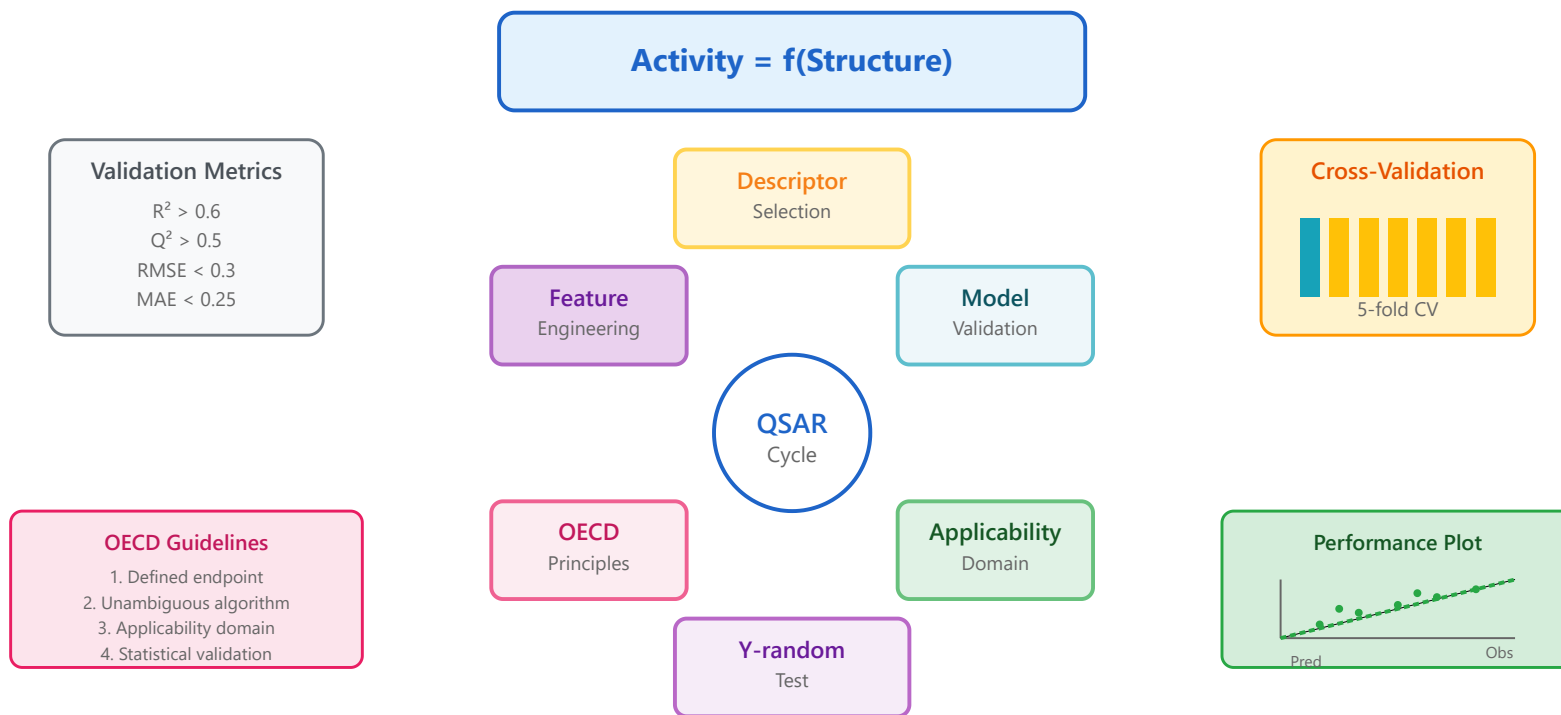
Graph-level representations



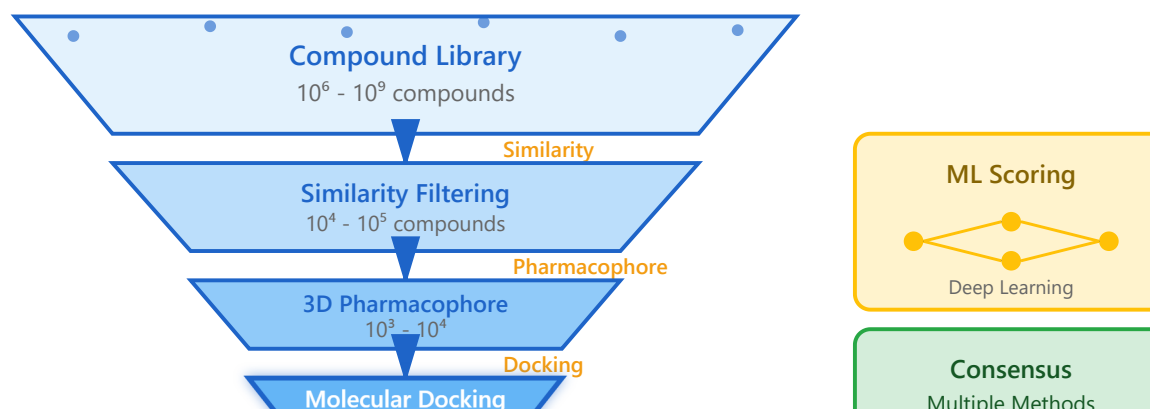
# Property Prediction



# QSAR Modeling



# Virtual Screening



## Similarity searching

Finding similar active compounds

## Pharmacophore modeling

3D feature-based screening

## Docking scores

Protein-ligand binding prediction

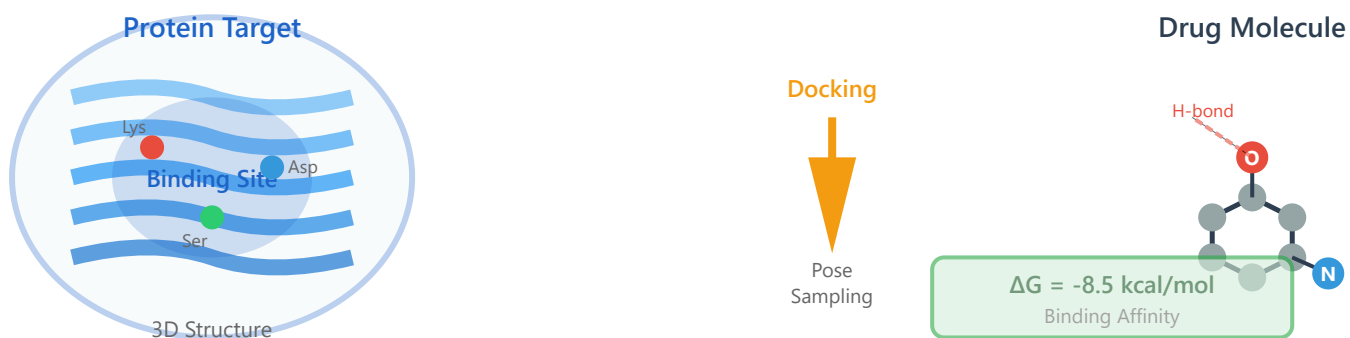
## ML scoring functions

Learning-based scoring

## Consensus approaches

Combining multiple methods

# Docking Simulation



Scoring: vdW + Electrostatic + H-bonds + Solvation + Entropy

## Protein preparation

Structure optimization

## Binding site detection

Active site identification

## Conformational sampling

Exploring binding modes

## Scoring functions

Binding affinity estimation

## Induced fit

Protein flexibility modeling

**Part 3/3:**

# **Practical Applications**

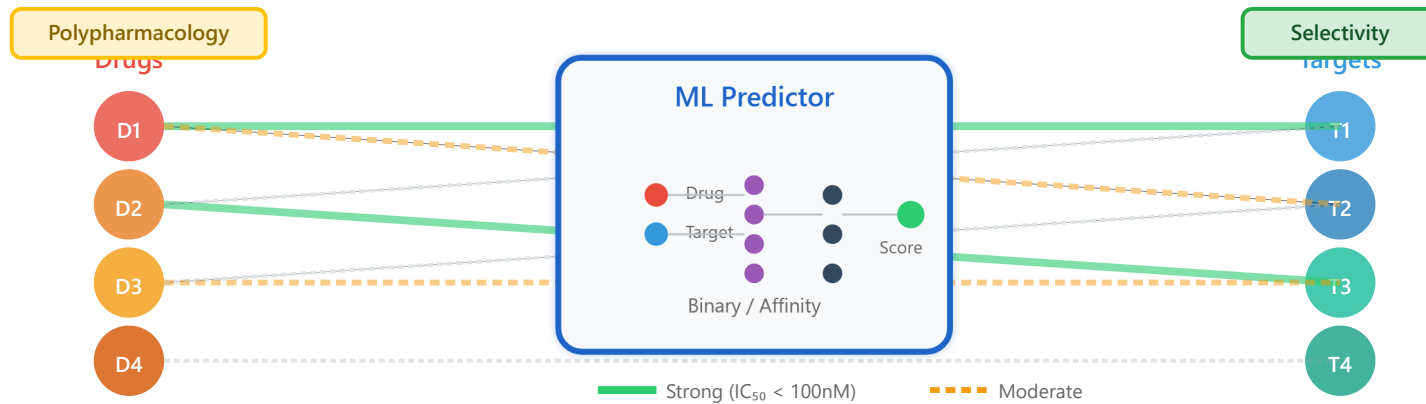
- Practical implementations
- Success metrics
- Future directions

## Part 3/3

# Applications

- Practical implementations
  - Success metrics
  - Future directions

# Drug-Target Interaction



## Binary classification

Predicting interaction likelihood

## Binding affinity

Quantitative affinity prediction

## Kinome profiling

Kinase selectivity analysis

## Polypharmacology

Multi-target interactions

## Off-target prediction

Safety profiling

# Side Effect Prediction

## ADR databases

Adverse drug reaction resources

## Network approaches

Drug-target-disease networks

## Chemical similarity

Structure-based prediction

## Target-based

Mechanism-based approaches

## Clinical translation

Preclinical to clinical



# Drug Repurposing

## Indication expansion

New therapeutic uses

## Signature matching

Disease signature comparison

## Network propagation

Disease module identification

## Clinical evidence

Real-world validation

## IP considerations

Patent and exclusivity

# Bioactivity Prediction

## Activity cliffs

Small structural changes, large activity differences

## Matched pairs

Systematic SAR analysis

## Free energy perturbation

Physics-based predictions

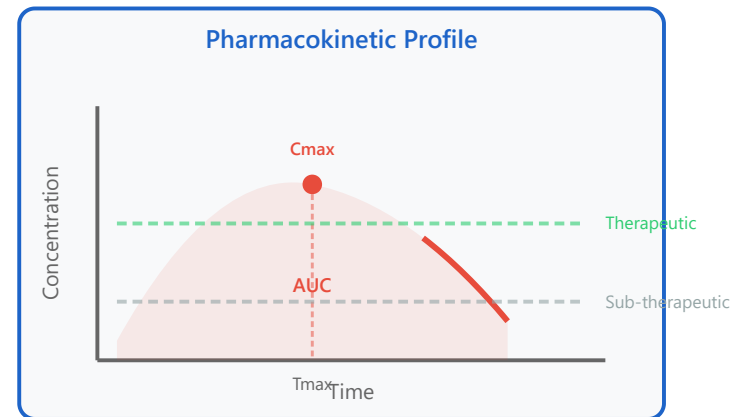
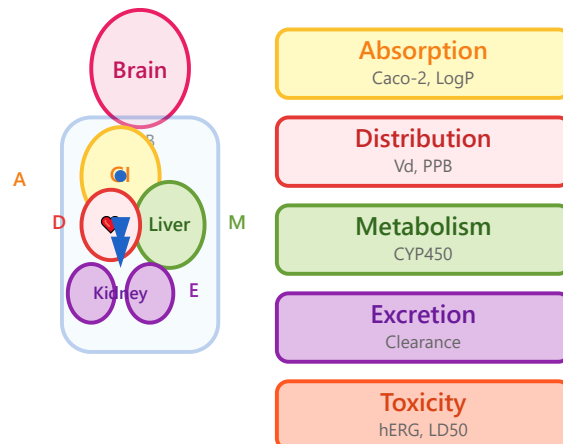
## Active learning

Iterative experiment design

## Experimental validation

Wet-lab confirmation

# ADMET Prediction



## Absorption models

Oral bioavailability prediction

## Distribution (BBB, Vd)

Tissue distribution modeling

## Metabolism (CYP)

Drug metabolism prediction

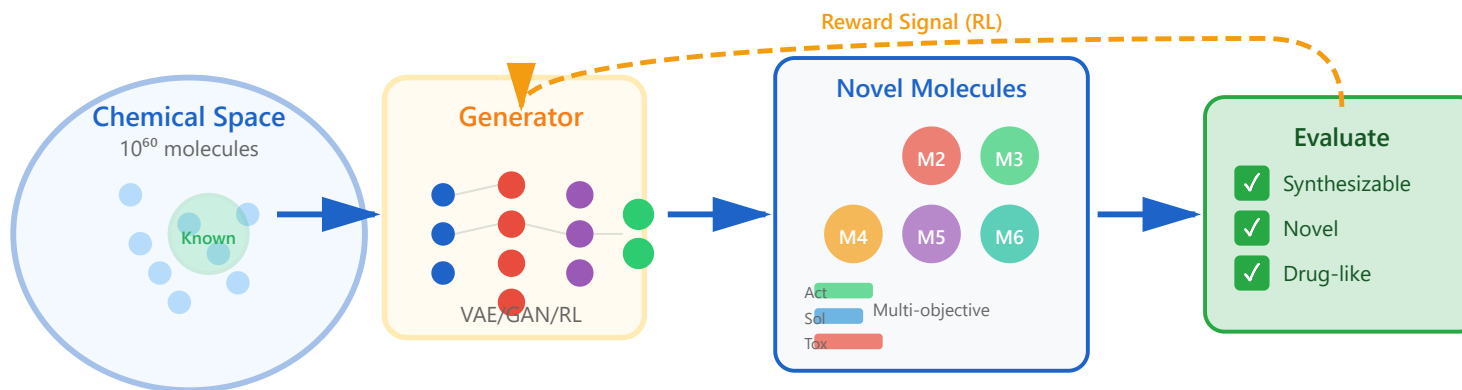
## Excretion (clearance)

Elimination pathway modeling

## Toxicity endpoints

Safety assessment

# De Novo Design



## Chemical space exploration

Novel compound generation

## Reinforcement learning

Goal-directed optimization

## VAE/GAN approaches

Generative architectures

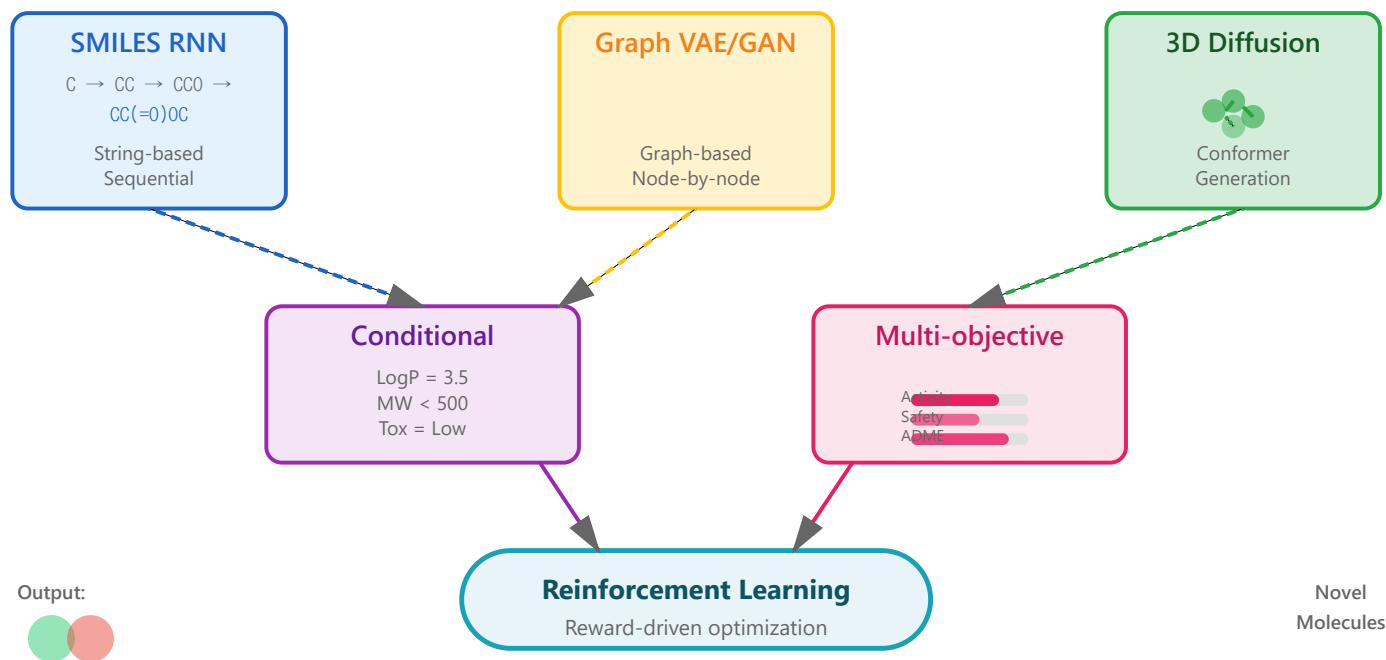
## Synthesizability

Chemical feasibility assessment

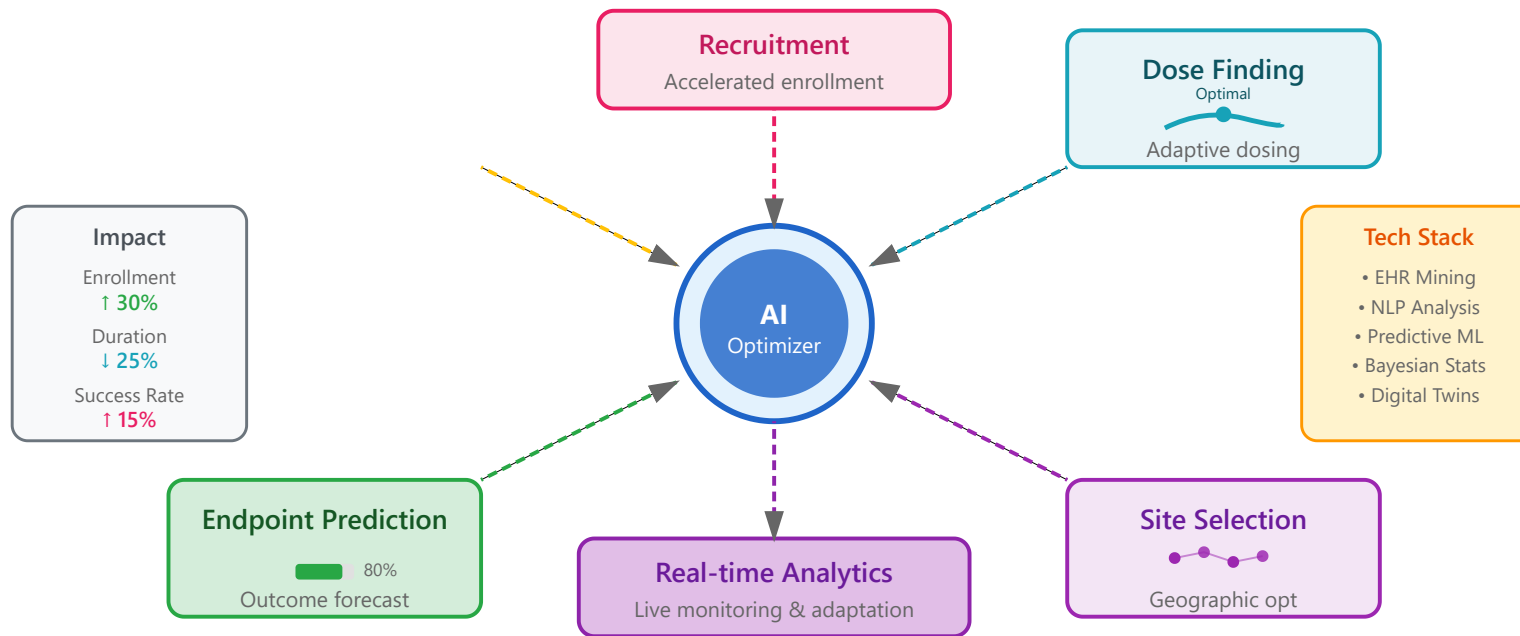
## Diversity metrics

Novelty quantification

# Generative Models



# Clinical Trial Optimization



# Pharmacovigilance

## Signal detection

Identifying safety signals

## Causality assessment

Determining drug-event relationships

## Risk-benefit analysis

Therapeutic decision support

## Literature mining

Automated safety surveillance

## Social media monitoring

Real-time safety signals

 HANDS-ON

## RDKit and DeepChem

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### Molecule manipulation

Loading, parsing, and modifying molecular structures

### Descriptor calculation

Computing physicochemical properties and fingerprints

### Model training

Building predictive models with DeepChem framework

### Scaffold splitting

Creating train/test splits based on molecular scaffolds

### Performance evaluation

Assessing model accuracy using appropriate metrics



```
# Example: RDKit & DeepChem workflow
from rdkit import Chem
import deepchem as dc

# Load molecules and compute descriptors
featurizer = dc.feat.CircularFingerprint()
loader = dc.data.CSVLoader(tasks=['activity'], featurizer=featurizer)
```

# Hands-on: RDKit and DeepChem

## Molecule manipulation

Reading and writing structures

## Descriptor calculation

Computing molecular features

## Model training

Building predictive models

## Scaffold splitting

Dataset partitioning strategies

## Performance evaluation

Metrics and validation

 HANDS-ON

## Molecular Generation

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### SMILES RNN

Recurrent neural networks for sequential generation

### Graph VAE

Variational autoencoders for graph-based generation

### Reinforcement learning

Policy-based optimization for desired properties

### Property optimization

Guiding generation toward specific target profiles

### Diversity analysis

Measuring chemical diversity in generated libraries

```
# Example: Generative model workflow
from rdkit import Chem
import torch

# Load pre-trained generative model
model = MolecularRNN.load('pretrained_model.pt')
generated_smiles = model.sample(n_molecules=100)
```

# Hands-on: Molecular Generation

## SMILES RNN

Recurrent neural network generation

## Graph VAE

Variational autoencoder for graphs

## Reinforcement learning

Policy-based optimization

## Property optimization

Multi-objective design

## Diversity analysis

Chemical space coverage

# Thank You

- Approved AI-discovered drugs
- Pipeline statistics & success rates
- Investment trends in AI drug discovery
  - Future outlook & opportunities

Introduction to Biomedical Datascience