

Drug Design

An Industrial Perspective

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OpenEye Scientific Software



Drug Discovery: A Little History

- “Folk” medicine
 - Digoxin (digitalis) from Foxglove plant
- Serendipity
 - Penicillin
 - Sildenafil (Viagra)
- Brute Force (Screen everything you can)
 - Most drugs developed during the 50s-70s
 - Many are derived from natural products
 - Example: Taxol (yew tree)
 - Companies often paid employees to bring back plants from exotic vacation destinations
- Design
 - HIV reverse transcriptase and HIV protease inhibitors



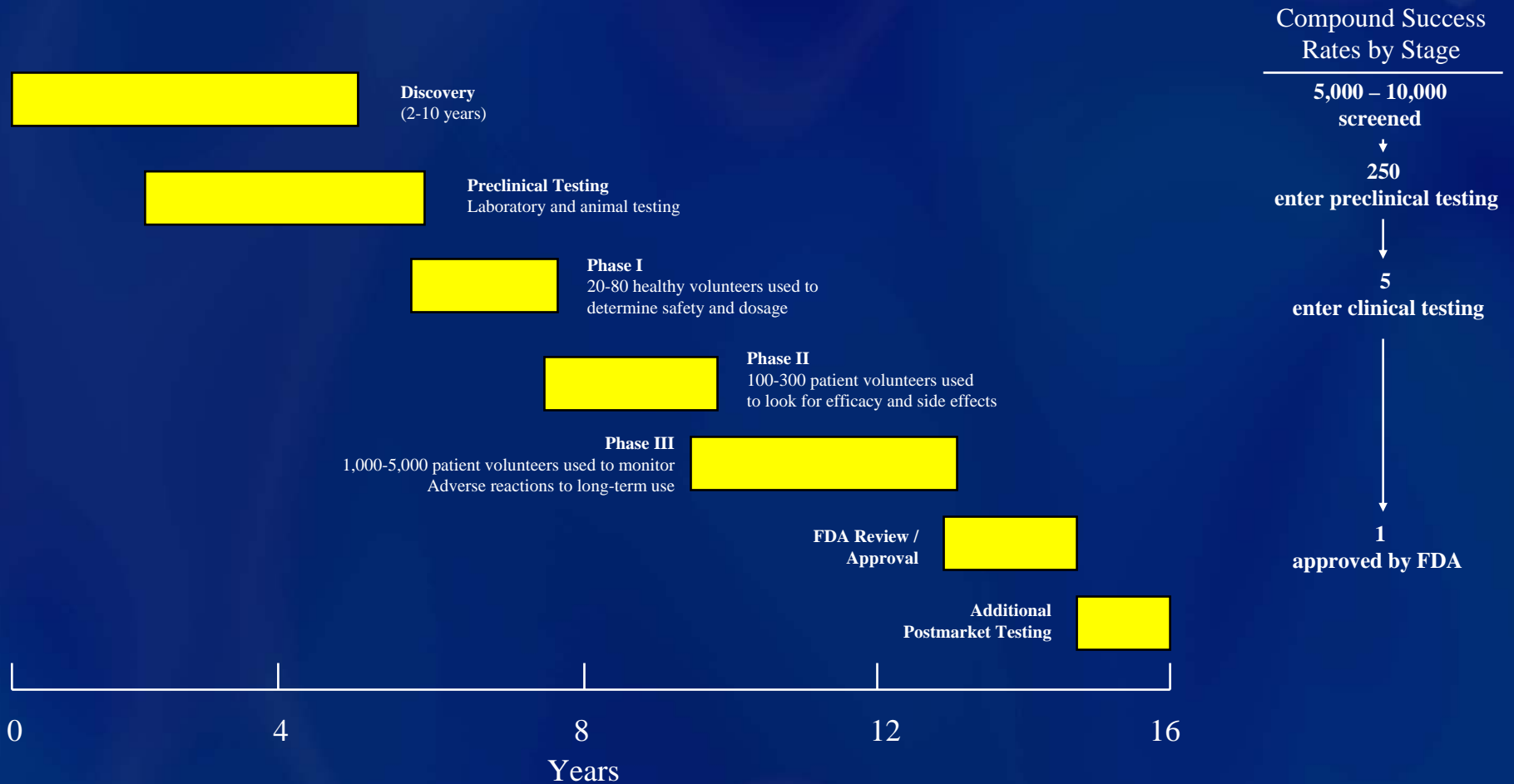
Drug Development: Overview

- Drug development is very expensive
 - Recent estimates put the cost of developing a single drug at ~\$800 million
 - Much of this cost is an amortization of all the other drugs that fail during development
- Drug development is very time consuming
 - Recent estimates put the time to market for a single drug in the range from 8-15 years
- Drug development is a high-risk game
 - For every 250 lead candidates only 1 will make it to FDA approval
 - Seven out of every ten drugs brought to market never generate enough revenue to recover the cost of development *

* PhRMA: Pharmaceutical Spending Facts and Figures 2004
<http://www.phrma.org/publications/publications/2005-08-30.1290.pdf>



Drug Development: Overview



Source: PhRMA, based on data from *Center for the Study of Drug Development*, Tufts University, 1995



Challenges to Making a Drug?

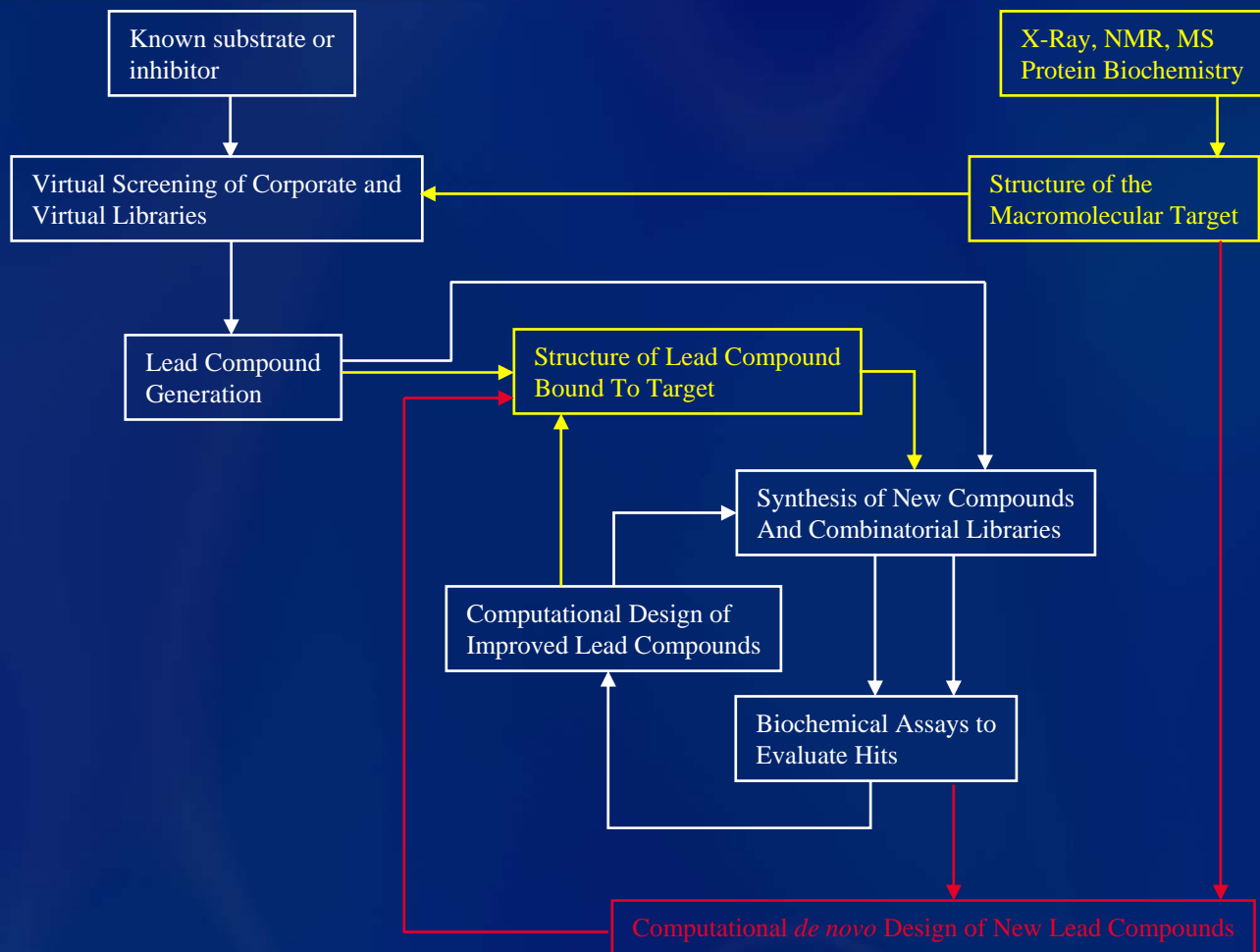
- Selection and determination of a feasible receptor target
- Finding an appropriate “key” for the “lock”
- Synthetic feasibility
- Biological effectiveness
- Bioavailability (ADME)
- Toxicity
- Patents
- Financial

Drug Discovery

- Disease / Target Selection (1 – 5 years)
- Lead Discovery (0.5 – 1 year)
- Lead Optimization (2 – 4 years)
- Preclinical Testing / Formulation (1 -2 years)



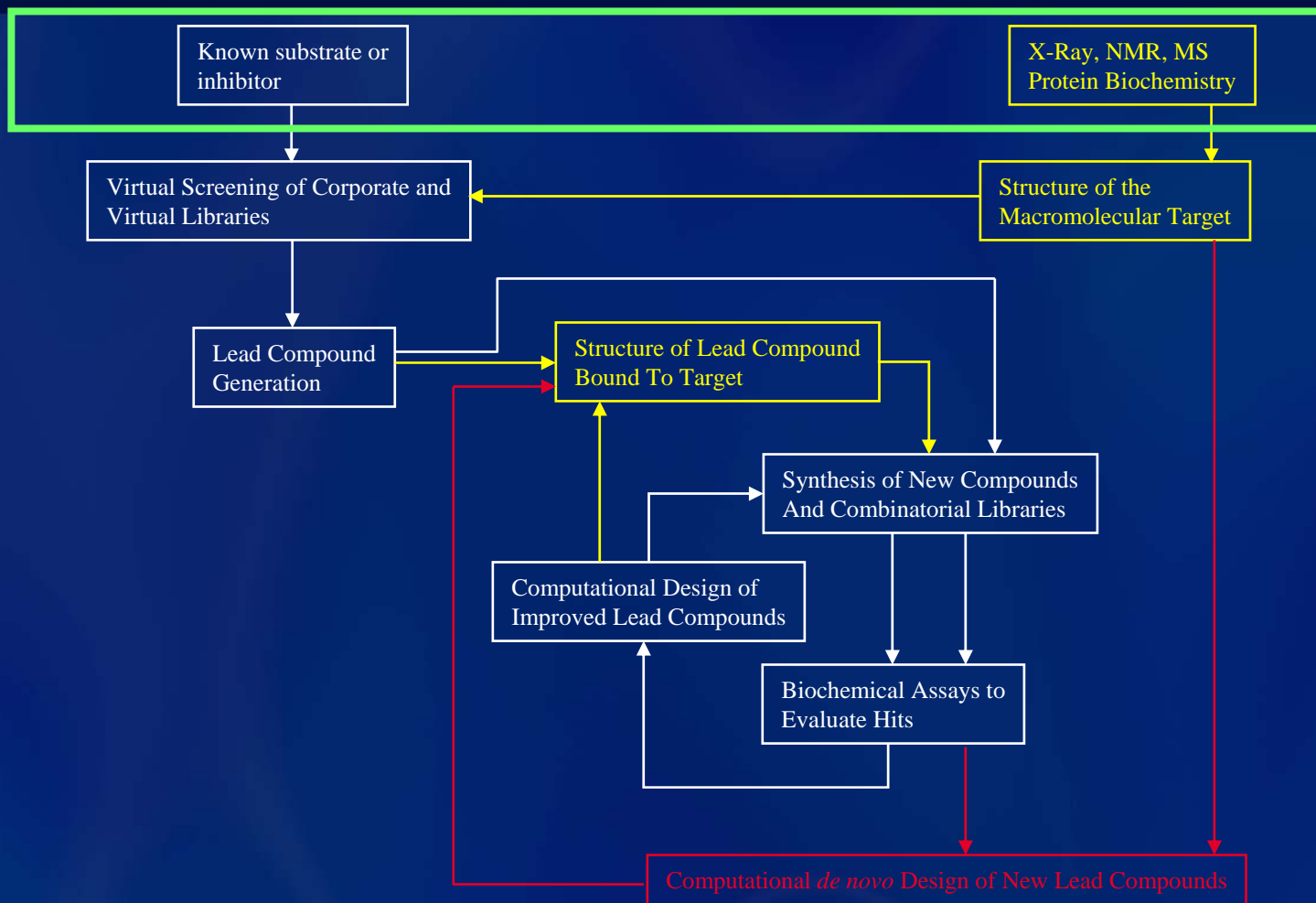
Drug Design: Process



Adapted from Figure 22.1: Bourne, Wessig, [Structural Bioinformatics](#).



Disease / Target Selection



Adapted from Figure 22.1: Bourne, Wessig, [Structural Bioinformatics](#).



Disease/Target Selection

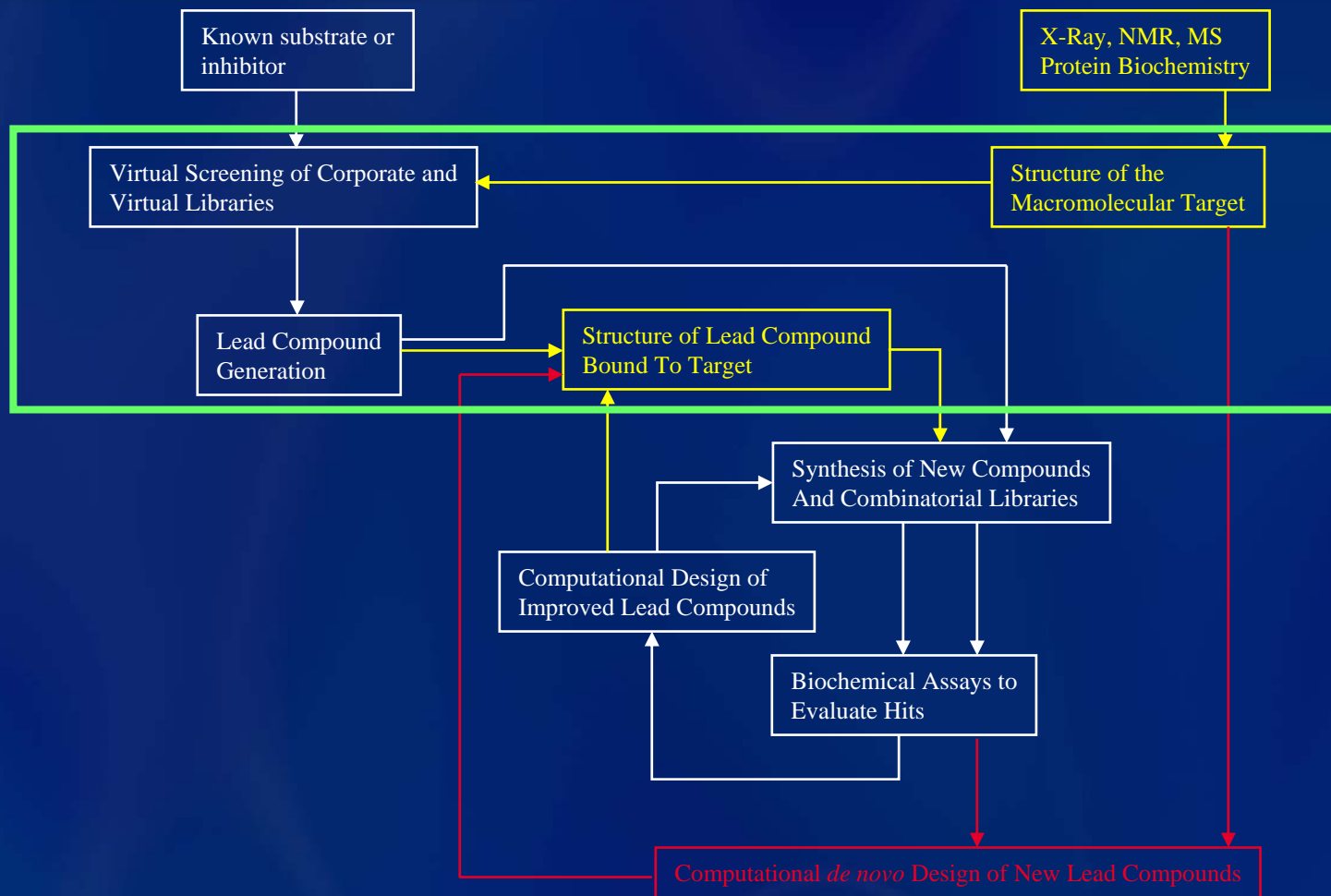
- Many financial considerations as many drugs never recoup investment costs.
- Best targets from a financial standpoint are chronic conditions with a high societal prevalence such as:
 - Asthma and Allergy
 - Diabetes
 - Hypertension
 - Joint Disease (arthritis)
 - Lipid Disorders (cholesterol and triglycerides)
 - Mental Disorders (depression, ADHD, etc.)
 - Ulcerative and Reflux Disease
- For rare and low prevalence diseases, pharmaceutical companies may apply for *orphan drug* status.
 - Government provides significant tax and marketing incentives
 - <http://www.fda.gov/orphan/designat/>



Disease/Target Selection: Cont'd

- In addition to financial considerations, there are many biological considerations:
 - The biological process has to be known such that a specific target protein can be selected
 - Must have the ability to produce and purify the protein of interest
 - Must be able to assay the *in vitro* activity of the protein in the setting of individual compounds
 - Must be able to scale-up the assay to allow for compound screening

Lead Discovery



Adapted from Figure 22.1: Bourne, Wessig, [Structural Bioinformatics](#).



Lead Discovery

- Iterative process of screening large numbers of compounds for biological activity
- Screening and testing compounds can be quite expensive
- Use “virtual screening” to reduce the number of compounds actually screened

Virtual Screening

- Process by which computational tools are used to reduce the number actual compounds screened
- Pharmaceutical companies and chemical vendors have very large databases of chemical compounds (millions) available for screening
- Goal is to reduce the number needed to screen while:
 - Increasing the probability of biological activity
 - Increasing the probability of oral absorption
 - Decreasing the probability of toxicity
 - Decreasing the number of false positives



Filtering: Lipinski's "Rule of Five"

- Seminal paper published in 1997 describing commonly found features of orally active drugs
- Features
 - ≤ 5 hydrogen bond donors
 - ≤ 10 hydrogen bond acceptors
 - Molecular weight < 500
 - $\text{LogP} < 5$
 - LogP is a measurement of the hydrophobicity / hydrophilicity of a given compound
 - Actual values are frequently not known
 - Frequently predicted using sum-of-fragment methods, neural networks trained on known structures with measured values, or a combination of these methods
 - Examples include xLogP and cLogP
- Widely extended to include many other properties to help identify "drug-like" molecules

Lipinski CA, Lombardo F, Dominy BW, Feeney PJ. "Experimental and computational approaches to estimate solubility and permeability in drug discovery and development settings." *Advanced Drug Delivery Reviews*. 15 January 1997; 23(1), pp 3-25.



Post-Filtering

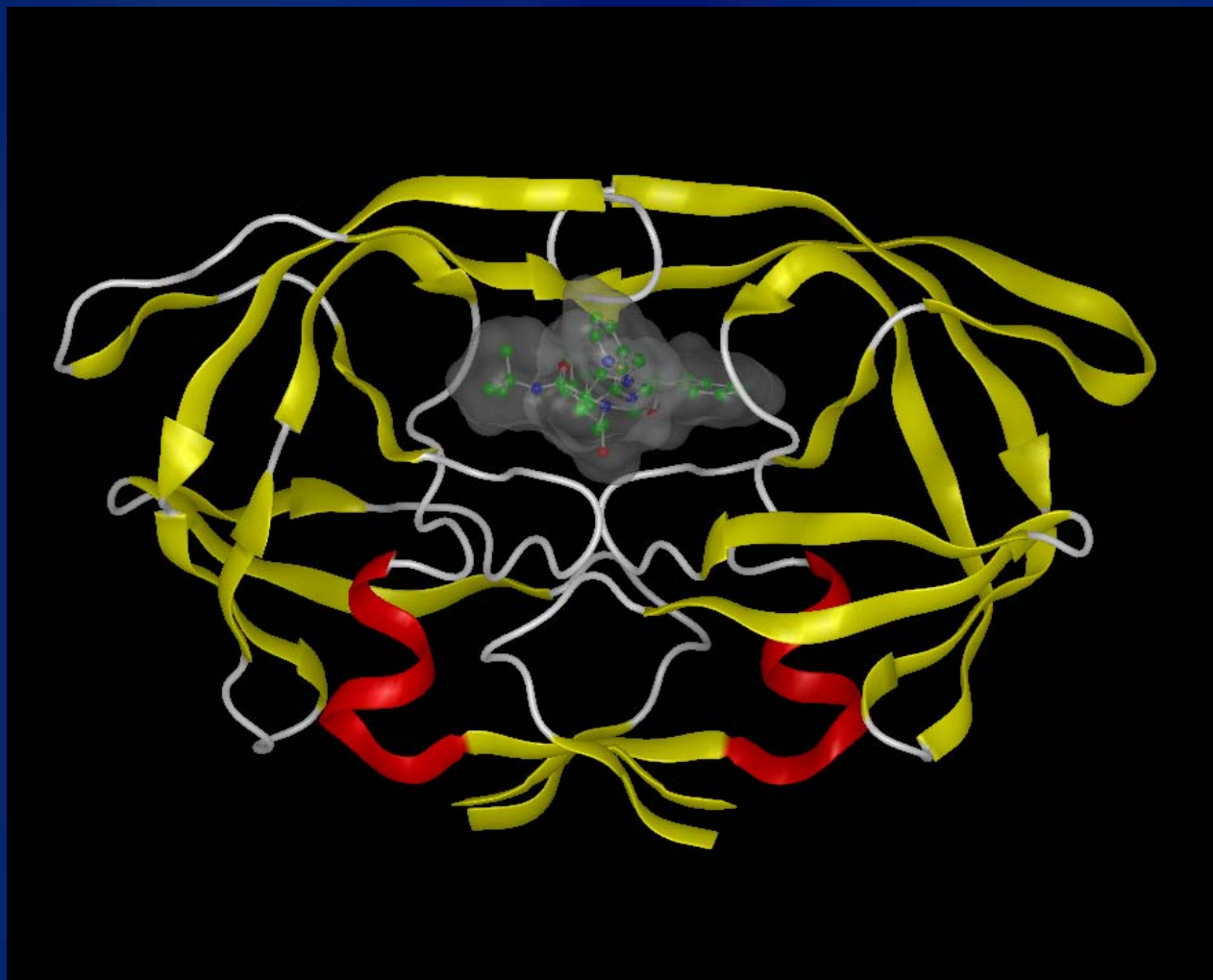
- **Structure-based design**
 - Requires 3D structure of target protein
 - Requires knowledge of location of ligand binding site
 - Example: HIV protease
- **Ligand-based design**
 - Requires known substrates or inhibitors of target
 - Example: HIV reverse transcriptase
- **Random screening**
 - Requires time, money, patience, and serendipity

Structure-Based Design Example: HIV Protease

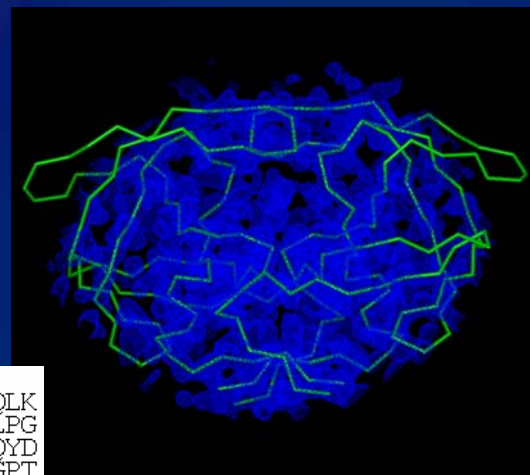
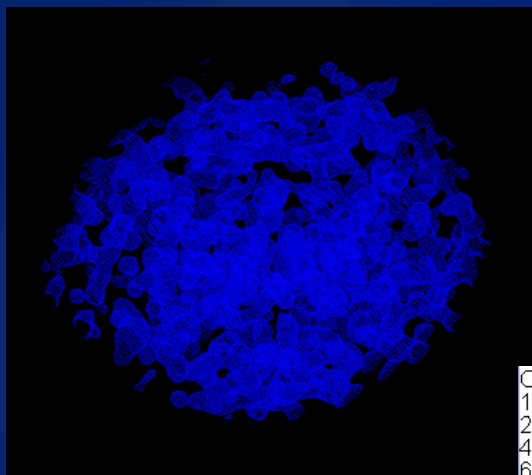
- HIV-1 protease was one among many potential proteins involved in the life-cycle of the HIV-1 virus
- HIV-1 protease was determined to be unique to HIV and absolutely required for replication
- As a result, there was a significant effort made to rapidly solve the crystal structure of this enzyme which helped elucidate the mechanism
- Using this information, a large number of protease inhibitors have been developed to date



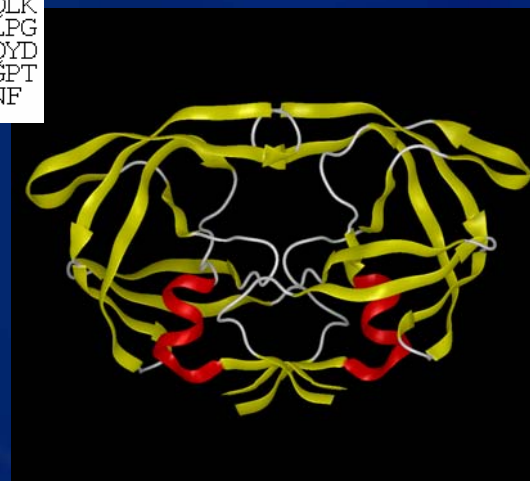
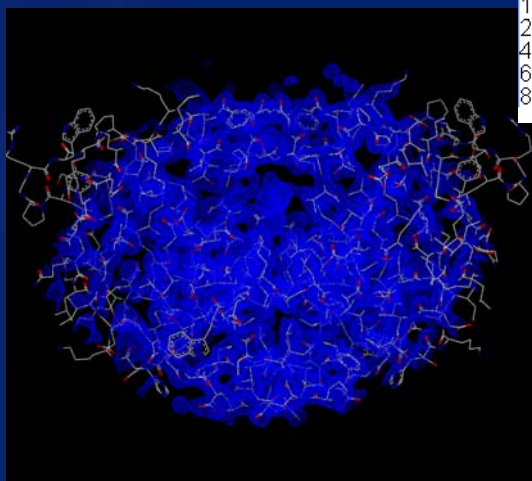
Structure-Based Design Example: HIV Protease Inhibitor (Indinavir)



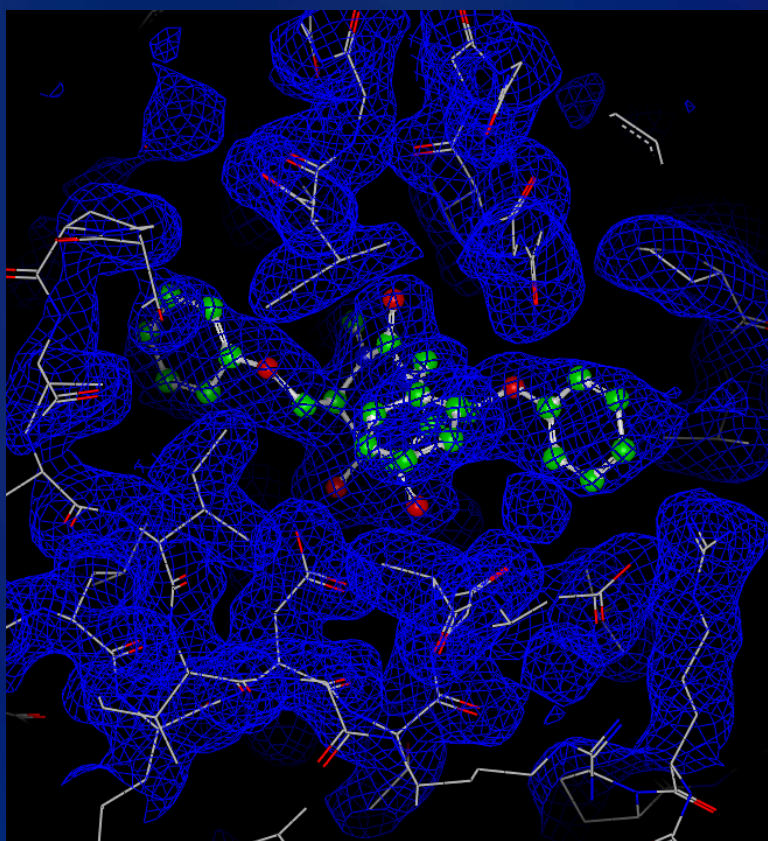
Structure-Based Design: Structure Determination



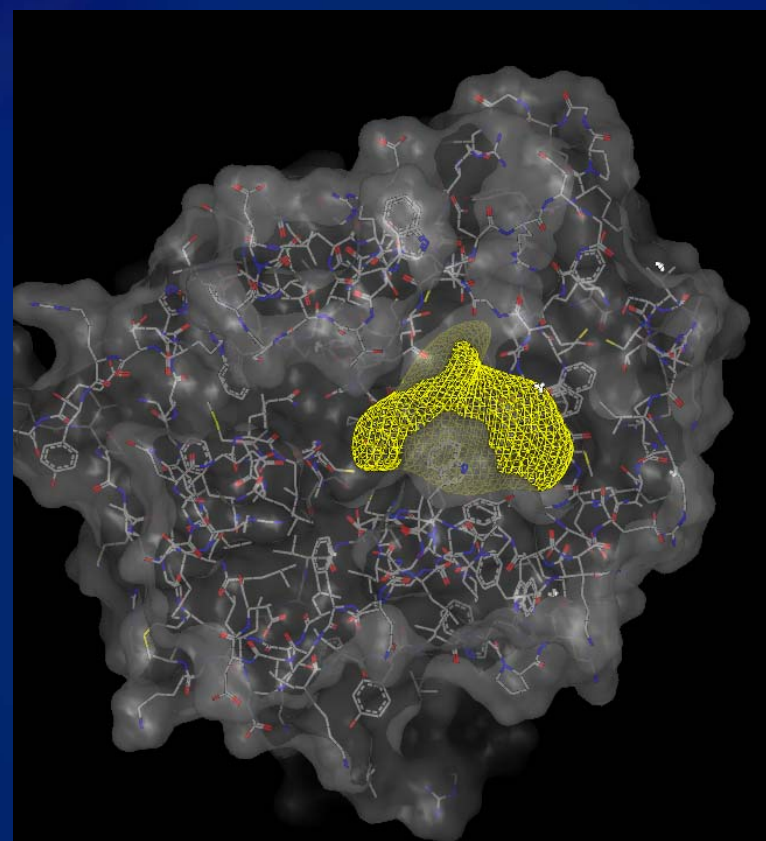
```
Chain A
1  PQITLWORPLVTIKIGGOLK
20 EALLDTGADDTVLEEMSLPG
40 RWKPKMIGGIGGFIKVROYD
60 QILIEICGHKAIGTVLVGPT
80 PVNIIGRNLLTQIGCTLNF
Chain B
1  PQITLWORPLVTIKIGGOLK
20 EALLDTGADDTVLEEMSLPG
40 RWKPKMIGGIGGFIKVROYD
60 QILIEICGHKAIGTVLVGPT
80 PVNIIGRNLLTQIGCTLNF
```



Structure-Based Design: Binding Site Determination



Co-crystallization of ligand



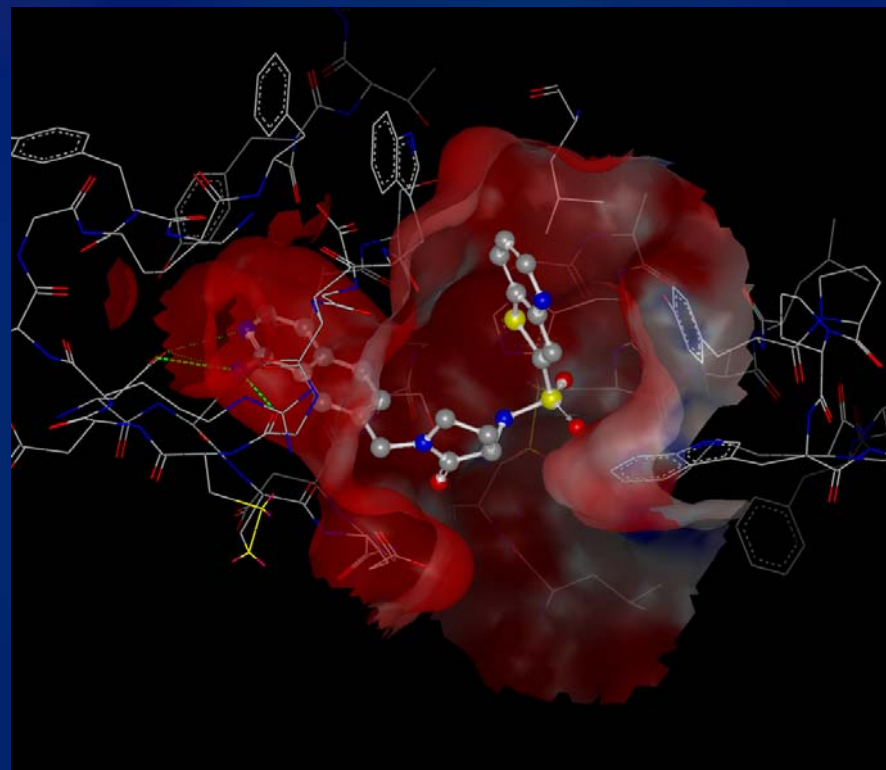
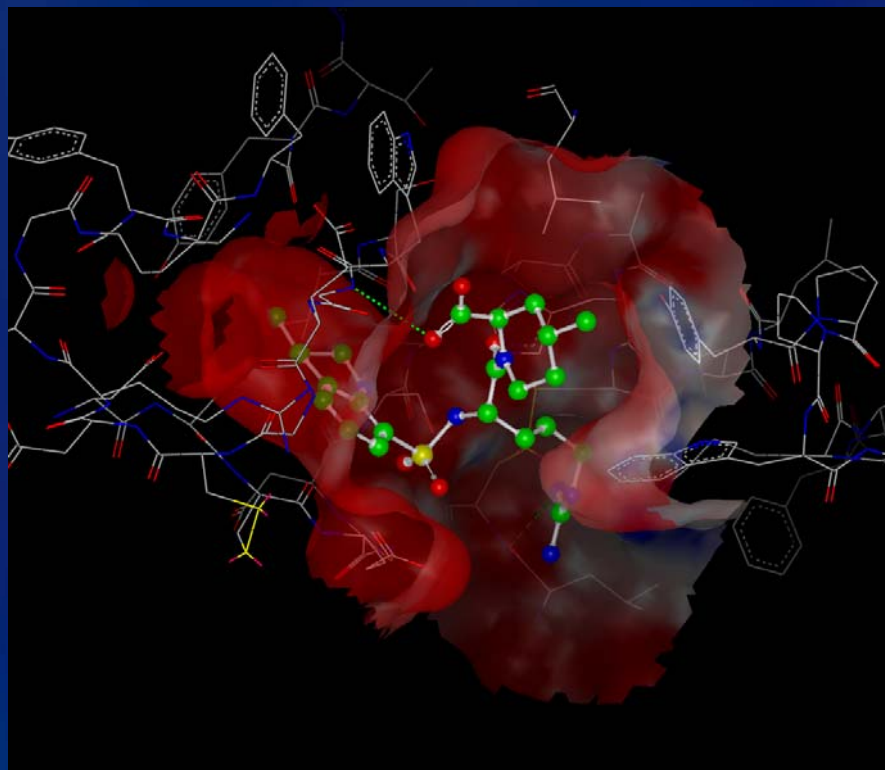
Computational methods

Structure-Based Design: Docking

- Docking is a widely used methodology for selecting compounds for initial screening
- It is now possible to systematically search all potential docking poses of a given compound in a reasonable period of time
- Docking's primary limitations:
 - Scoring functions
 - Protein flexibility



Docking: Continued



Ligand-Based Design

- Select potential leads based on other compounds that are known bind to the protein
- Natural substrate
 - Looking for compounds that bind in a similar fashion thus preventing binding of the natural substrate
- Known inhibitors / competitors' drugs
 - Looking for compounds that bind in a similar fashion but with stronger binding capability and sufficiently different structure / chemistry to avoid patent infringement

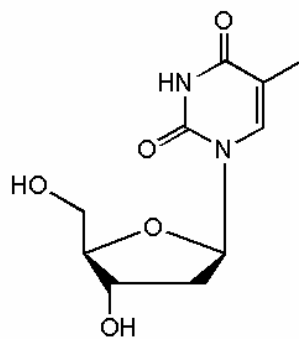
Ligand-Based Design Example: HIV Reverse Transcriptase

- HIV is a retrovirus, meaning that it's genetic material is RNA (not DNA) and that it contains an enzyme *reverse transcriptase* which converts RNA to DNA
- When HIV infects a human T-cell, it's RNA is converted to DNA (via *reverse transcriptase*) which is then incorporated into the genetic material of the cell enabling continued reproduction of the virus
- Given this knowledge, *reverse transcriptase* was an obvious target for drug design

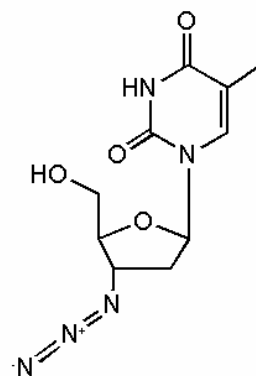
Ligand-Based Design Example: HIV Reverse Transcriptase Cont'd

- As *reverse transcriptase* produces DNA, its substrates were obvious: the four DNA nucleosides – adenosine, cytidine, guanosine, thymidine
- It was hypothesized that creation of an analogue of one of these nucleosides might inhibit *reverse transcriptase* thus preventing viral reproduction

Ligand-Based Design Example: HIV Reverse Transcriptase Cont'd



Thymidine



AZT

Ligand-Based Design Methodologies

- 2D Similarity
 - Fingerprint-based
- 3D Similarity
 - Pharmacophores
 - Shape
 - Electrostatics



2D Similarity: Fingerprints

- A molecular descriptor is a numerical or binary value for a given structure
 - 1D: logP, molecular weight, num atoms
 - 2D: graph-based, functional groups
- A fingerprint is a collection of molecular descriptors for a given molecule
- Can be time consuming to calculate, so are often calculated once and stored in a database

2D Similarity: Fingerprints

- Fingerprint searching is fast with binary fingerprints

| | | | | | | | | | | | | | | | | | |
|---|---|---|---|---|---|---|---|---|---|---|---|---|-----|---|---|---|---|
| A | 0 | 0 | 1 | 0 | 0 | 1 | 1 | 0 | 1 | 0 | 0 | 1 | ... | 0 | 1 | 0 | 1 |
| B | 1 | 0 | 1 | 0 | 0 | 0 | 1 | 0 | 0 | 1 | 0 | 1 | ... | 1 | 1 | 1 | 0 |

- Similarity is assessed by calculating a Tanimoto coefficient

$$T(a,b) = \frac{N(a \cap b)}{N(a) + N(b) - N(a \cap b)}$$

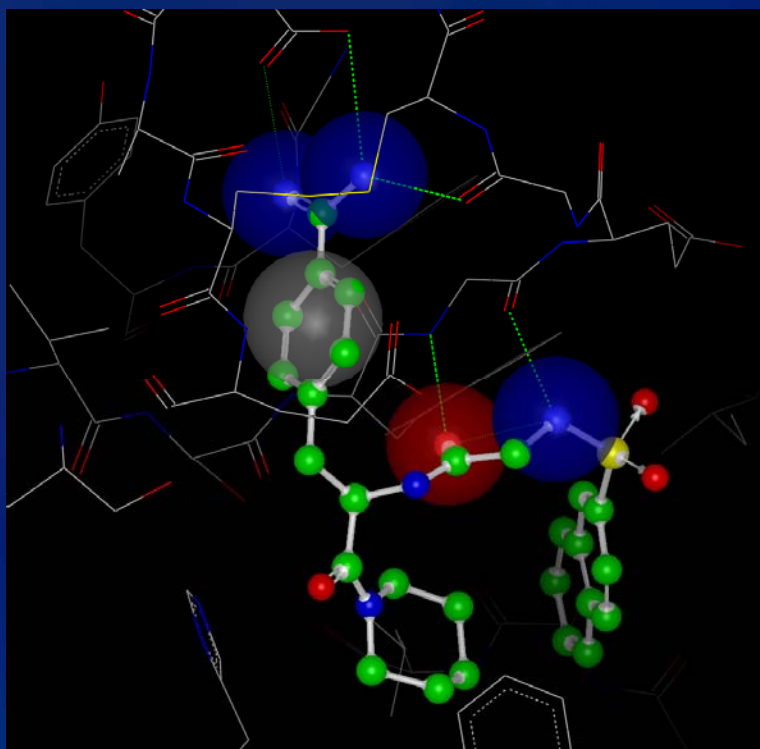
- Tanimoto has a range [0,1]

3D Similarity: Pharmacophores

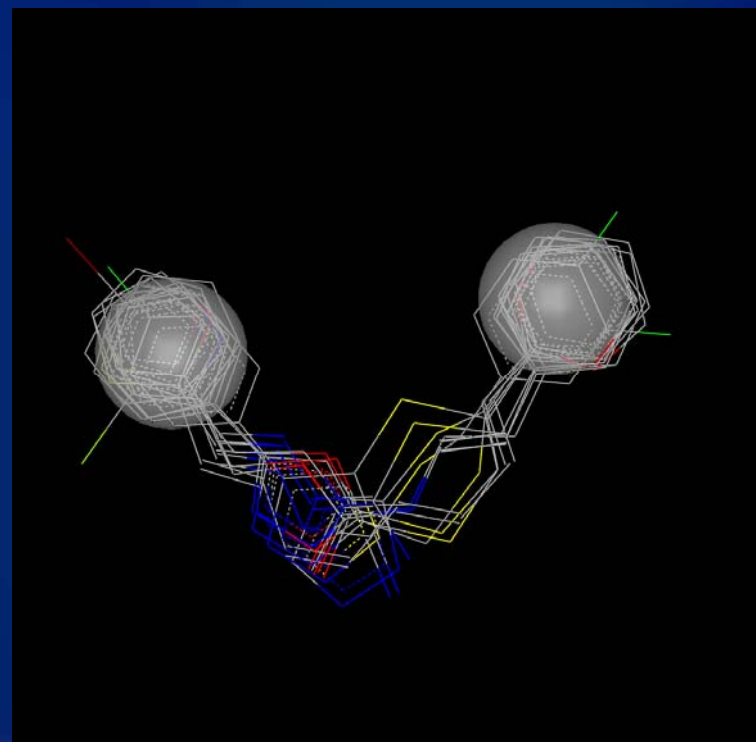
- Traditional definition is the minimum functionality a molecule has to contain in order to exhibit activity
- Typically defined in terms of atoms or centers which can interact with a receptor and are categorized into six types:
 - Hydrogen bond acceptors
 - Hydrogen bond donors
 - Anion
 - Cations
 - Aromatic ring centers
 - Hydrophobic ring centers

3D Similarity: Pharmacophores

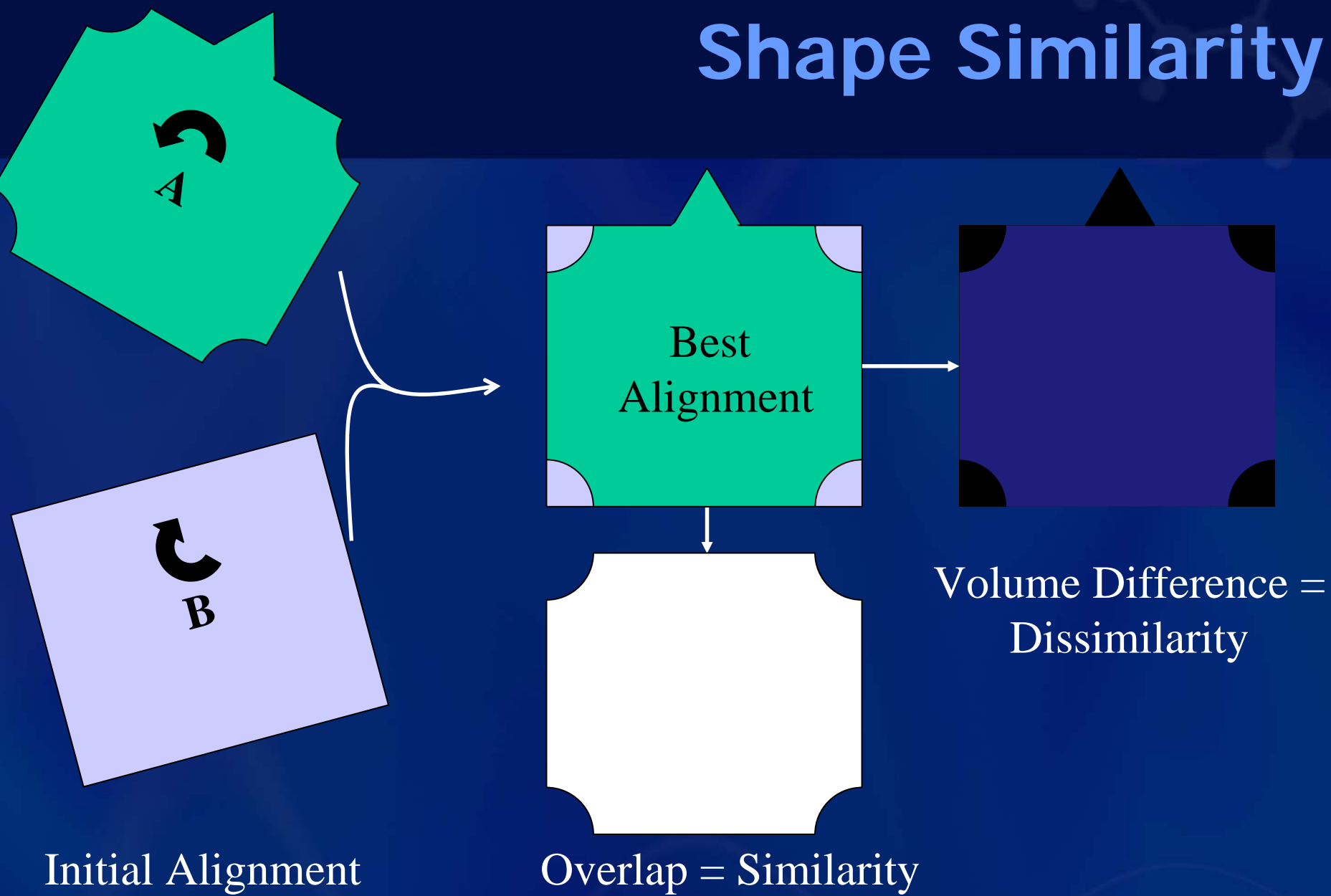
Receptor based



Ligand based



Shape Similarity



Initial Alignment

Best Alignment

Volume Difference =
Dissimilarity

Overlap = Similarity

Shape: Mathematically...

$$\Delta^2 = \int_{\text{all space}} (\chi_1 - \chi_2)^2 dv$$

$\chi_1 = \textit{defines shape 1}$

$\chi_2 = \textit{defines shape 2}$

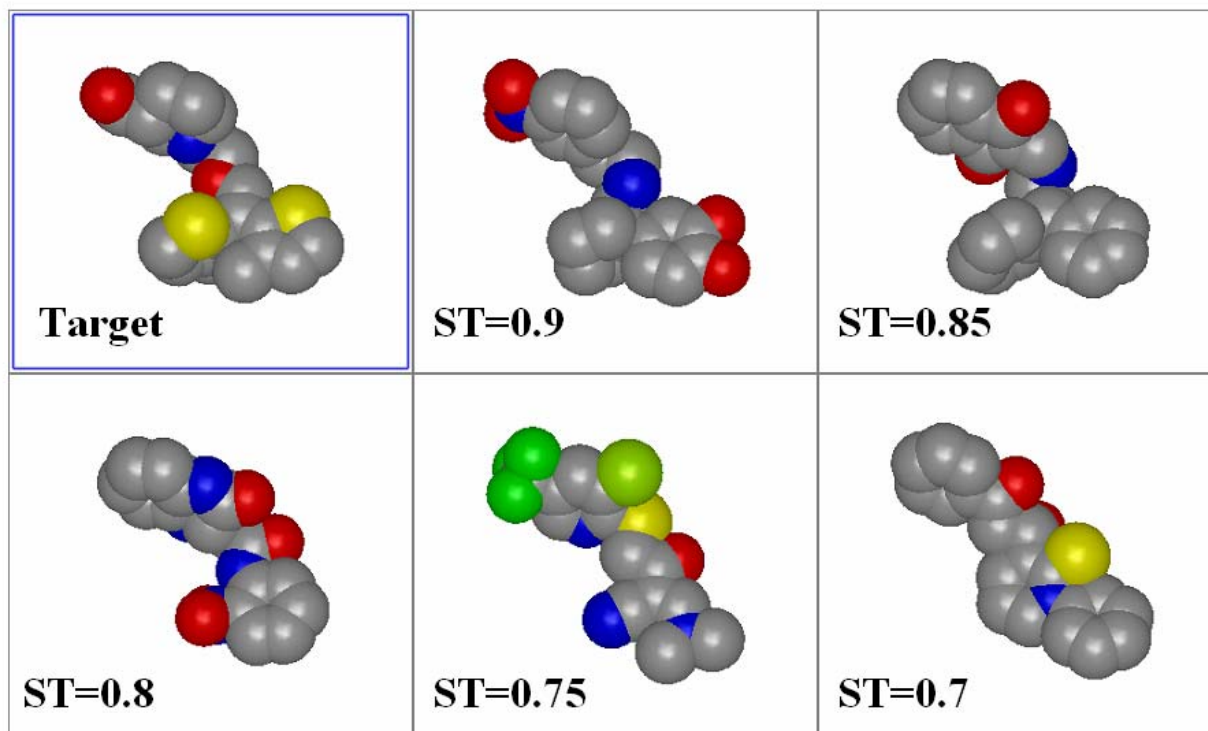
$\Delta = \textit{shape distance}$

Shape: Tanimoto (ST)

$$\frac{\text{Overlap}(A, B)}{\text{Overlap}(A, A) + \text{Overlap}(B, B) - \text{Overlap}(A, B)}$$

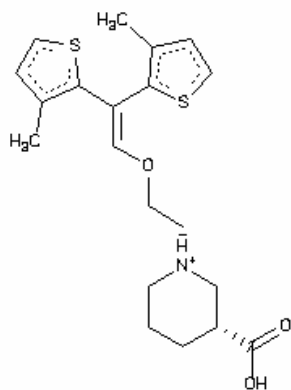
- Larger Tanimoto = More similar = Better
- Smaller Tanimoto = Less similar = Worse
- Range = [0,1]
- Value understood, e.g. $T > 0.75$ = Shape similar

Shape Similarity

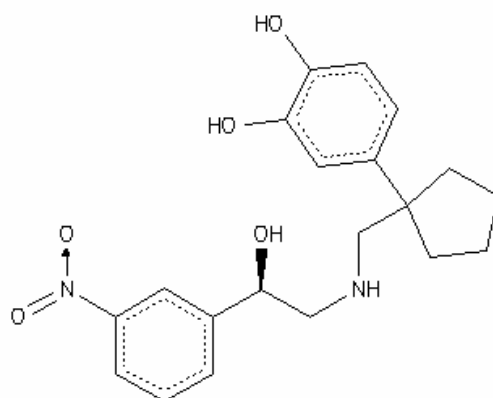


Chemical Dissimilarity

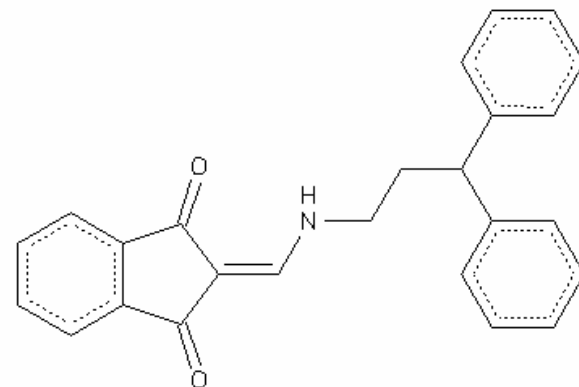
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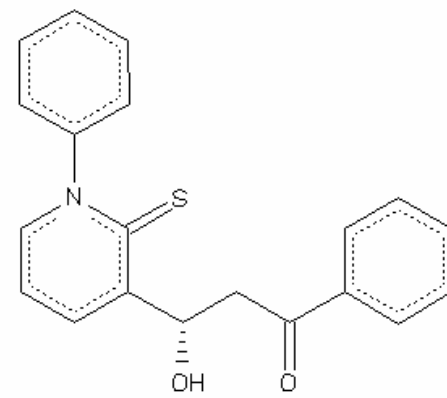
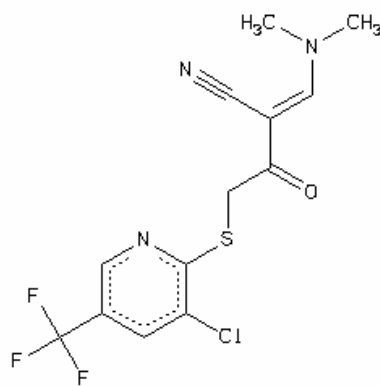
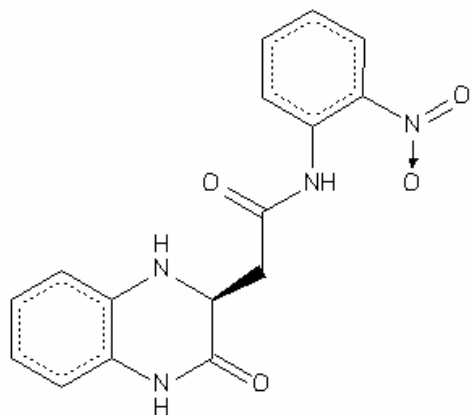
chembridge.03.204741_858



ibs.99.76040_892



asinx.01.155513::asinx.02.134 bionet.00.24579::bionet.01.2857 specs.03.189350::specs.03.2697:



Docking: Revisited

- How does ligand-based design compare with structure-based design?
- Compared OpenEye's ligand shape-based search tool (ROCS) with common docking programs
- Docking results obtained from published papers

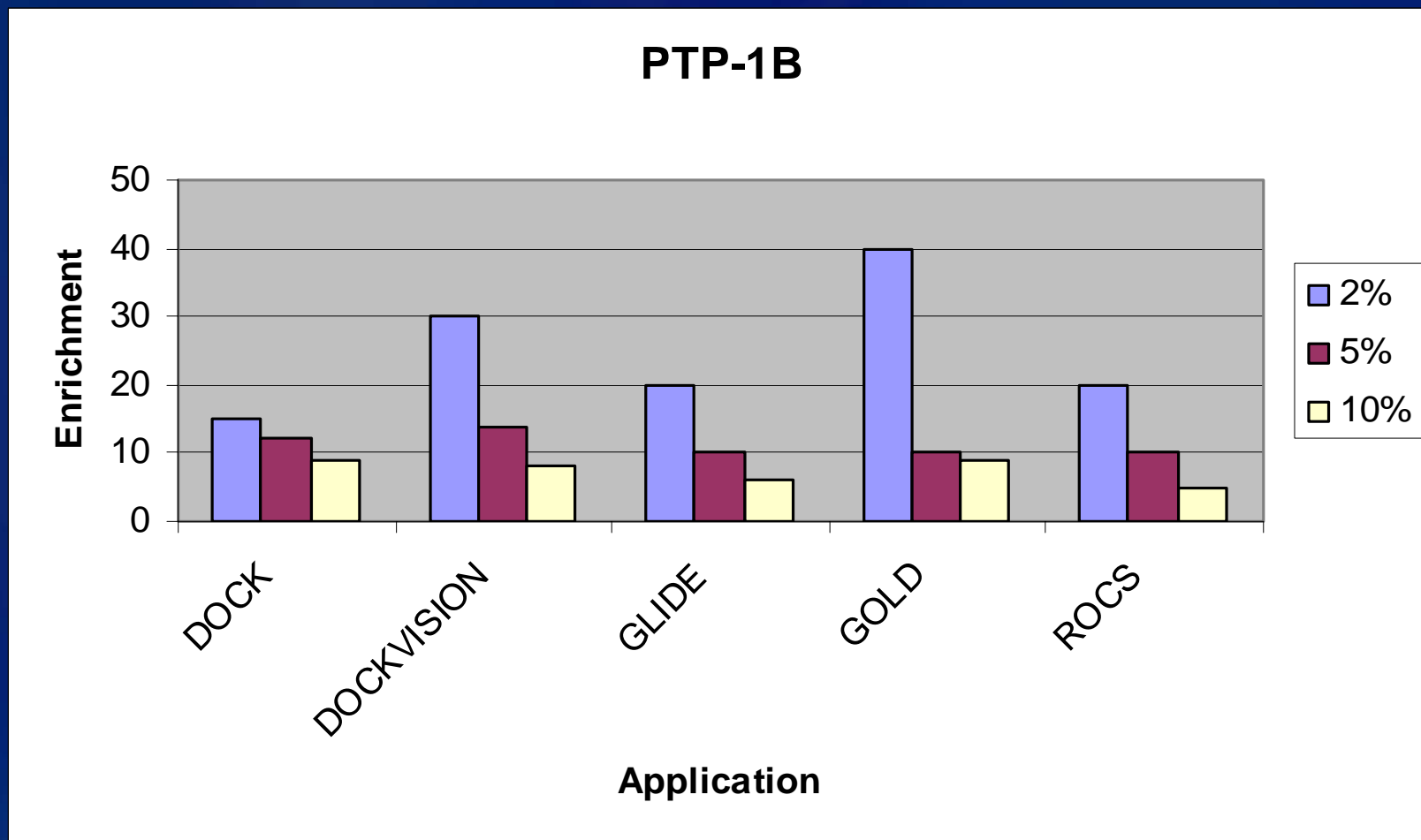


Docking: First Study

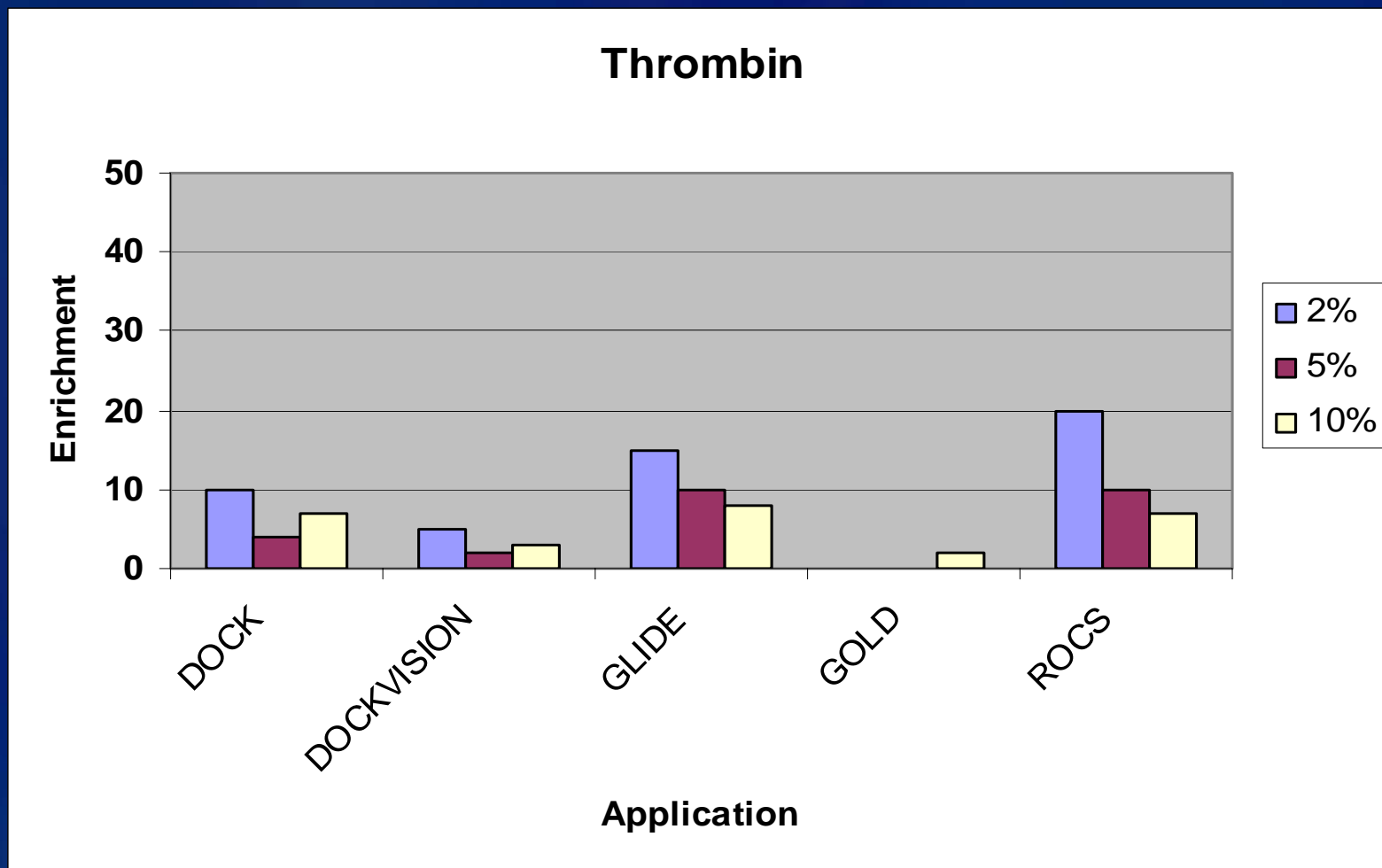
- Cummings *et al.*, *J. Med. Chem.*, **2005**, *48*, 962
 - compared 4 programs on 3 public and 2 proprietary datasets
 - DOCK, Dockvision, GOLD, GLIDE
 - HIV-PR, PTP-1B, thrombin
 - Decoys randomly selected from MDDR
 - Decoys and actives publicly available



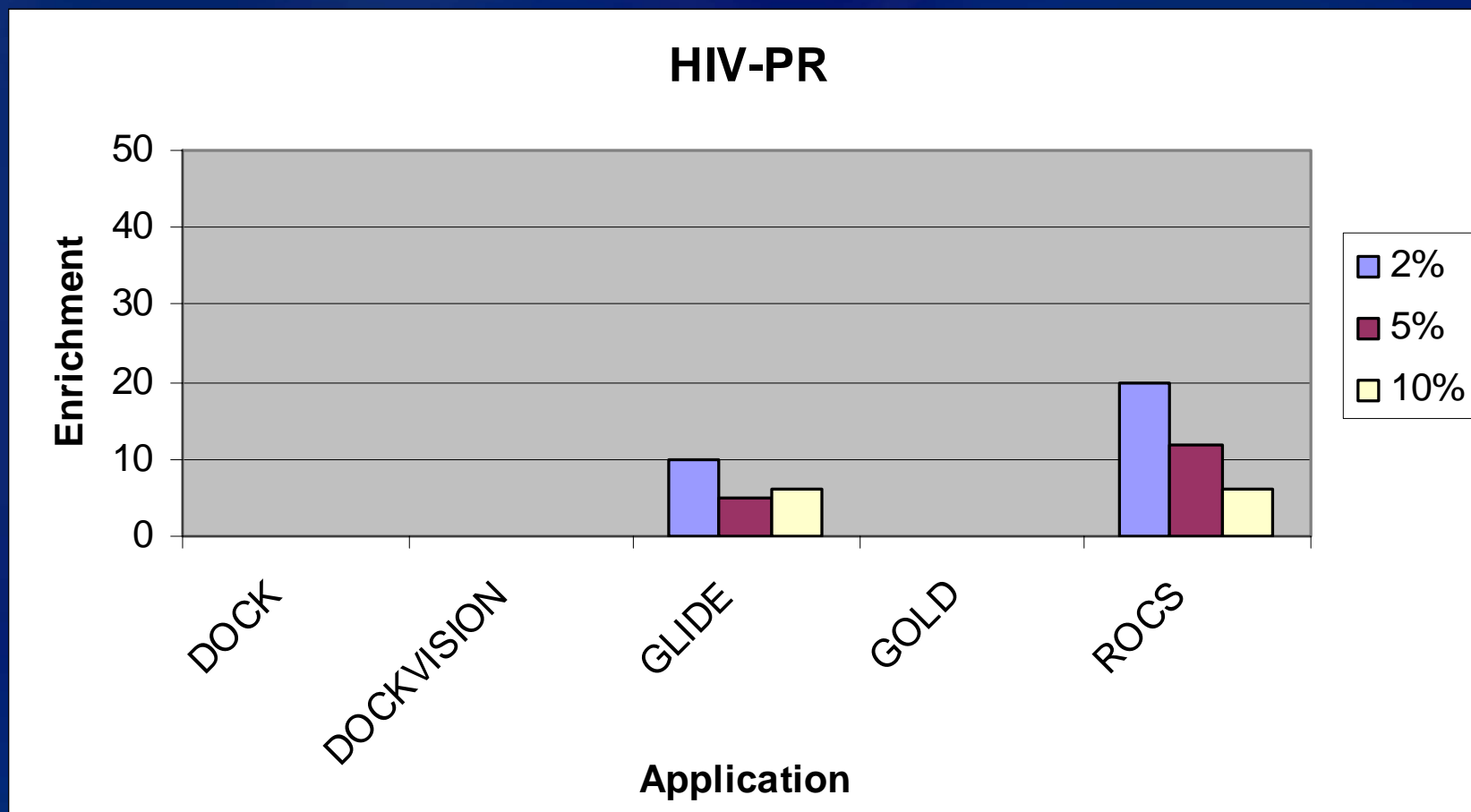
PTP-1B



Thrombin



HIV-Protease



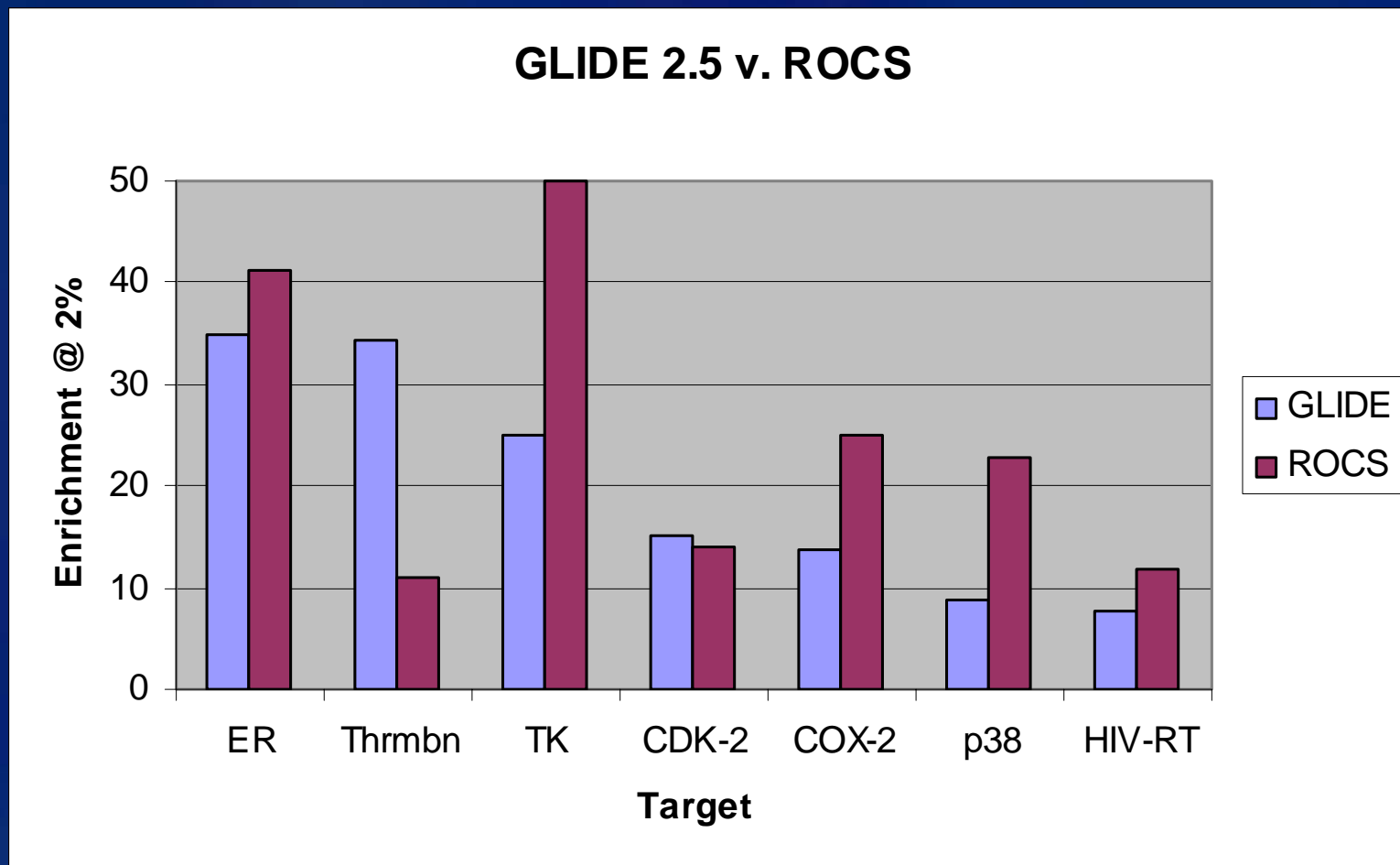
First Study: Conclusion

- Docking tools have inconsistent performance
- Ligand-based shape techniques shows consistent performance

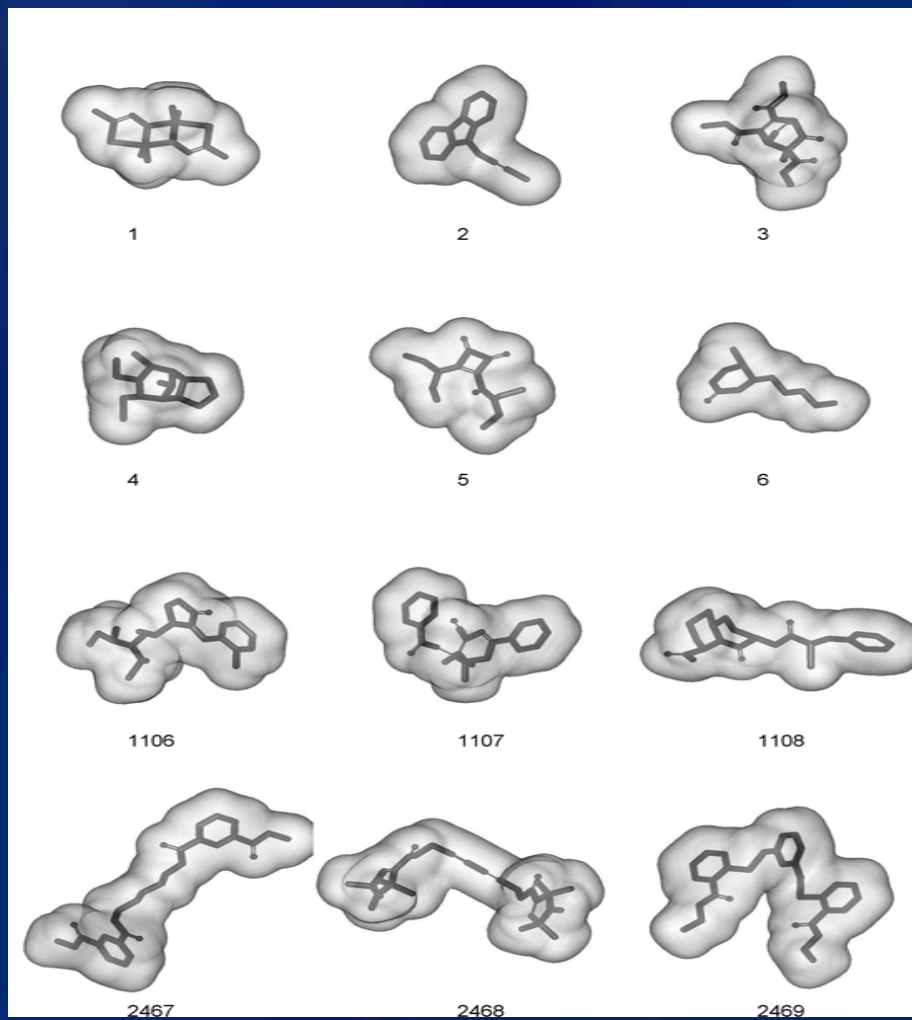
Docking: Second Study

- Halgren *et al.*, *J. Med. Chem.*, **47**, 1750 (2004)
 - GLIDE 2.5
 - Thymidine kinase, ER, CDK-2, p38, HIV-PR, thrombin, thermolysin, COX-2, HIV-RT
 - Decoys chosen to match a property profile
- No compounds publicly available

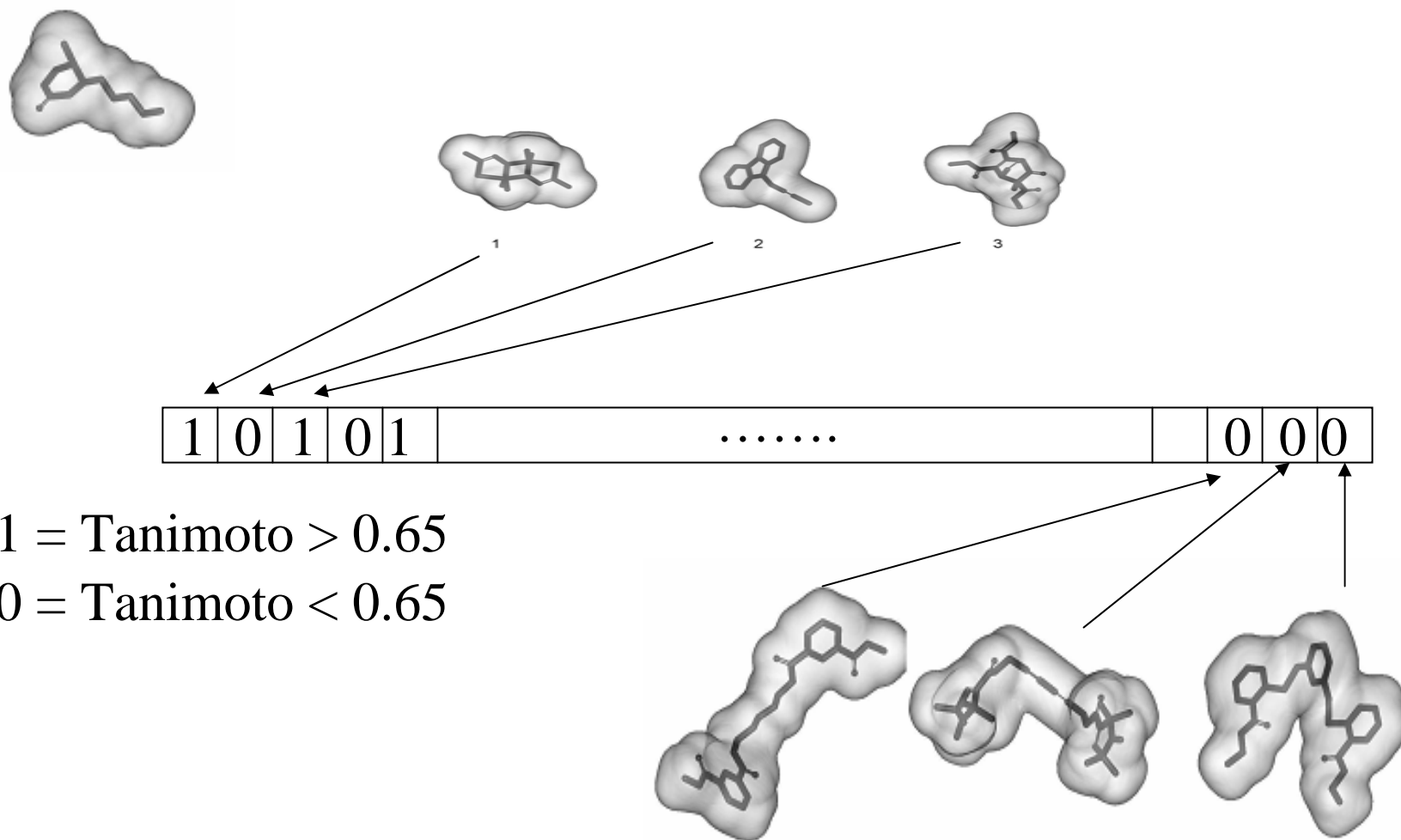
ROCS & GLIDE



Representative Shapes



Binary Shape Fingerprints



1 = Tanimoto > 0.65

0 = Tanimoto < 0.65

Electrostatic Similarity

$$\Delta^2 = \int_{\text{all space}} (\chi_1 - \chi_2)^2 dv$$

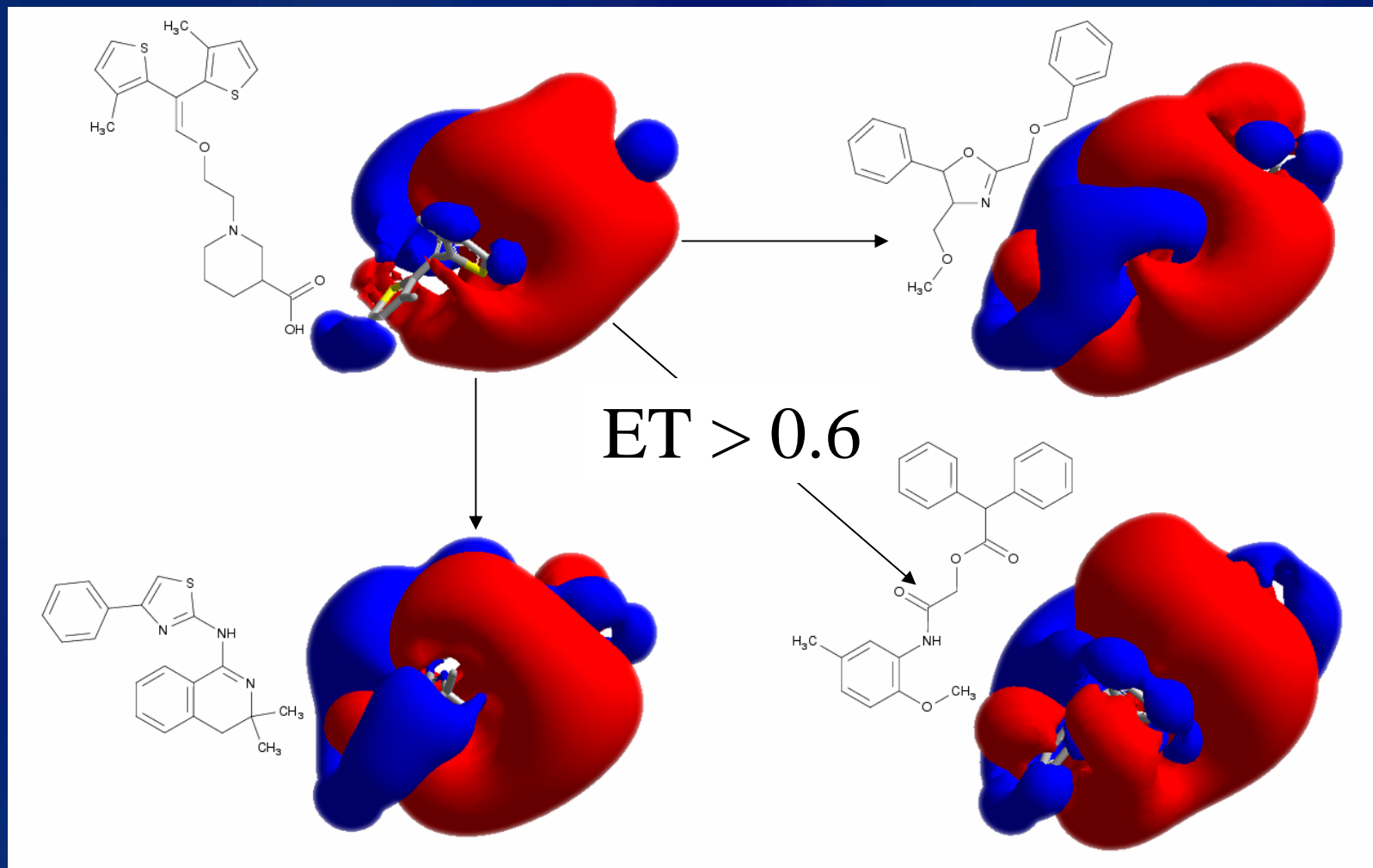
$$\Theta^2 = \int_{\text{all space}} (\phi_1 - \phi_2)^2 dv$$

$\phi_1 = \text{defines potential field 1}$

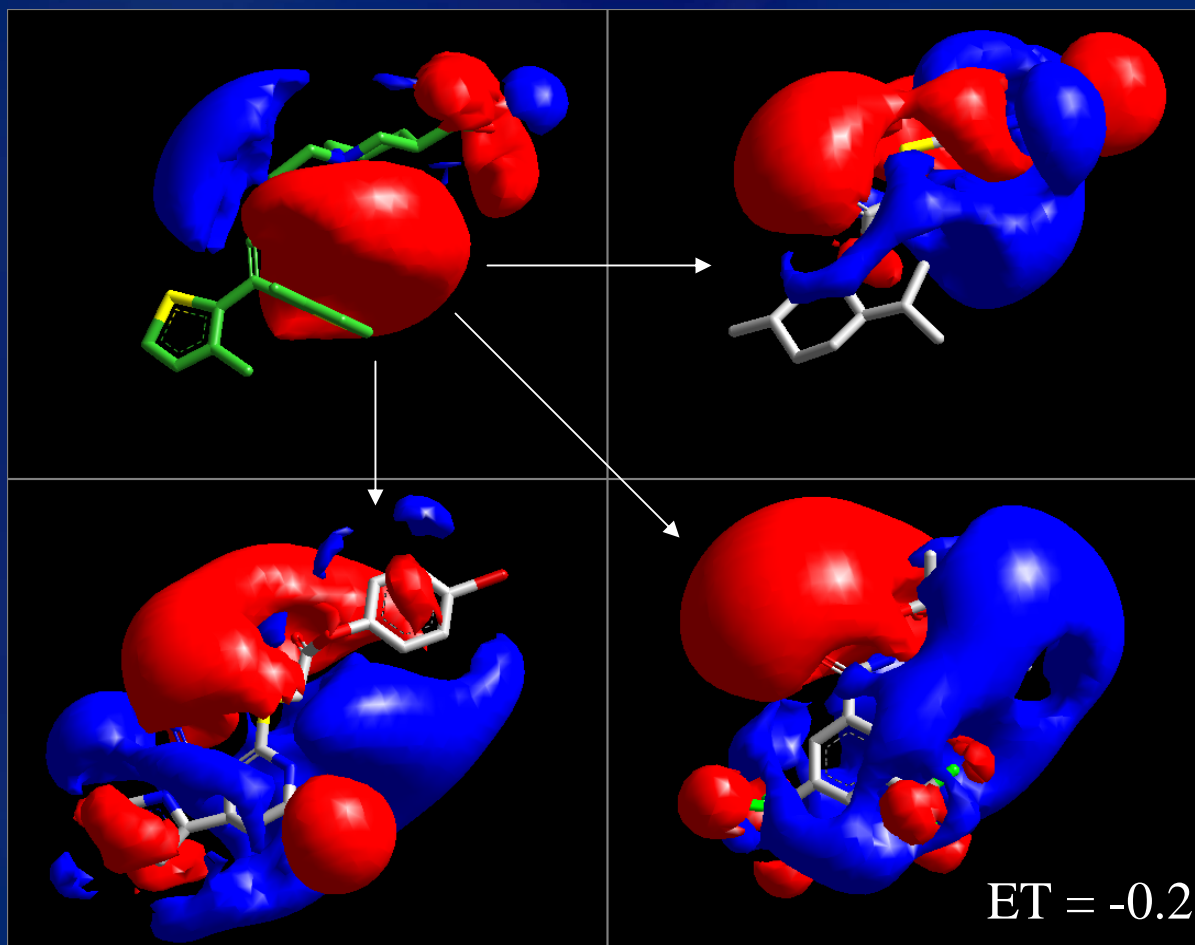
$\phi_2 = \text{defines potential field 2}$

$\Theta = \text{Electrostatic Distance}$

Electrostatic Tanimoto: Good



Electrostatic Tanimoto: Bad



Lead Optimization

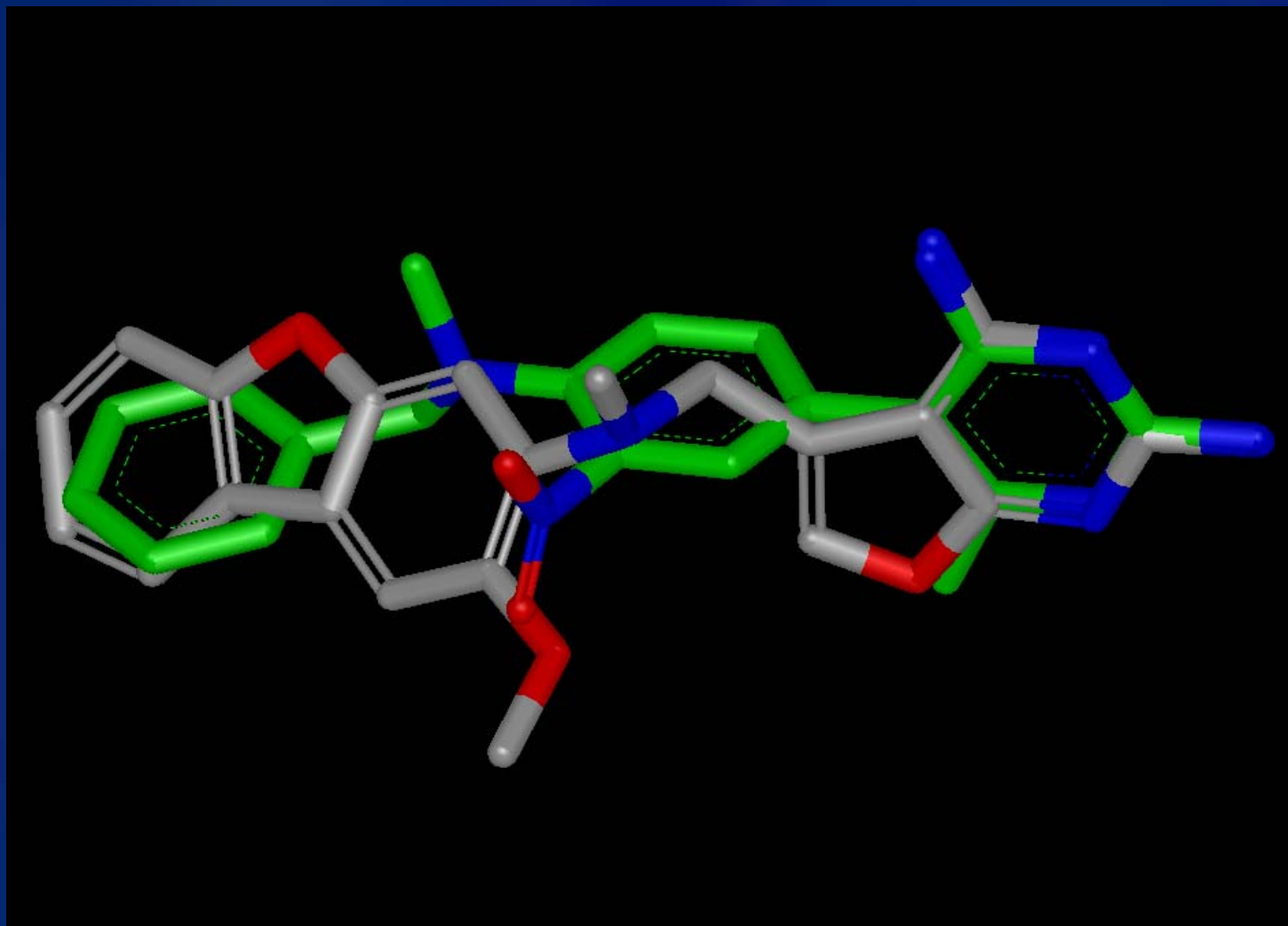
- Leads discovered in the first stage are rarely potent enough to be considered drug candidates
- Initial candidates must be optimized for binding affinity and other “drug-like” properties
- Medicinal chemists play a large role in making optimizations based on their experience
- Molecular graphics tools are particularly important in structure based projects to help maximize receptor interactions



Lead Optimization

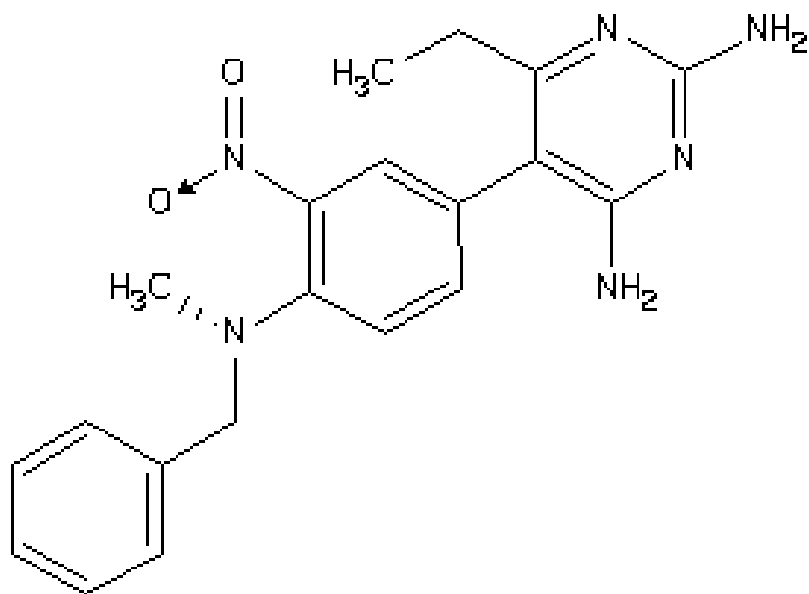
- Look for compounds with same shape (isosteres) but with different chemical and/or electrostatic profiles
- Fix the important parts, but change the rest
- Look for un- or under-utilized regions in the receptor

Scaffold Hopping: Shape



Scaffold Hopping: Shape

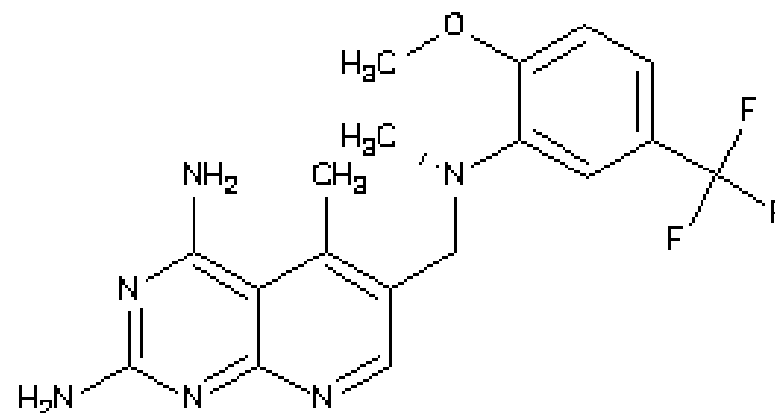
IC50 1.6 μ M



Query



IC50 93 nM

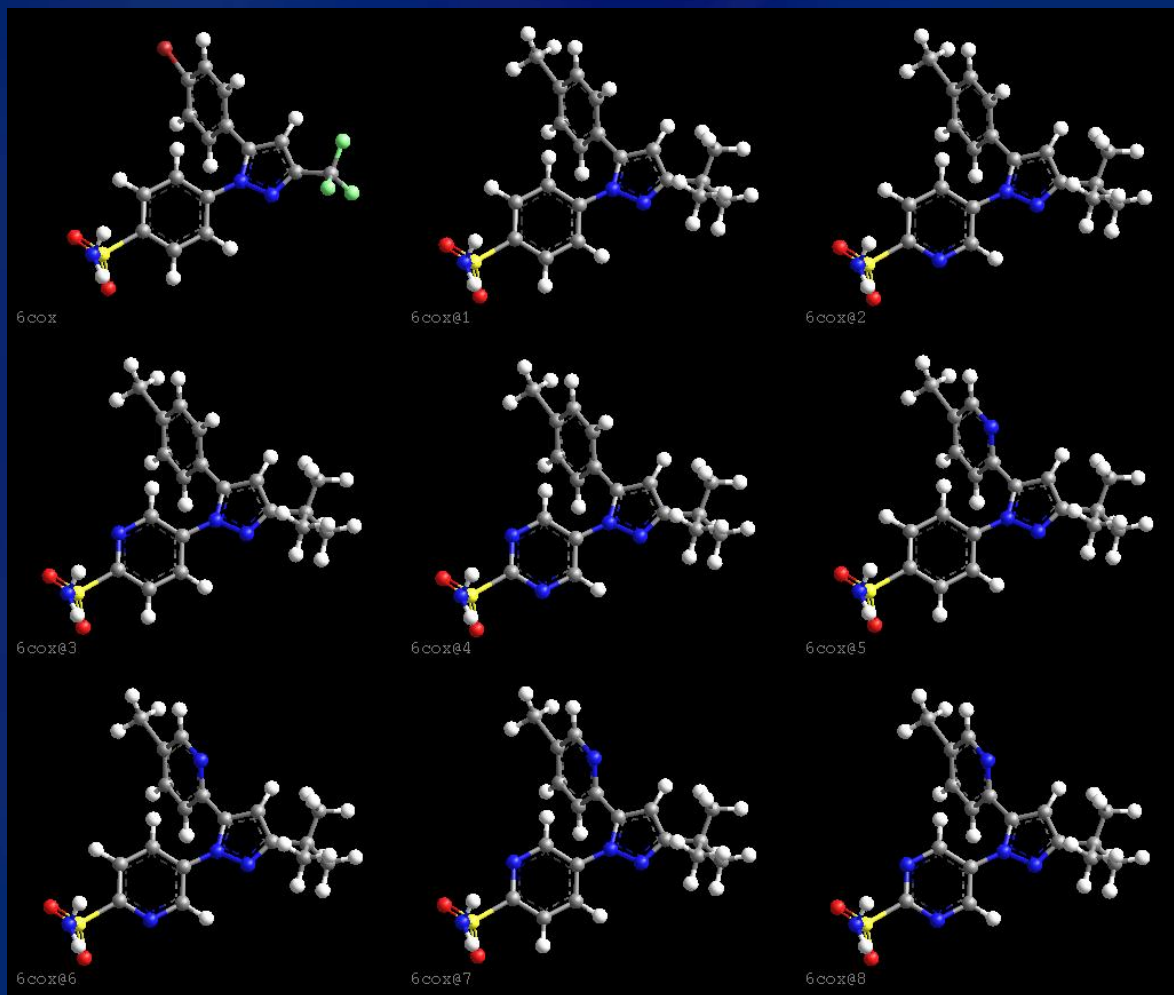


Result

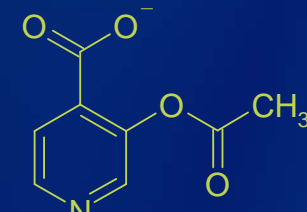
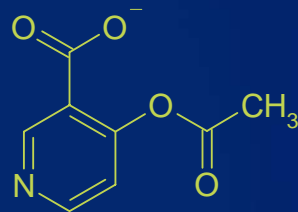
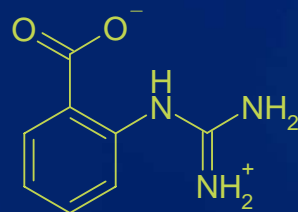
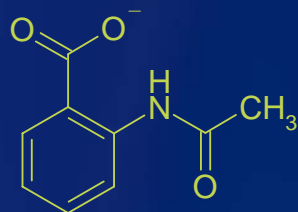
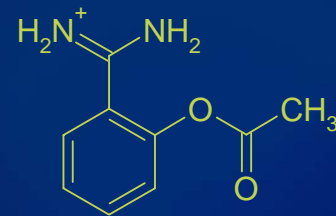
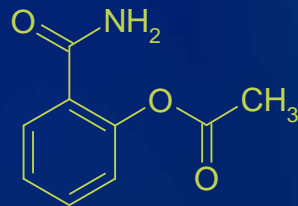
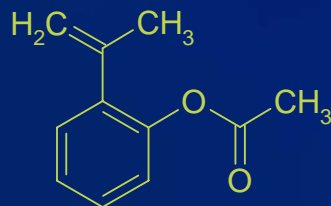
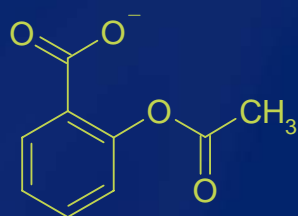


2dsim 0.55

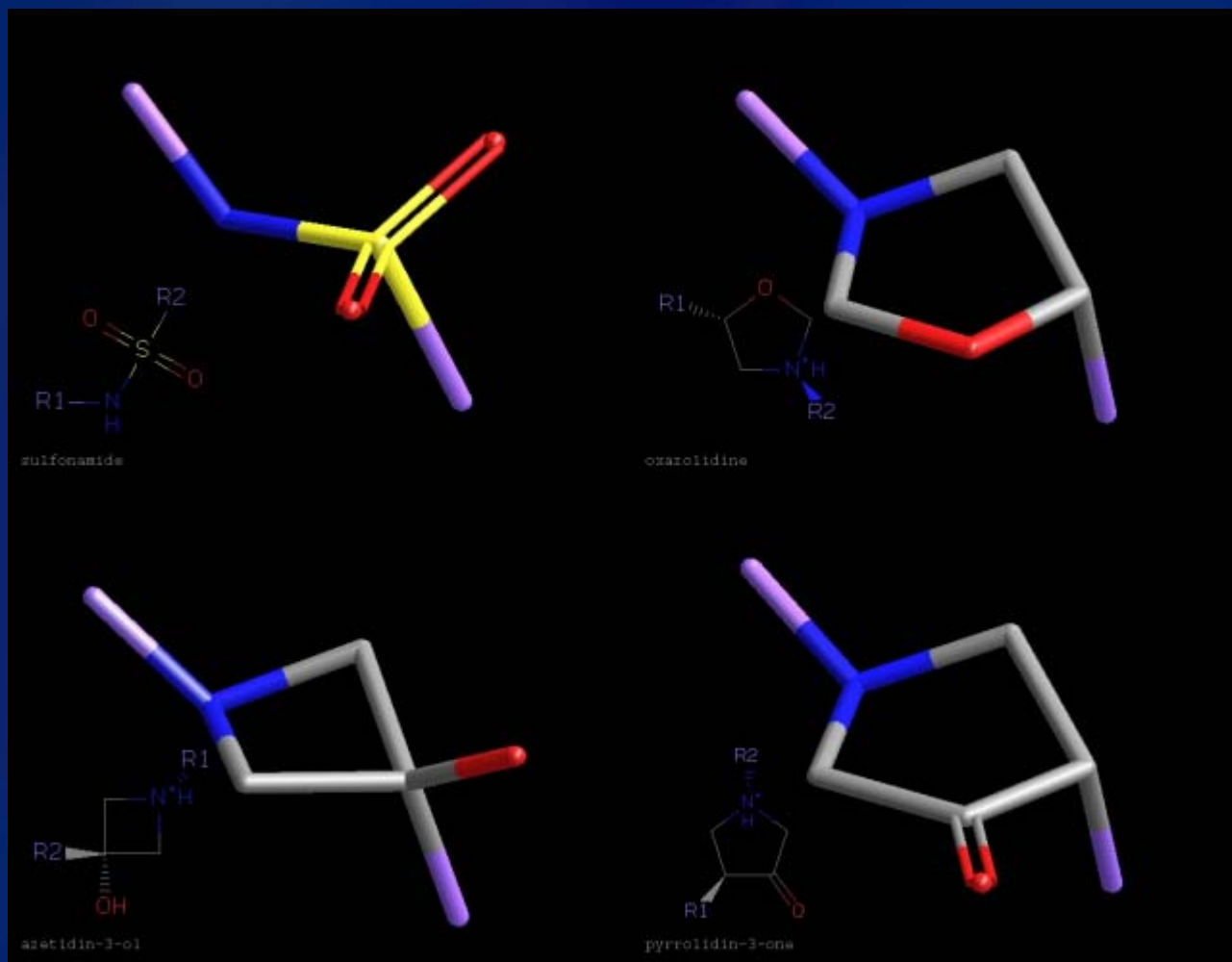
Electrostatic Modification



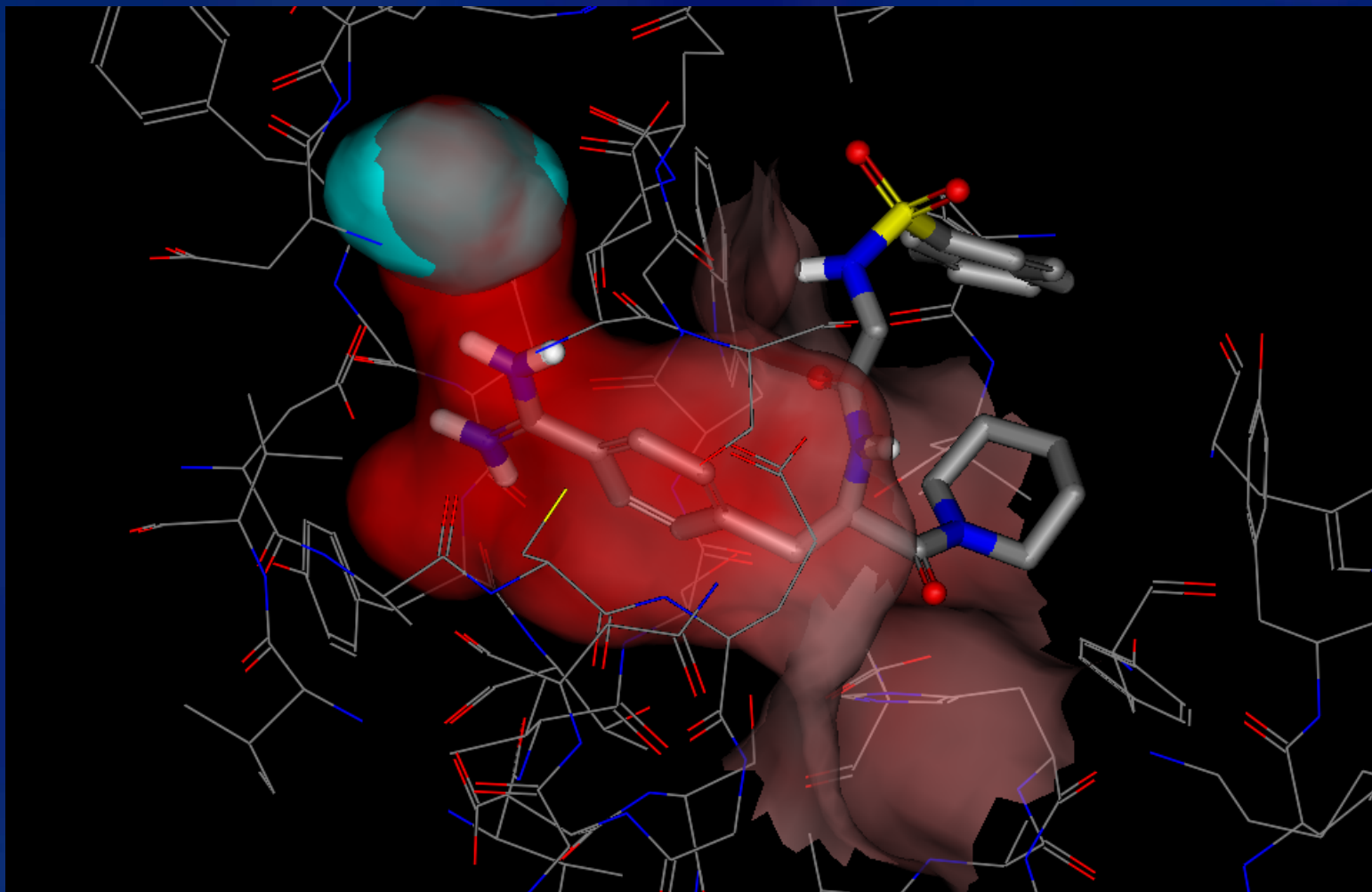
Electrostatic Modification



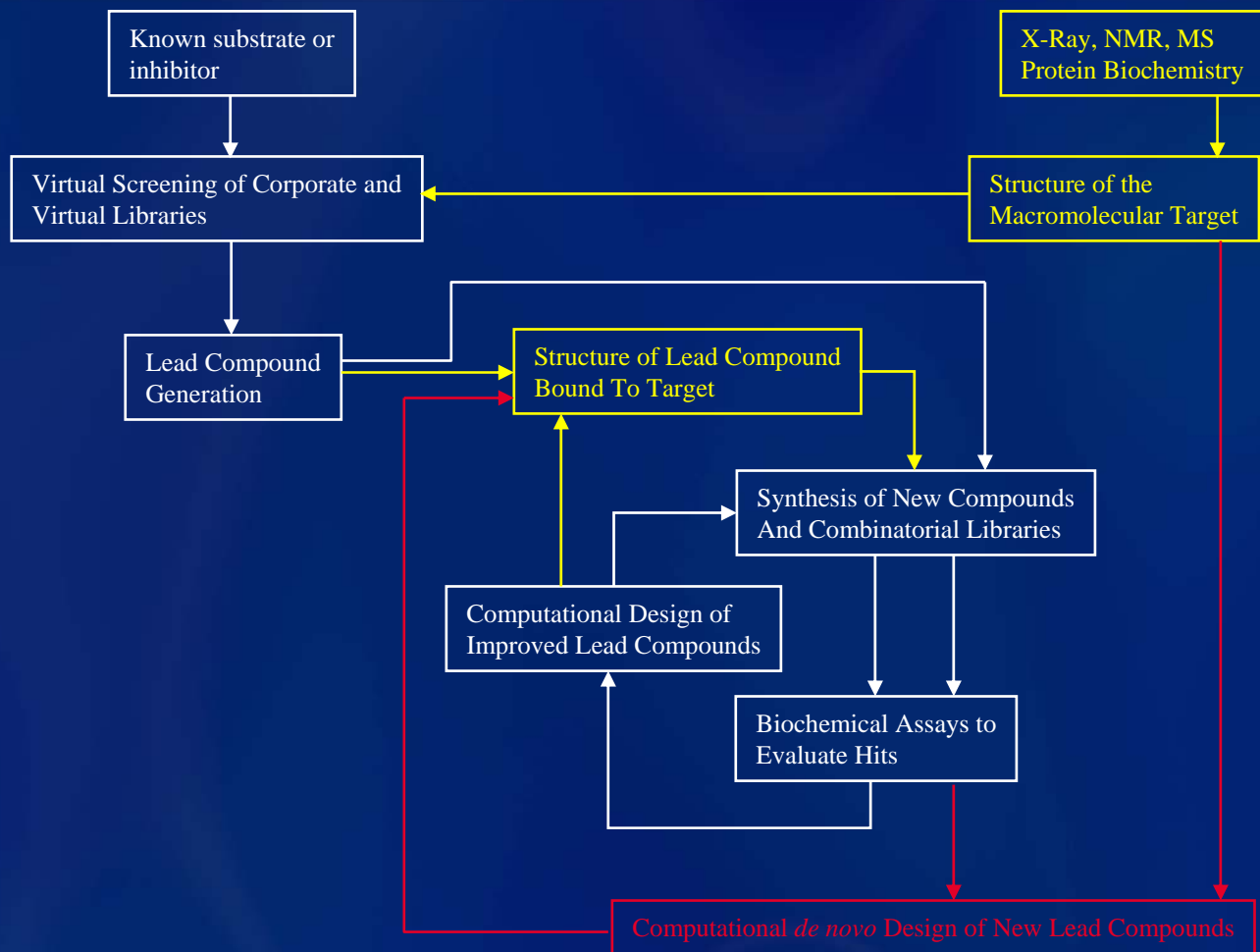
Scaffold Substitution



Void Volumes



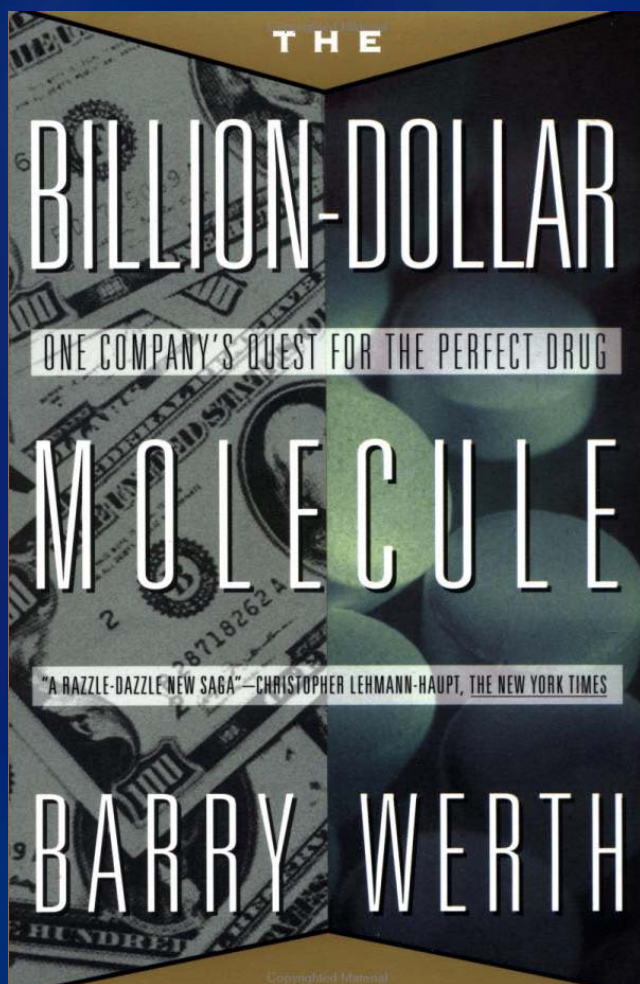
Drug Discovery



Adapted from Figure 22.1: Bourne, Wessig, [Structural Bioinformatics](#).



Fun Reading



The Billion-Dollar Molecule by Barry Werth

Story of the creation of Vertex Pharmaceuticals and their efforts to develop a blockbuster drug using rational drug design methods

*Image obtained from Amazon.com

Present your class project!

- OpenEye's 7th Annual CUP Meeting
 - March 6th to 8th, Santa Fe, NM
- Poster submissions still welcome (contrary to what it says on website)
- <http://www.eyesopen.com>

