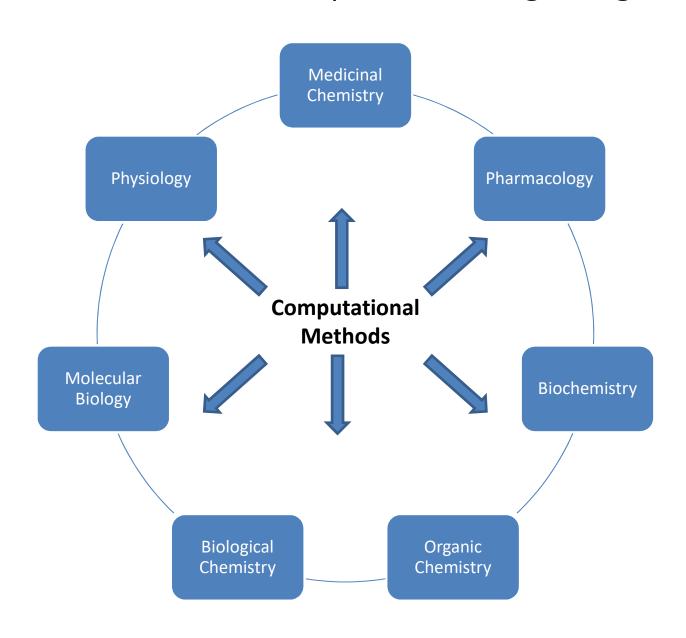
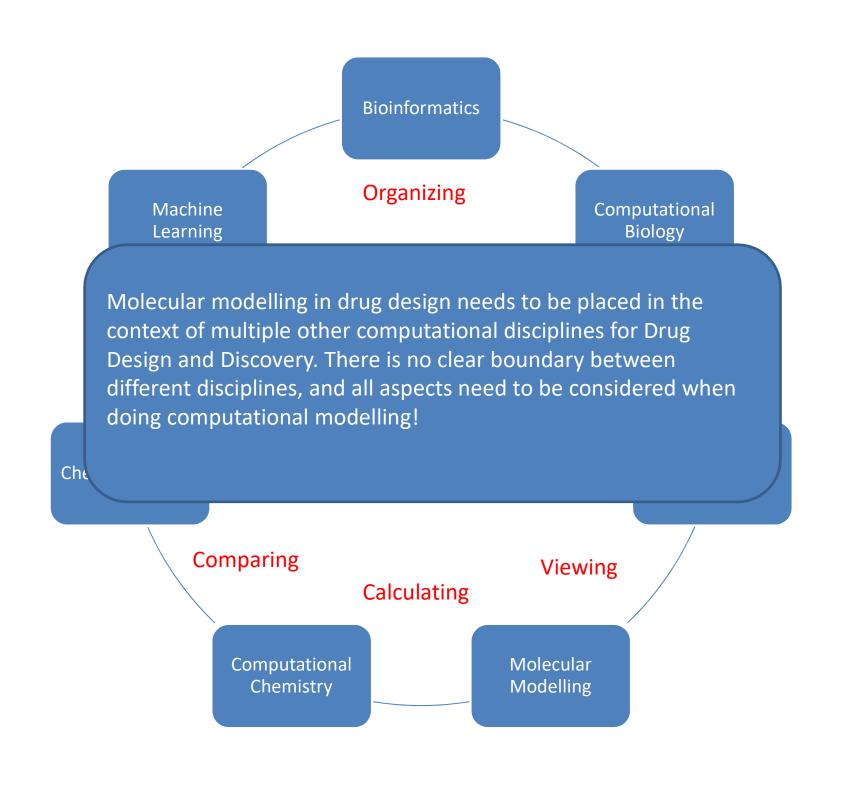
Computational Drug Design: what is it?

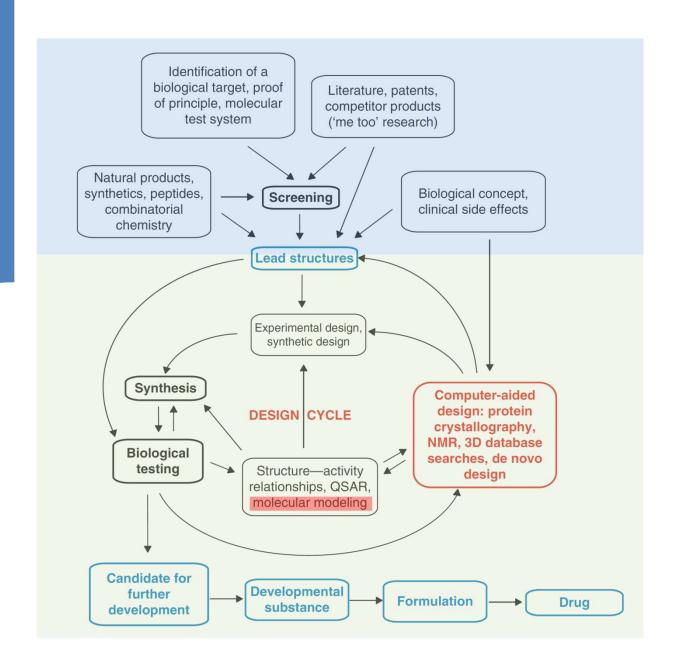
- Modern Drug Design arises from the convergence of multiple disciplines
- The chemical space is extraordinarily big and computational tools are required to fully explore it (too large for synthetic chemistry)
- Abstract and computational representation of small molecule structures
- Management of very large small molecule virtual databases
- Target are very large molecules (generally proteins) whose structure determination requires special methods where the computer is a necessary tool
- Analysis of target structures requires computational methods (very large structures with many thousands of atoms.
- Interaction between ligands and potential targets is a physicochemical process that can be modelled in a computer (docking)
- Computational techniques for molecular similarity can be used to identify new molecules sharing essential features with know ligands (pharmacophores, molecular fields, 3D QSAR)
- Sets of features (descriptors) can be used to classify and cluster molecules according to desired properties (Rule of 5, Golden Triangle, etc.)
- Automated machine learning methods can be used to classify molecules and predict potential activities, sites of metabolism or ADMET properties

The "classical" disciplines of Drug Design





How do computational techniques integrate into the Drug Discovery process?



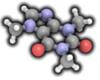
Klebe, G. Drug Design. Springer 2013, chap 1

Techniques in Molecular Modelling

Technique	Objective				
Interactive computer graphics	Display of 3D structures				
Modeling small molecules	3D Structure generation (CONCORD, CORINA)				
	Molecular mechanics—force fields				
	Molecular dynamics				
	Quantum mechanical techniques				
	Conformational analysis				
	Calculation of physicochemical properties				
Comparing molecules	Superimposition of molecules according to their similarity				
	Volume comparisons				
	3D-QSAR (e.g., CoMFA methods)				
Protein modeling	Sequence comparisons				
	Protein homology modeling				
	Protein-folding simulations				
Modeling of protein-ligand	Binding constant calculations				
interactions	Ligand docking				
Ligand design	Searches in 3D databases				
	Structure-based ligand design				
	de novo design				
	Virtual screening				

Representing chemical structures

Representation Name	Representation of Caffeine	
Common Name	Caffeine	
Synonyms	Guaranine	
	Methyltheobromine	
	1,3,7-Trimethylxanthine	
	Theine	
Empirical Formula	$C_8H_{10}N_4O_2$	
IUPAC Name	1,3,7-trimethylpurine-2,6-dione	
CAS Registry Number	58-08-2	
ChEMBL ID	CHEMBL113	
Wiswesser Line Notation	T56 BN DN FNVNVJ B F H	
(WLN)		
SMILES	CN1C=NC2=C1C(=O)N(C(=O)N2C)C	
Aromatic SMILES	CN1C(=O)N(C)c2ncn(C)22C1=O	
InChI	1S/C8H10N4O2/c1-10-4-9-6-	
	5(10)7(13)12(3)8(14)11(6)2/h4H,1-3H3	
InChIKey	RYYVLZVUVIJVGH-UHFFFAOYSA-N	
Topography	8	

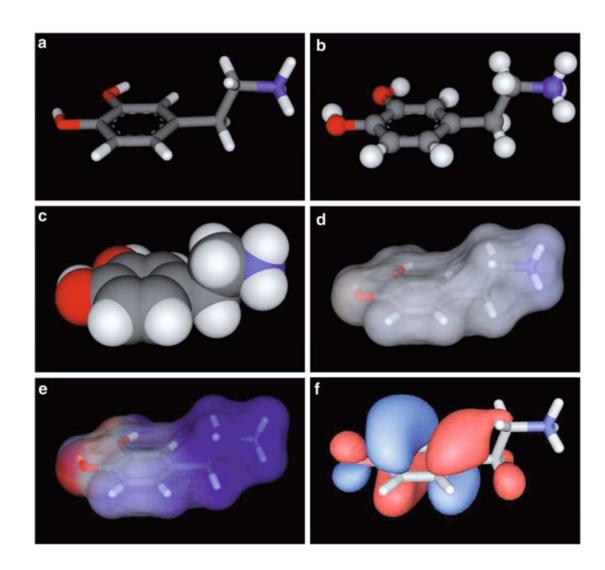


Surface

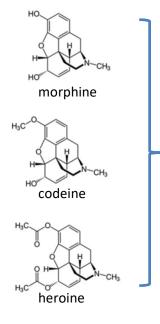


Vsizualizing chemical structures

- **a** dreiding model
- **b** ball-and-stick
- $\mathbf{c} \text{vdW (CPK)}$
- **d** molecular surface
- **e** surface potential
- **f** HOMO orbitals



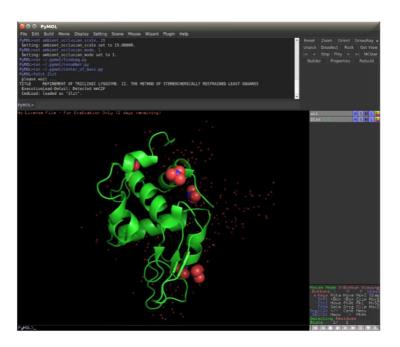
The Importance of molecular similarity

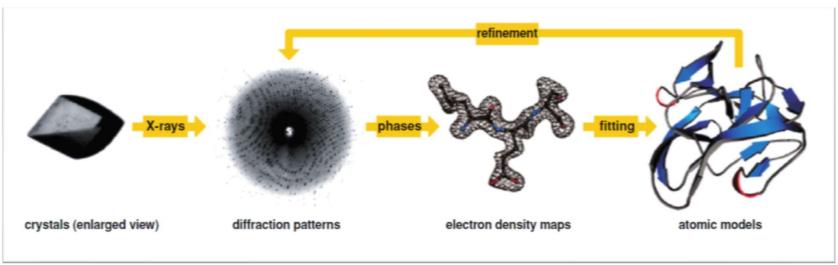


Similar structures similar functions

Chemical similarity	A	Mol. weight	LogP 5.23	Rotatable bonds	Aromatic rings	Heavy atoms		
	В	463.5	4.43	4	5	35		
Molecular similarity	NN N H							
2D similarity	A B H							
3D similarity	A B							
Biological similarity	A		ar endothelia actor recepto active active		ne-protein k 2 inactive active	inase TIE-		
Global similarity	and and a							
Local similarity	~				В	, n		

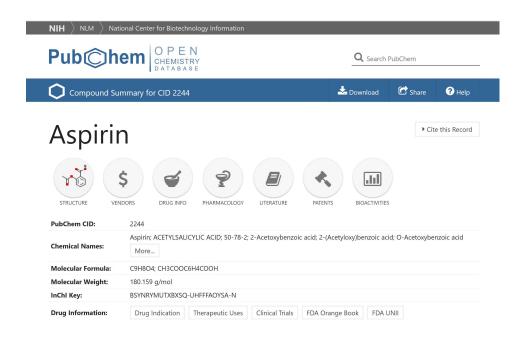






Protein structure determination by X-ray crystallography

Chemical Databases



Database	Description	Size	web addresses
DrugBank ^[5]	Collection of approved and experimental drugs	7895	https://www.drugbank.ca/
CTD[6]	Toxicogenomics database	12 K	http://ctdbase.org/about/dataStatus.go
NCI[7]	National cancer institute chemical database	265 K	https://cactus.nci.nih.gov/
BindingDB[8]	Bioactive small molecules annotated with experimental data	600 K	https://www.bindingdb.org/bind/index.jsp
ChEMBL ^[9]	Bioactive small molecules annotated with experimental data	1.7 M	https://www.ebi.ac.uk/chembldb
SureChEMBL[10]	Collection of patented compounds	17 M	https://www.surechembl.org/search/
eMolecules	Commercial small molecules for screening	7 M	https://www.emolecules.com/
ChemSpider	Collection of compounds from various institutions and commercial companies	58 M	http://www.chemspider.com/
PubChem[11]	NIH repository of molecules	93 M	http://pubchem.ncbi.nlm.nih.gov
ZINC 15[12]	Commercial small molecules for screening	378 M	http://zinc15.docking.org/
GDB-11[13]	Possible small molecules up to 11 atoms of C, N, O, F	26 M	http://gdb.unibe.ch
GDB-13[14]	Possible small molecules up to 13 atoms of C, N, O, S, Cl	980 M	http://gdb.unibe.ch
GDB-13.FL[15]	Fragrance-like subset of GDB-13	59 M	http://gdb.unibe.ch
GDB-17 ^[16]	Possible small molecules up to 17 atoms of C, N, O, S and halogens	166 B	http://gdb.unibe.ch
FDB-17[17]	Fragment like subset of GDB-17	10 M	http://gdb.unibe.ch

What Makes a Good Drug?

Finding the essential chemical descriptors (dimensionality reduction), classifying, filtering, selecting.

Machine learning-methods

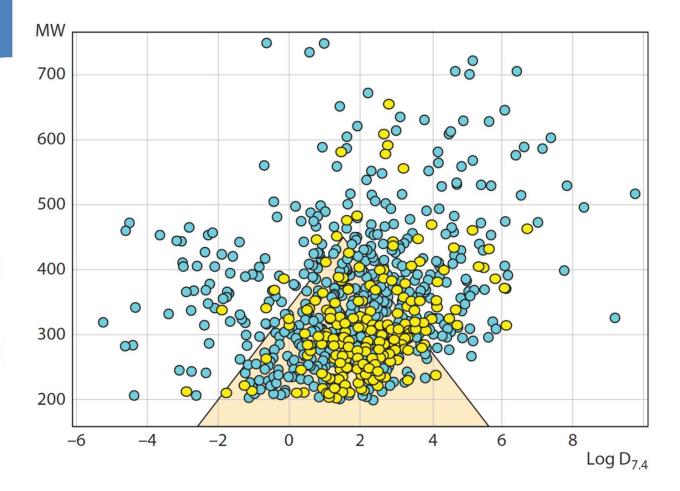
Lipinski's rule of 5

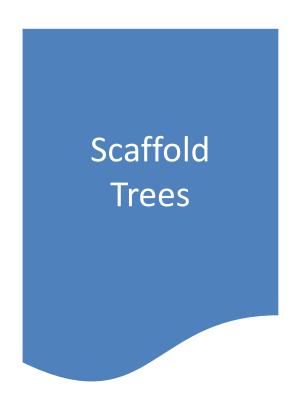
Peripheral drugs

84% Ro5 compliant 53% inside the Golden Triangle 70% have CNS MPO score > 4

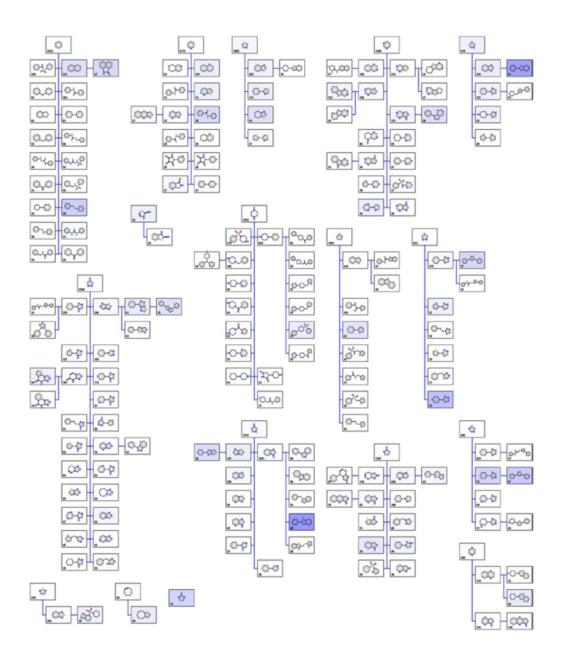
CNS drugs

92% Ro5 compliant 77% inside the Golden Triangle 70% have CNS MPO score > 4



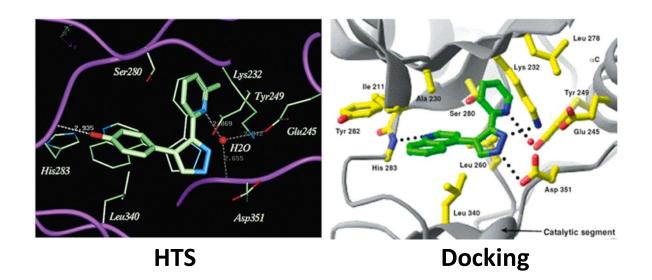


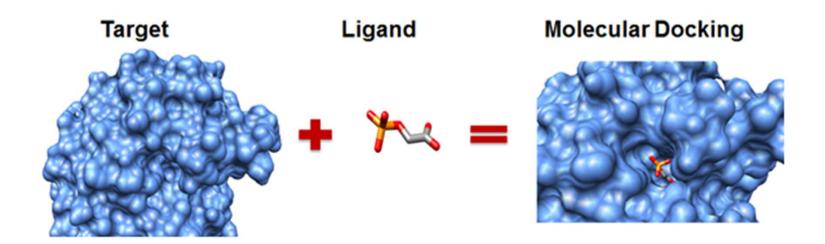
Guided search through chemical space.



Color intensity represents potency



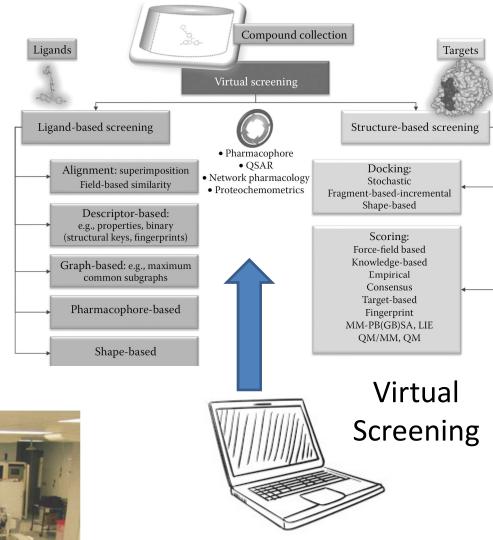


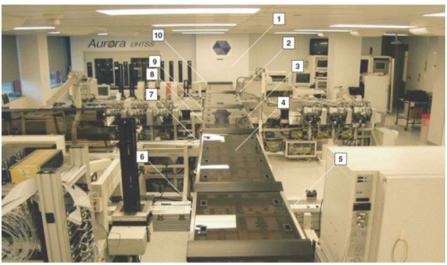


Virtual
Screening

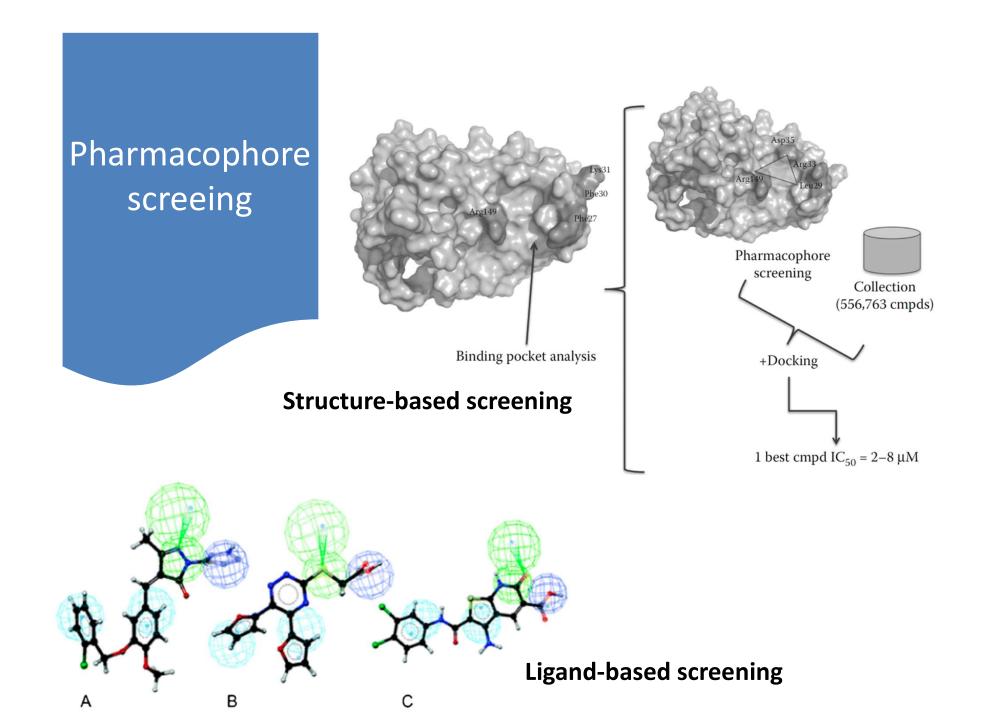
versus

Real
Screening

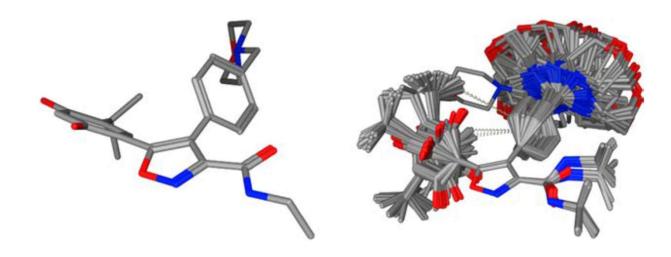




Hight-Throughput Screen Laboratory

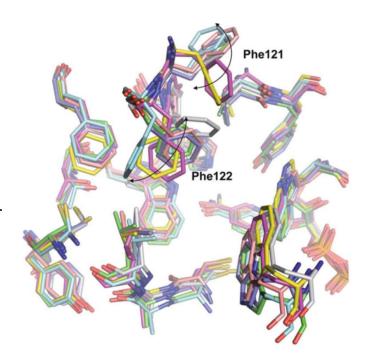


Importance of Conformational Search



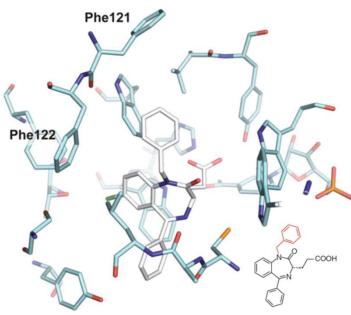
Importance of of Molecular Dynamics Simulations

MD simulation shows wide movement of Phe121 residue, enlarging the binding pocket of the receptor



Benzodiazepine-like inhibitor

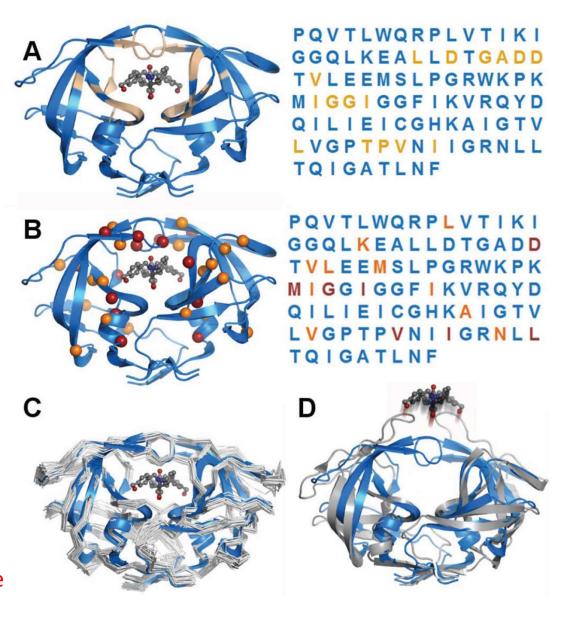
The open conformation can accommodate ligands with extended functional groups, like the red group of the benzodiazepine-like inhibitor,



Sequence and
Structure
Analysis of
Protein
Targets

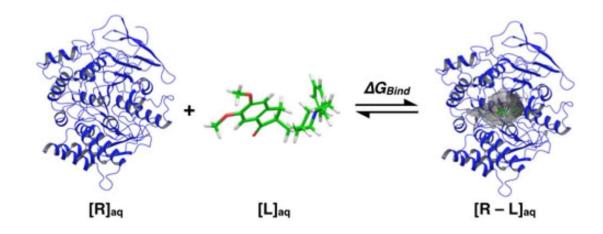
HIV protease

- A Residues near bound inhibitor
- **B** Mutations leading to resistance
- **C** Mutations can affect flexibility
- **D** Dynamics of ligand free protein (studied by *MD simualtions*)



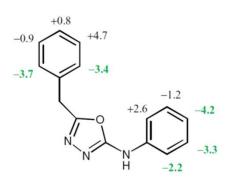
Urrutia (2016). F1000, 5:766

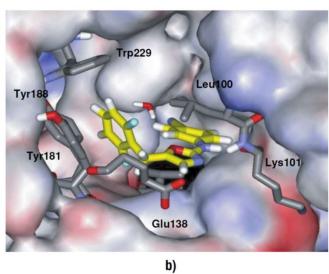
Calculation
of binding
free energies
of
ligands



$$K_{\rm d} = \frac{[R][L]}{[RL]}$$

$$\Delta G = RT \ln K_{\rm d}$$





a)

Prediciton of binding afinities

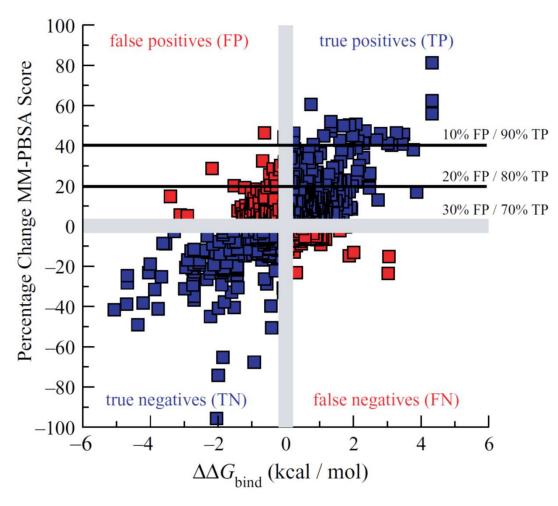
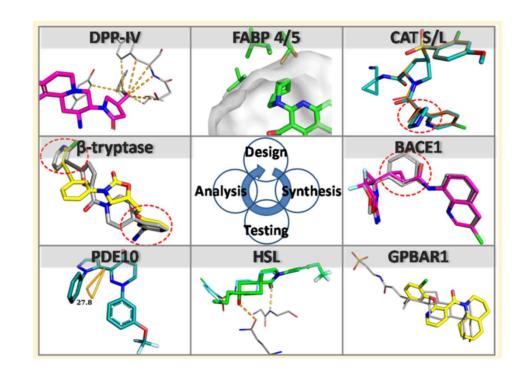


Figure 5.3. Data showing change in compound potency (relative to a reference compound) versus percentage change in MM-PBSA score (relative to same reference compound) for 480 compounds across eight targets, which span 292 x-ray crystallographic complexes.

CADD works in the "real world"





Perspective

pubs.acs.org/jmc

A Real-World Perspective on Molecular Design

Miniperspective

Bernd Kuhn, Wolfgang Guba, Jérôme Hert, David Banner, Caterina Bissantz, Simona Ceccarelli, Wolfgang Haap, Matthias Körner, Andreas Kuglstatter, Christian Lerner, Patrizio Mattei, Werner Neidhart, Emmanuel Pinard, Markus G. Rudolph, Tanja Schulz-Gasch, Thomas Woltering, and Martin Stahl*



Roche Pharmaceutical Research and Early Development, Roche Innovation Center Basel, F. Hoffmann-La Roche Ltd., Grenzacherstrasse 124, 4070 Basel, Switzerland

CADD and diseases

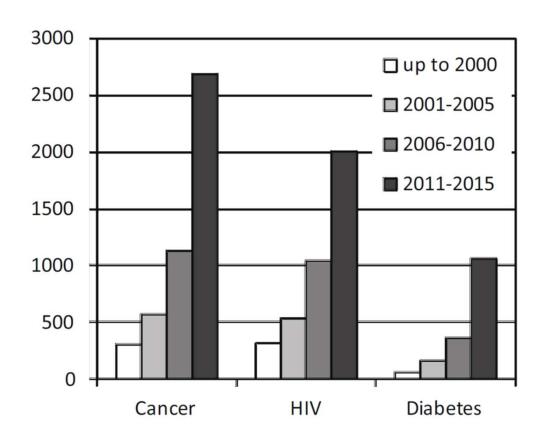


Fig. (1). The number of publications related to computer-aided drug design and diseases. Key words used in the Google Scholar search [16] were as follows: computer-aided drug design and disease; *e.g.* diabetes.