

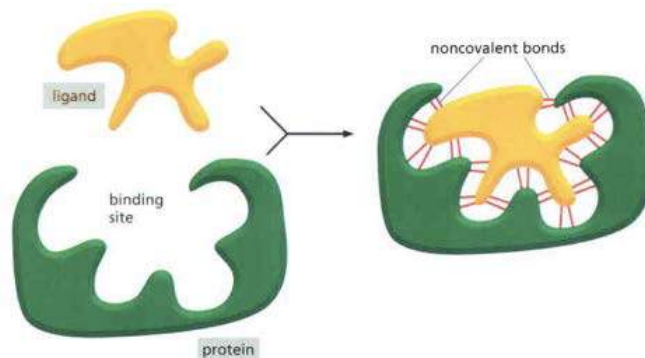
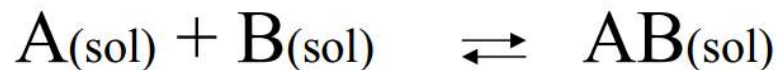
# Binding free energy theory and MM/PBSA method

Speaker: Xiaoxiao Zhang

2014.9.24

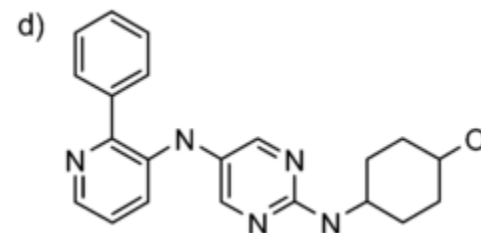
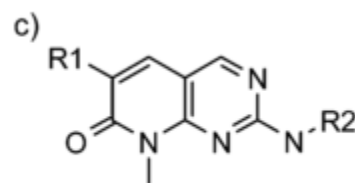
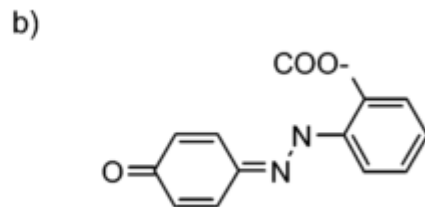
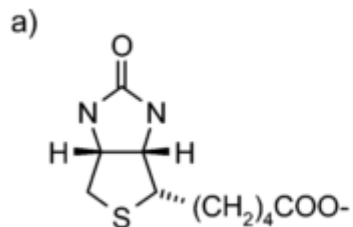
# Overview

## Quasi-Chemical Description



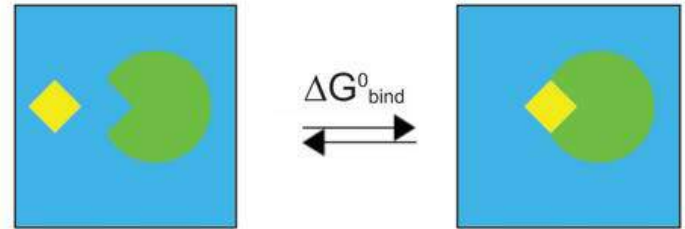
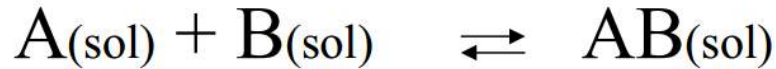
$$\Delta G^0 = \Delta G^0_{\text{tr}} + \Delta G^0_{\text{rot}} + \Delta G^0_{\text{int}} + \Delta G^0_{\text{solvn.}}$$

## Molecular Mechanics/ Poisson-Boltzmann Surface Area (MM/PBSA)



# Protein-ligand binding

## Quasi-Chemical Description



$$\mu_{\text{sol},i} = \mu_{\text{sol},i}^{\circ} + k T \ln \frac{C_i}{C^{\circ}}$$

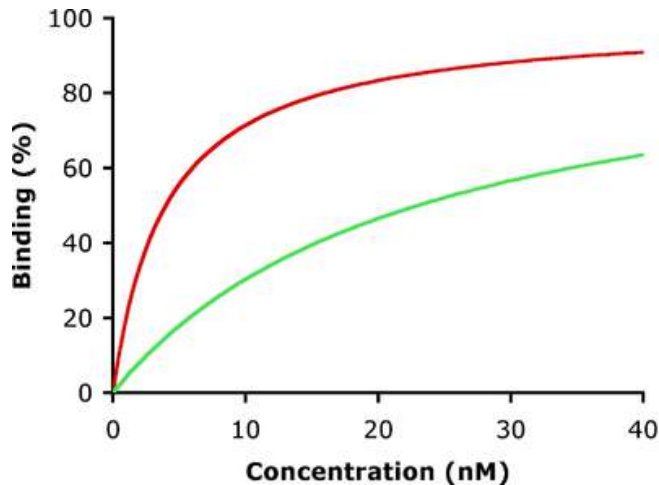
$$\Delta G_{AB}^{\circ} = \mu_{AB}^{\circ} - \mu_A^{\circ} - \mu_B^{\circ} = -k T \ln \frac{C_{AB}/C^{\circ}}{(C_A/C^{\circ})(C_B/C^{\circ})} = -k T \ln K_{AB}$$

$$K_{AB} = e^{-\beta \Delta G_{AB}^{\circ}} = \frac{e^{-\beta \mu_{AB}^{\circ}}}{e^{-\beta(\mu_A^{\circ} + \mu_B^{\circ})}} = \frac{C^{\circ}}{8 \pi^2} \frac{Z_{AB}}{Z_A Z_B};$$

The standard binding free energy:

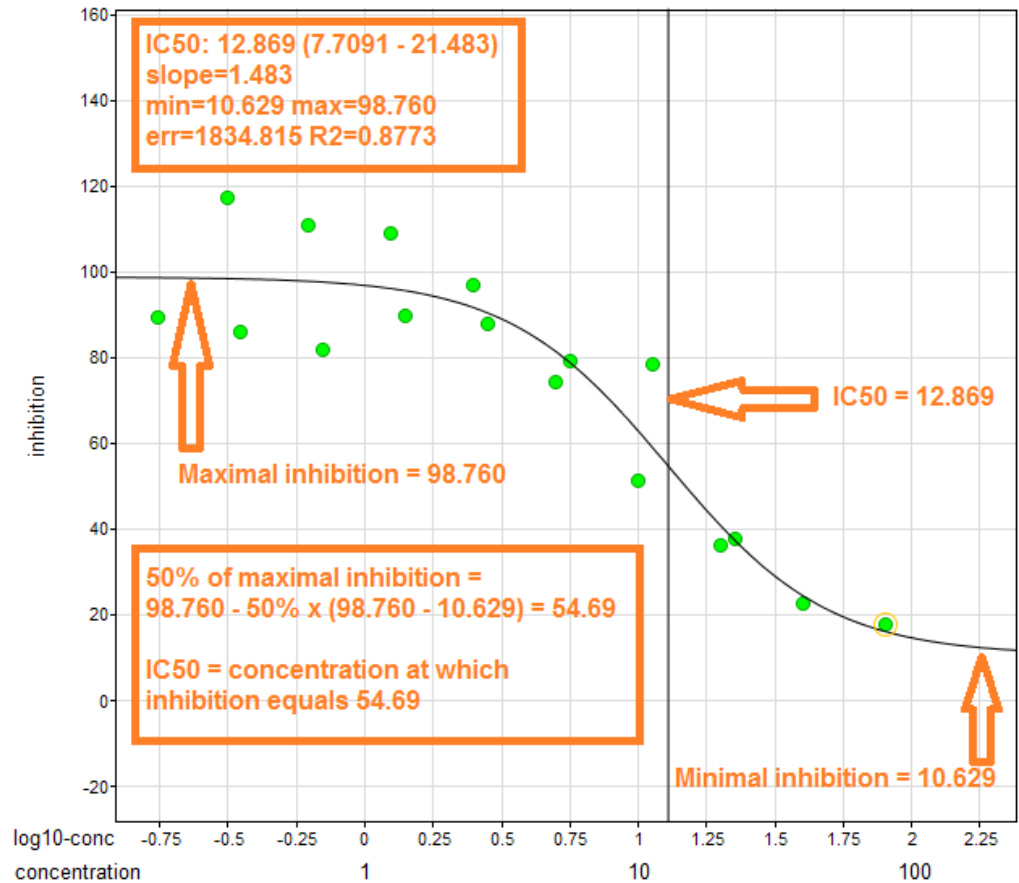
- Include a definition of the species “AB”.
- Depend on the standard state concentration.

# Binding affinity measurements



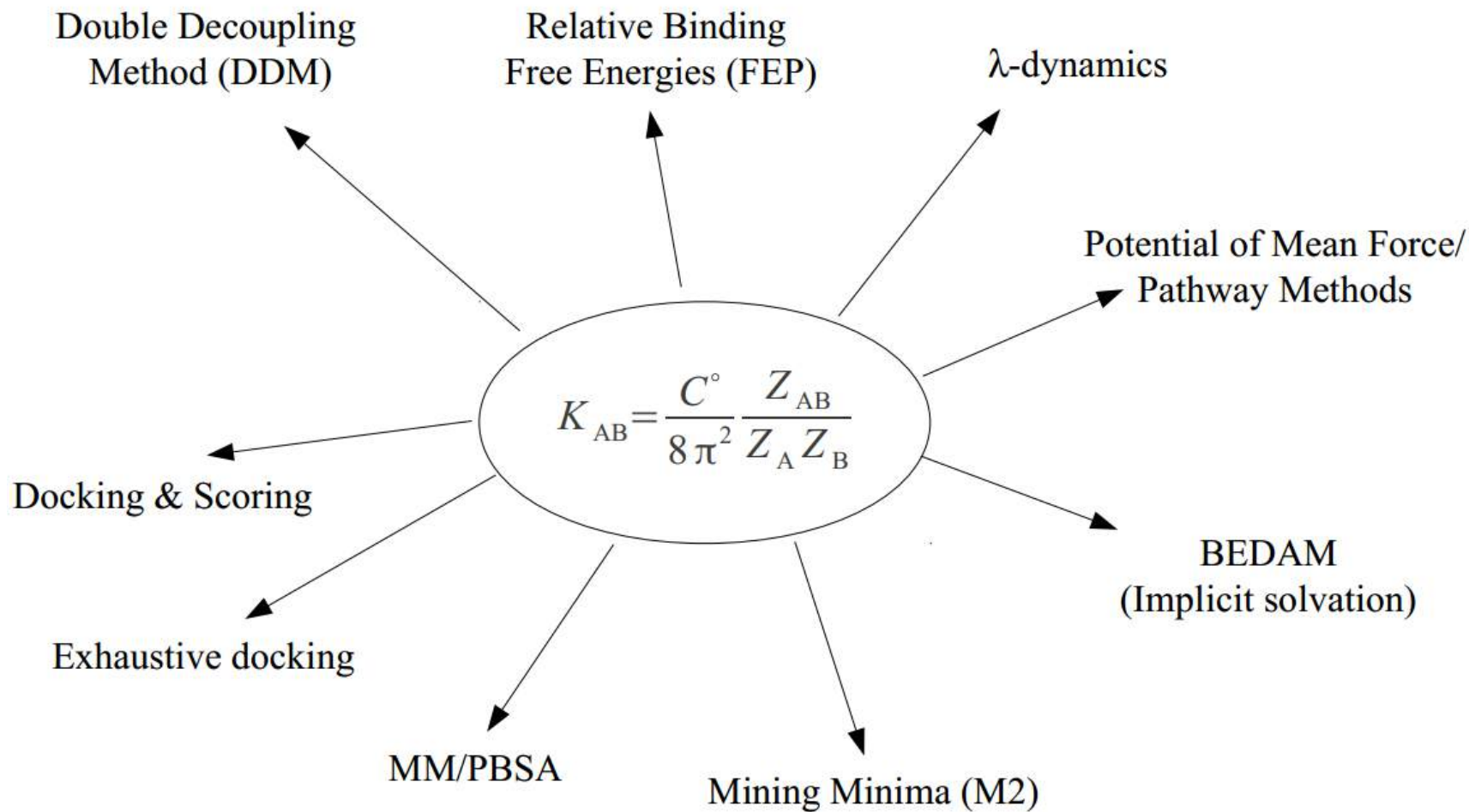
## Cheng-Prusoff Equation:

$$K_i = IC_{50} / (1 + [L]/K_d)$$



- The affinity (equilibrium inhibition constant  $K_i$ ) of a drug for a receptor is a measure of how strongly that drug binds to the receptor.
- The  $IC_{50}$  stands for the half maximal inhibitory concentration.
- The lower  $K_i$ , the higher binding affinity.

# Free Energy Computational Methods

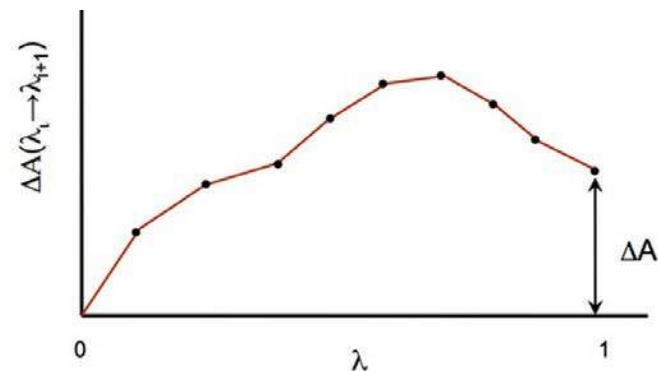
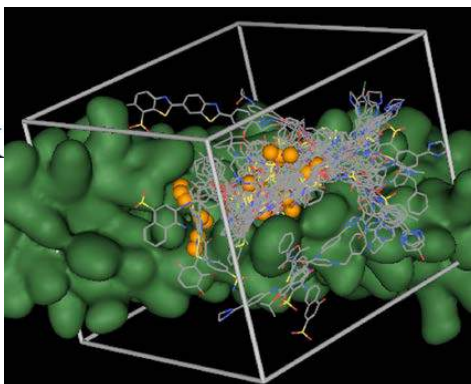
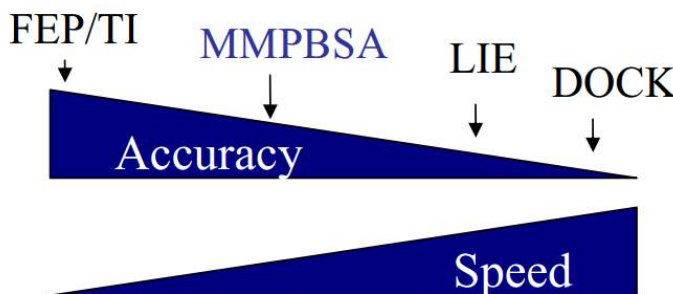


# Free Energy Computational Methods

The theoretical basis for the additivity assumed in free energy computation:

$$\Delta G^0 = \Delta G^0_{\text{tr}} + \Delta G^0_{\text{rot}} + \Delta G^0_{\text{int}} + \Delta G^0_{\text{solvn.}}$$

Docking & Scoring	Screening & Enrichment	1000's of compounds
MM-PB/SA Linear Interaction Energy	Ranking	100's of compounds
Absolute/Relative Binding Free Energy Methods	Ranking, Optimization, Specificity, Resistance	10's of compounds

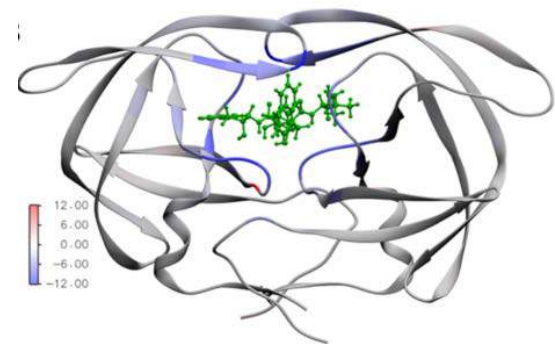
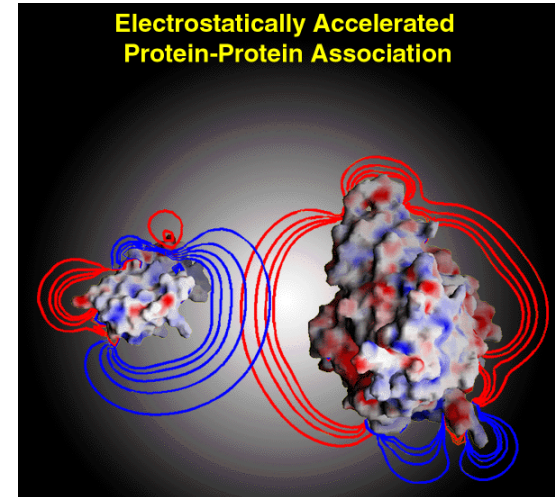


# MM/PBSA & MM/GBSA

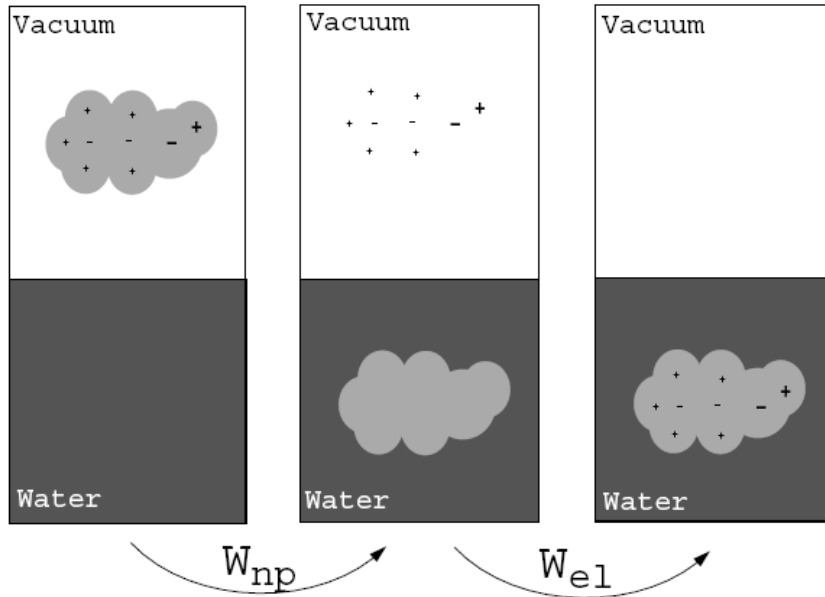
- Molecular Mechanic/ Poisson-Boltzmann Surface Area (MM/PBSA)
- MM = internal energy of the system (bonded & non-bonded)
- PB = electrostatic contribution to solvation
- SA = nonpolar contribution to solvation
- S = entropic contribution to binding

## Advantage over FEP:

- Generalizable.
- Can take into account large structure changes.
- Calculates G at endpoints, don't sample intermediates.
- With errors larger, still get good agreement with experiment.
- Continuum solvent has integrated out solvent coordinates.



# Solvation Model



$$\Delta G_{sol} = \Delta G_{np} + \Delta G_{el}$$

$$\Delta G_{np} = \Delta G_{vdW} + \Delta G_{cav}$$

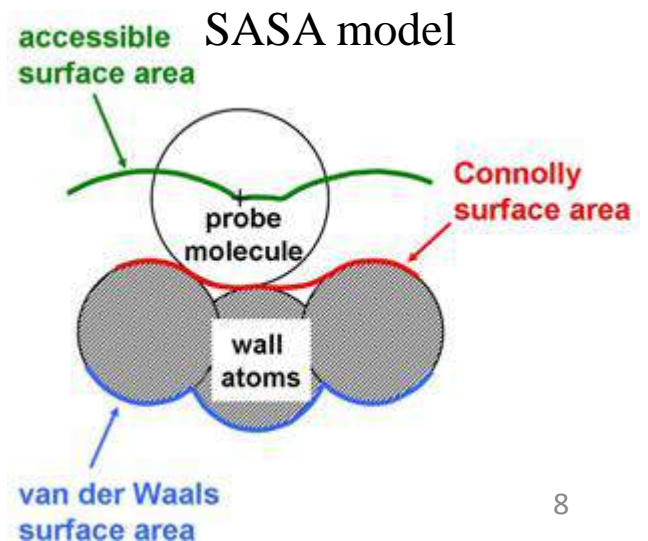
- Net electrostatics.
- Van der Waals interactions.
- Cavitation effects.

## Generalized Born Equation

$$G_{pol} = -166 \left( 1 - \frac{1}{\epsilon} \right) \sum_{i=1}^n \sum_{\substack{j=1 \\ j \neq i}}^n \frac{q_i q_j}{f_{GB}} - 166 \left( 1 - \frac{1}{\epsilon} \right) \sum_{i=1}^n \frac{q_i^2}{\alpha_i}$$

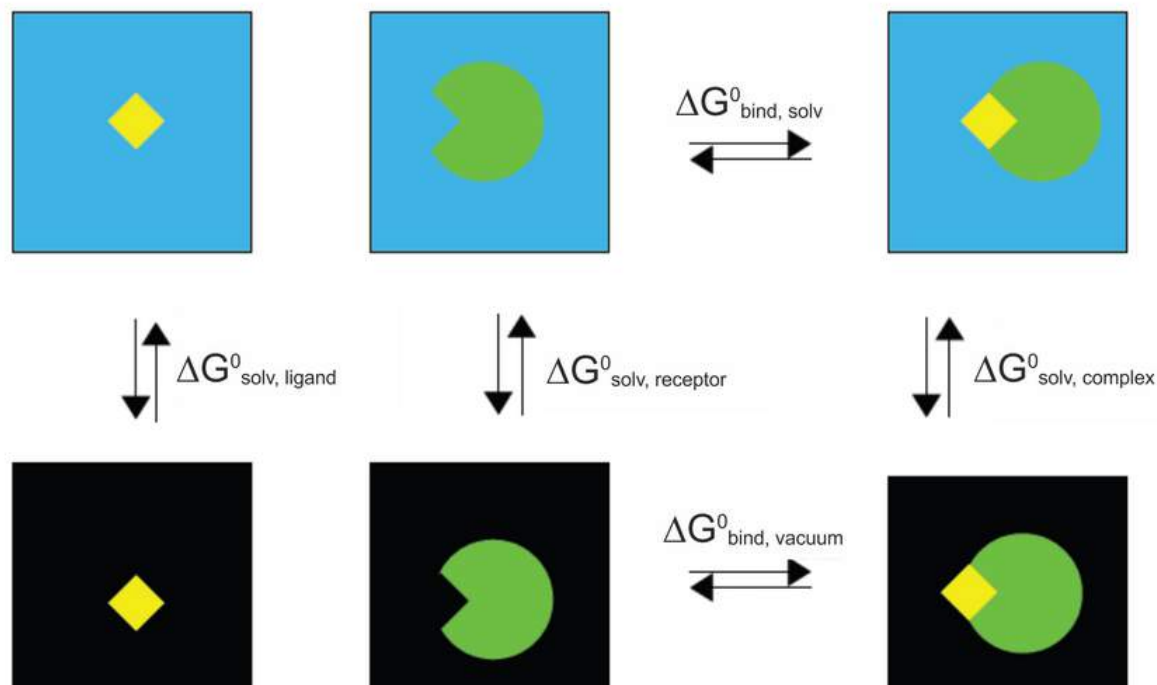
## Poisson-Boltzmann Equation

$$\nabla \left[ \epsilon(r) \nabla \phi(r) \right] = -4\pi \left( \rho_{solute}(r) + \sum_{i=1}^N q_i c_i^{bulk} \exp(-q_i \phi(r) / RT) \right)$$





# Methodology



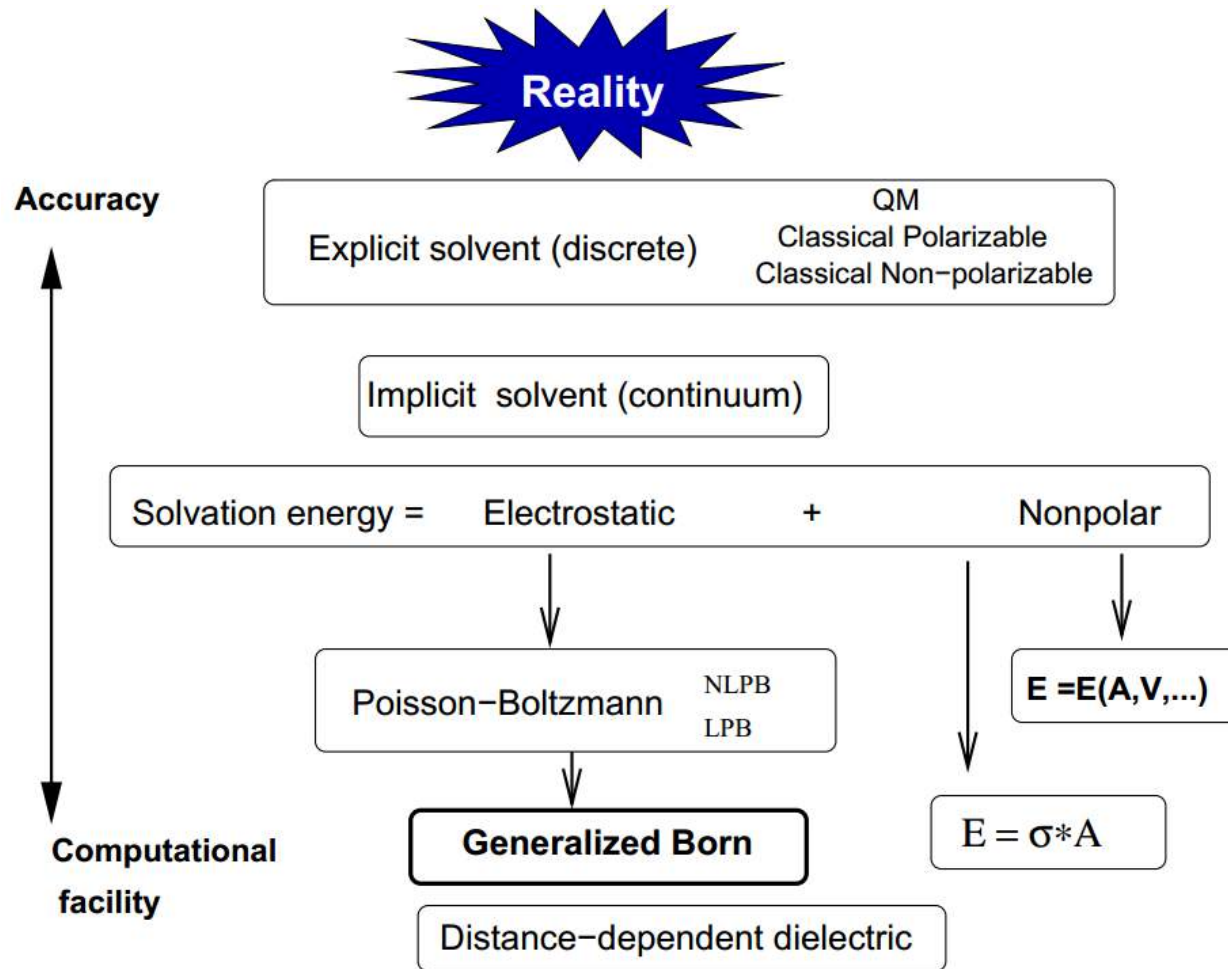
$$\Delta G^0_{bind, solv} = \Delta G^0_{bind, vacuum} + \Delta G^0_{solv, complex} - (\Delta G^0_{solv, ligand} + \Delta G^0_{solv, receptor})$$

$$\Delta G^0_{solv} = G^0_{electrostatic, \epsilon=80} - G^0_{electrostatic, \epsilon=1} + \Delta G^0_{hydrophobic}$$

$$\Delta G^0_{vacuum} = \Delta E^0_{molecular\ mechanics} - T \cdot \Delta S^0_{normal\ mode\ analysis}$$

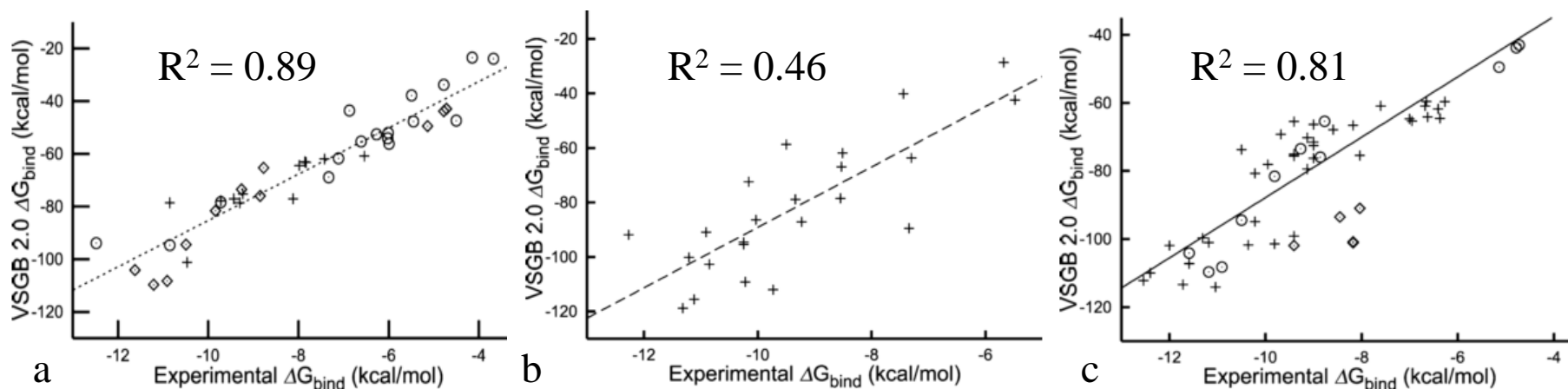
Takes advantage of multiple snapshots from a trajectory to get an average of energies.

# Accuracy



The hierarchy of representations of solvent effects in molecular modeling. The GB model is separated from reality by several layers of approximation.

# Accurate enough to correlate with binding affinity?



- the most accurate data-set.
- human aldose reductase with high variability in the affinity data.
- PDPK1 family with affinity data but no crystal structures.

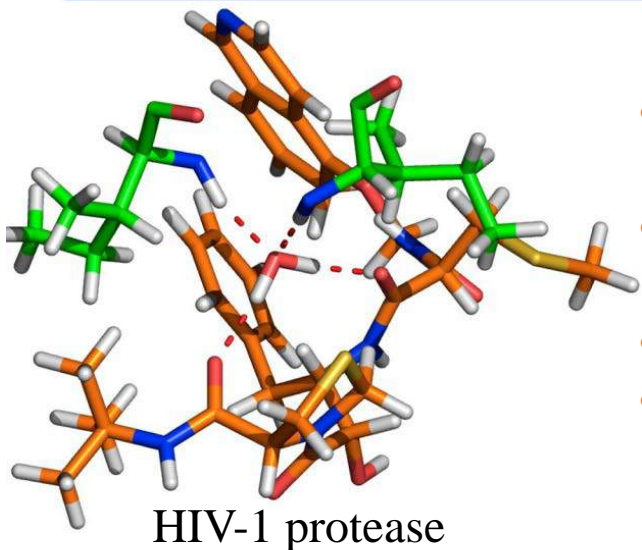
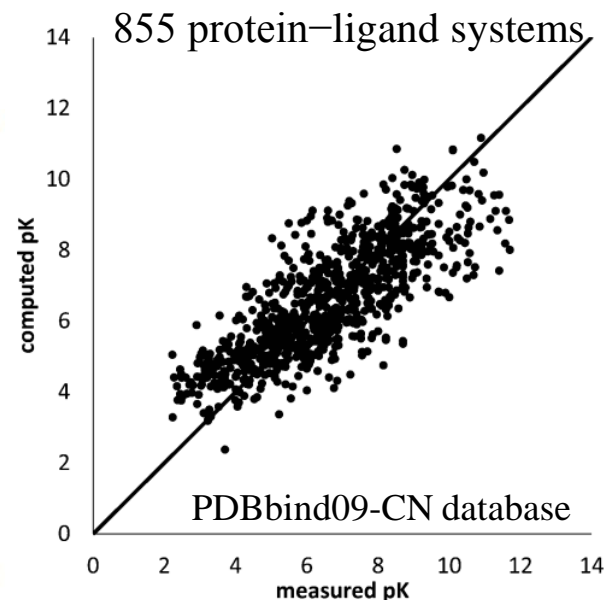
- Demonstrate the importance of **affinity data variability** and errors in the modeled complexes used for computing binding free energies.
- MM/PBSA could be approaching the accuracy required for evaluating relative binding free energies.

# Accurate enough to correlate with binding affinity?

$$\text{computed } pK = -0.0456 * \text{MM-GBSA Energy} + 2.564$$

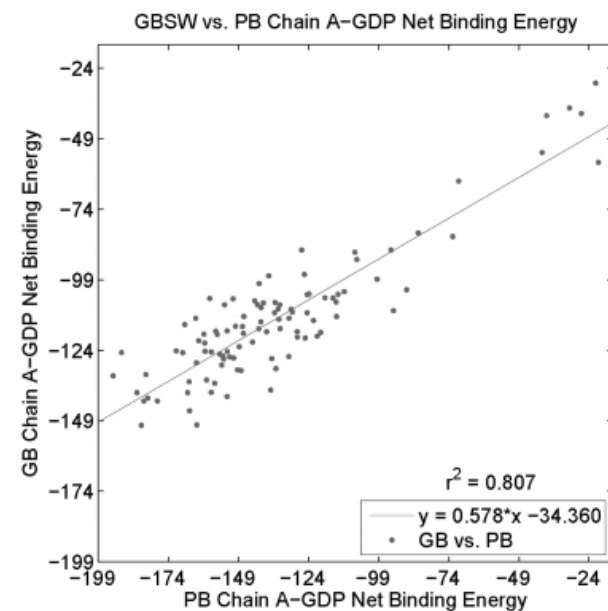
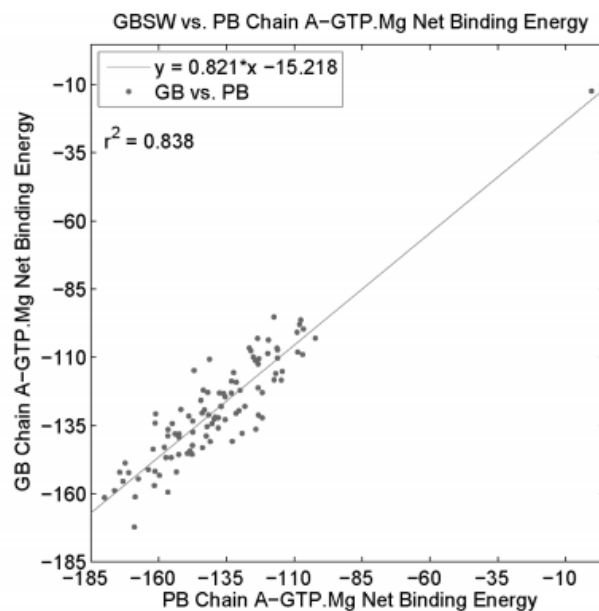
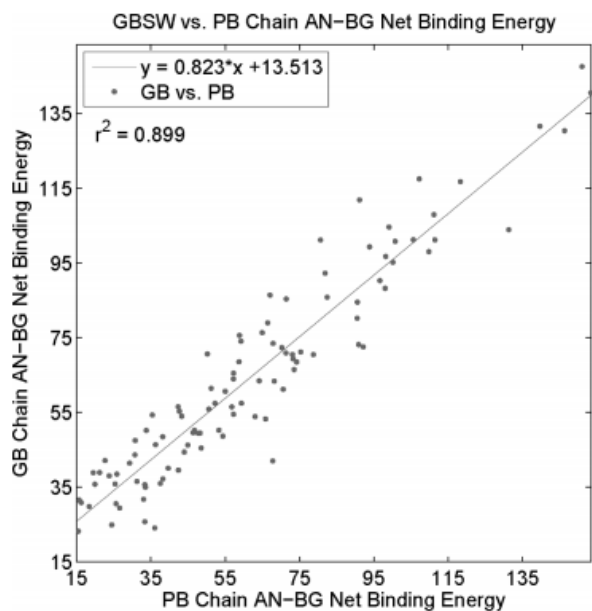
Table 1. Summary of  $R^2$ , RSE, and MUE Values for MM/GBSA Calculations with Different Conditions

conditions	$R^2$	RSE	MUE
ligand strain/water	0.58	1.29	1.03
ligand strain/no water	<b>0.63</b>	<b>1.20</b>	<b>0.96</b>
no ligand strain/no water	0.61	<b>1.20</b>	0.98
no ligand strain/water	0.56	1.32	1.06

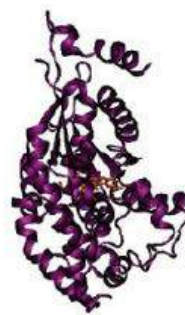


- A significant **correlation** between MM/GBSA energies and experimental  $K_i$ .
- MM/GBSA approach can yield values that are comparable even among different targets.
- The inclusion of **water deteriorates** the predictive quality.
- The inclusion of ligand strain slightly improves the overall accuracy.

# Performance of MM/PBSA vs. MM/GBSA



$G_{\alpha}^{\text{GDP}}/G_{\beta\gamma}$



$G_{\alpha}/\text{GTP} \cdot \text{Mg}$

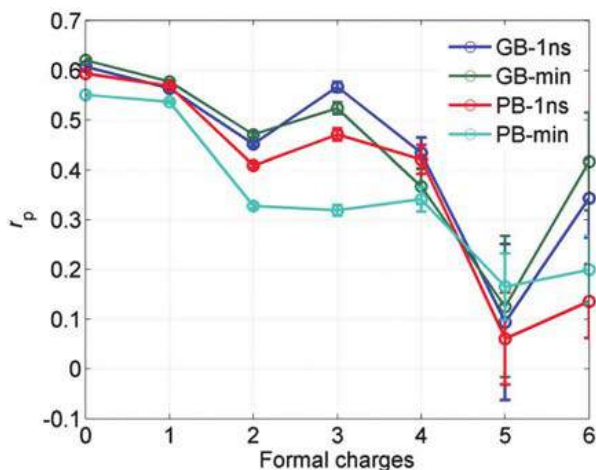
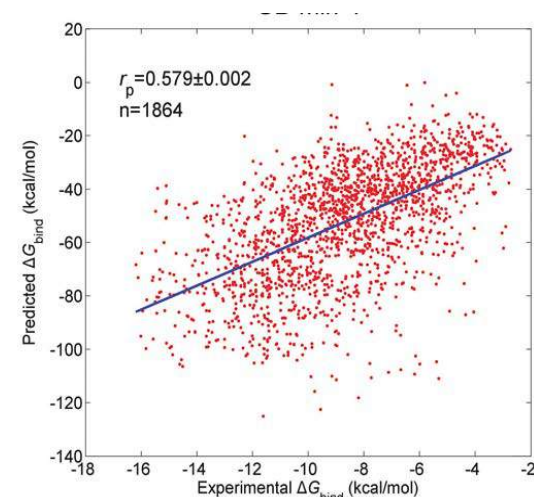


$G_{\alpha}/\text{GDP}$

# Performance of MM/PBSA vs. MM/GBSA

Dielectric constant: 4

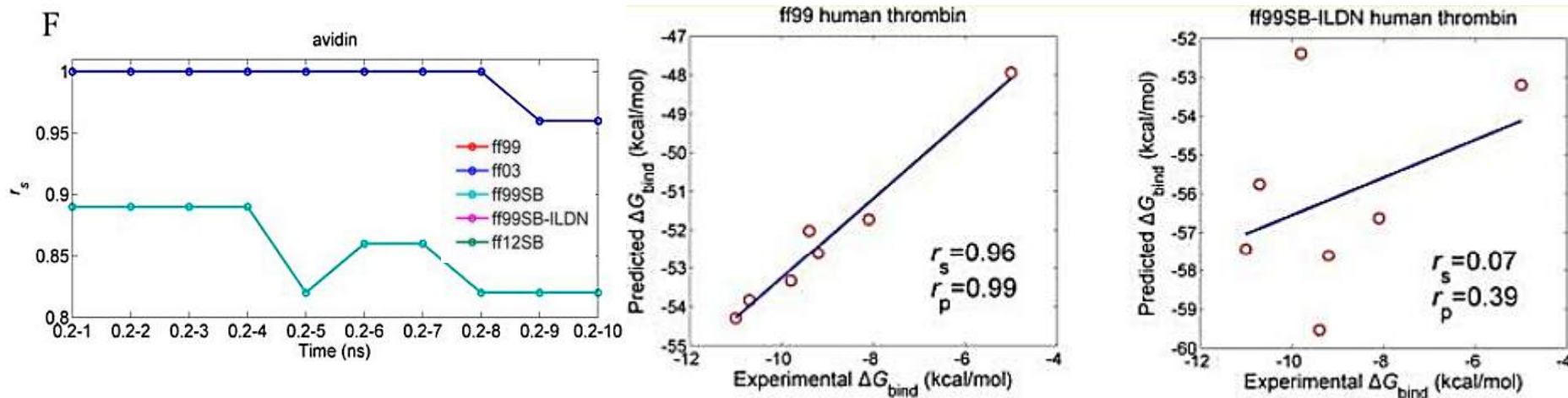
MM/GBSA		MM/PBSA	
Classification <sup>a</sup>	$r_p^b$	Classification	$r_p^b$
GB-1ns-1	$0.353 \pm 0.002^d$	PB-1ns-1	$0.071 \pm 0.002$
GB-1ns-2	$0.521 \pm 0.002$	PB-1ns-2	$0.306 \pm 0.004$
GB-1ns-4	$0.564 \pm 0.002$	PB-1ns-4	$0.491 \pm 0.003$
GB-min-1	$0.352 \pm 0.002$	PB-min-1	$-0.043 \pm 0.002$
GB-min-2	$0.535 \pm 0.002$	PB-min-2	$0.152 \pm 0.002$
GB-min-4	$0.579 \pm 0.002$	PB-min-4	$0.412 \pm 0.003$



- The predicted binding affinities by MM/GB(PB)SA are unrealistically large.
- Accuracy decreases with the increase of the ligand formal charge.
- MM/PBSA is more sensitive to the investigated systems, more suitable for the individual target prediction.



# Influence of the MM methods



- The ff99 yields better correlation with experimental  $\Delta G$  values than ff99SB-ILDN.
- For 2–4 ns MD simulations, MM/GBSA based on the ff99 force field yields the best predictions, while MM/PBSA based on the ff99SB force field does the best.
- 5 ns MD simulations may not be quite necessary.
- the RESP charges show the best performance for both MM/PBSA and MM/GBSA.

# MM/PBSA Approach for Drug Discovery

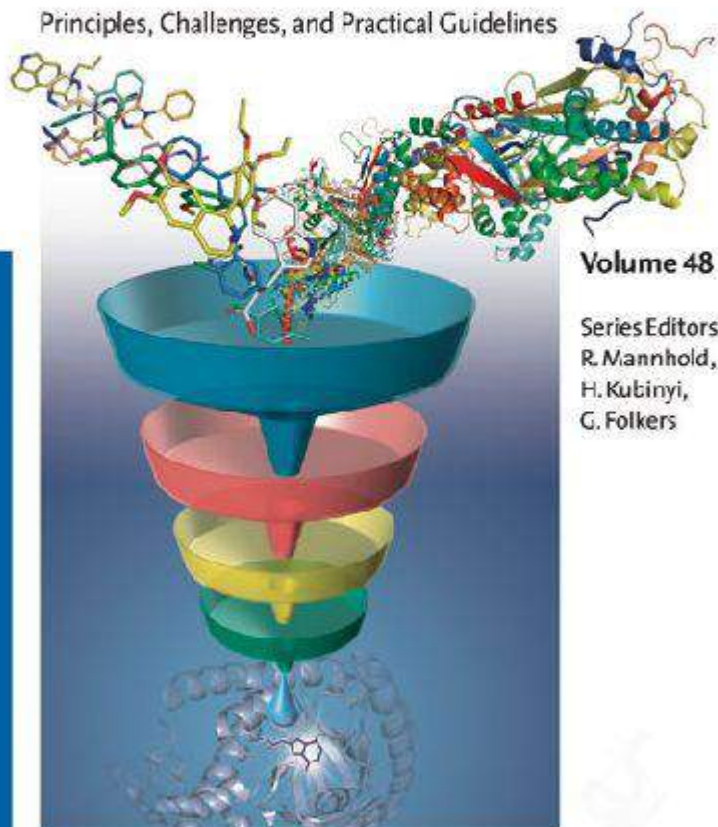
Methods and Principles in Medicinal Chemistry

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## Virtual Screening

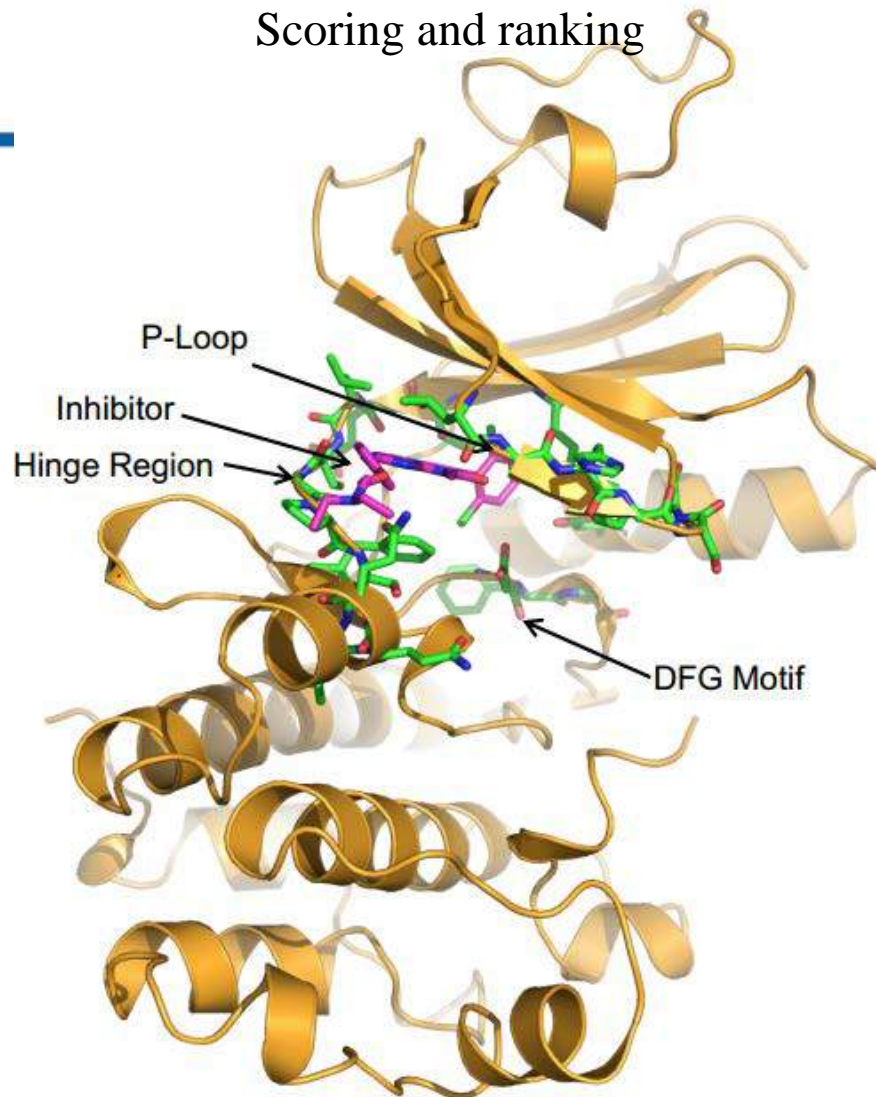
Principles, Challenges, and Practical Guidelines



Volume 48

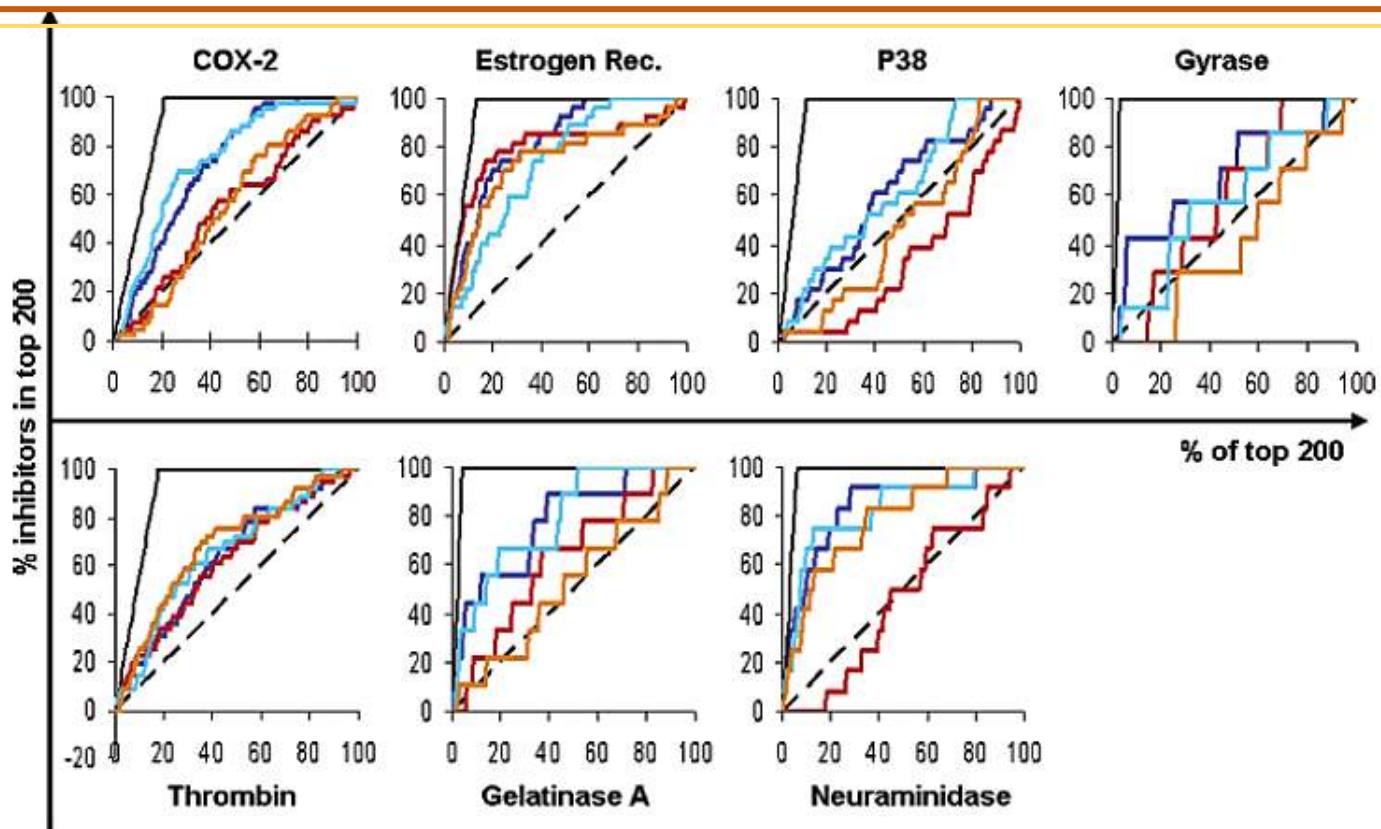
Series Editors:  
R. Mannhold,  
H. Kubinyi,  
G. Folkers

Scoring and ranking





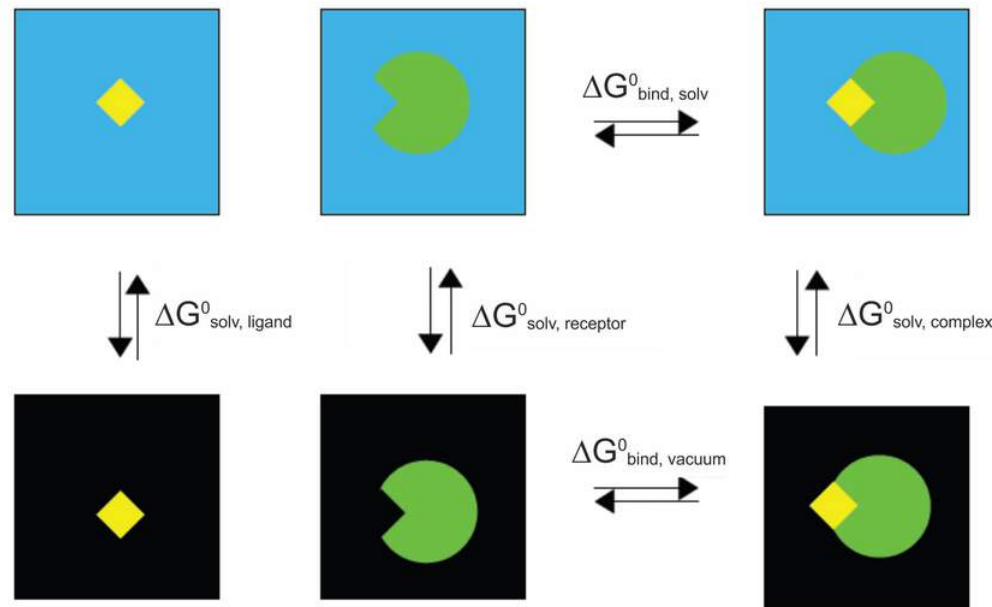
# Performance of MM/PBSA in VS



- Comparison of virtual screening performance for seven different proteins.
- Random selection (black, dashed), ideal performance (black, solid), FRED/ChemScore ranking (red), MM-RDIEL ranking with MAB\* force field (orange), MM-PBSA ranking with MAB\* force field (blue), and MM-PBSA ranking with GAFF force field (cyan).

# Summary

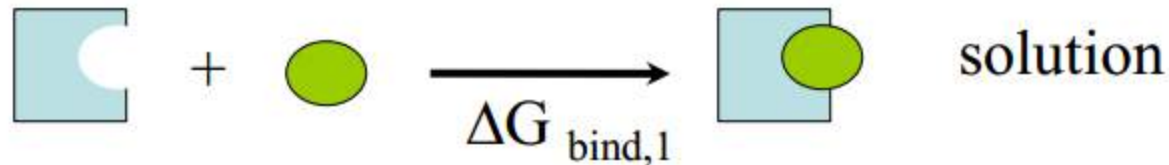
- The MM/PBSA method is a reasonable approximated free-energy calculation method, which considered the solvation energy in terms of Poisson-Boltzmann model and SASA.
- The accuracy of MM/PBSA is sensitive to the dielectric constant, the ligand formal charge, the initial protein structure and MD sampling.
- The MM/PBSA method is widely used in virtual screening, performs much better than docking and empirical scoring methods.



Thank You !!!



# One snapshot



$$\Delta \bar{G}_{bind,1} = \Delta \bar{E}_{MM} + \Delta \bar{G}_{Solv}^{PBSA} - T\Delta \bar{S} \quad (1)$$

$$\Delta E_{MM} = (E_{MM}^{complex} - E_{MM}^{Receptor} - E_{MM}^{Ligand})$$

$$\Delta G_{solvation} = (\Delta G_{solv}^{complex} - \Delta G_{solv}^{Receptor} - \Delta G_{solv}^{Ligand})$$

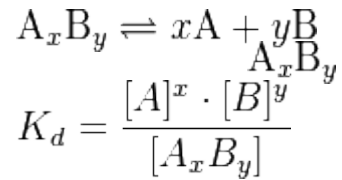
$$\Delta S = (S^{complex} - S^{Receptor} - S^{Ligand})$$

$$\bar{E}_{MM} = \bar{E}_{Bond} + \bar{E}_{Angle} + \bar{E}_{torsion} + \bar{E}_{VDW} + \bar{E}_{Elec}$$

$$\Delta G_{solv}^{PBSA} = \Delta G_{solv}^{nonpolar} + \Delta G_{solv}^{electrostatic}$$

$$\Delta \Delta G_{Bind} = \Delta G_{Bind,1} - \Delta G_{Bind,2} \quad (2)$$

$$\Delta G_{\text{vacuum}}^0 = \Delta E_{\text{molecular mechanics}}^0 - T \cdot \Delta S_{\text{normal mode analysis}}^0$$



For a general reaction:

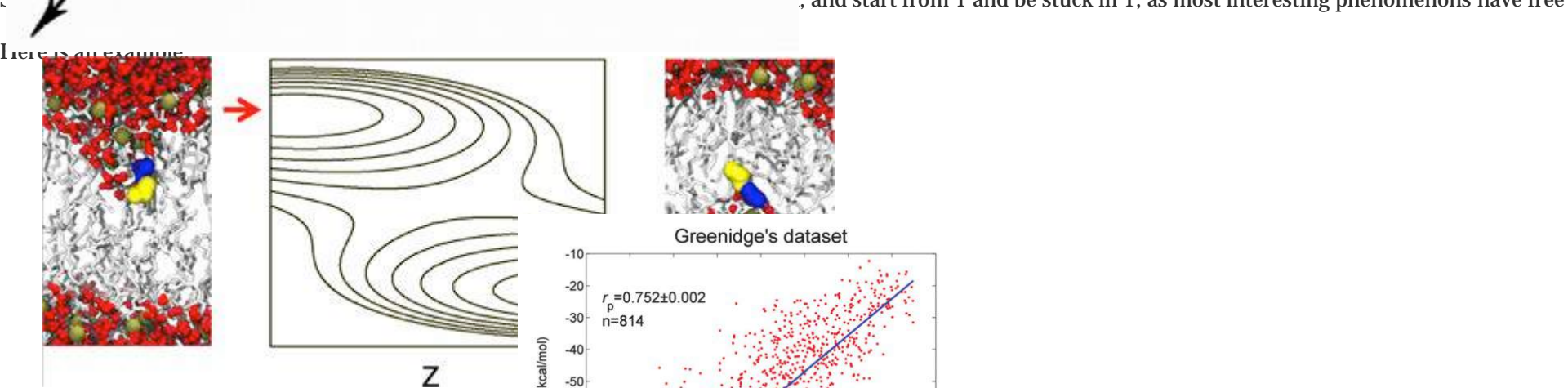
in which a complex  $A_x B_y$  breaks down into  $x$  A subunits and  $y$  B subunits, the dissociation constant is defined

Affinity is not potency, EC50

**In the MM-PBSA approach the different contributions to the binding free energy are calculated. Solvation free energies are calculated by either solving the linearised Poisson Boltzmann equation or using the Generalized Born model.**

**$\Delta G_{\text{vacuum}}$  is obtained by calculating the average interaction energy between receptor and ligand and taking the entropy contribution into account.**

**The entropy contribution can be found by performing normal mode analysis on the three species but in practice entropy is often neglected. The average interaction energies of receptor and ligand are usually obtained by performing calculations on an ensemble of structures. In this tutorial we will demonstrate the use of the MM/PB(GB)SA scripts included with Amber and AmberTools to automate these calculations.**



The above figure depicts a drug (yellow and blue) moving

How do we improve sampling?

FEP is one such method used to improve sampling, by

Change between initial and final states are controlled by

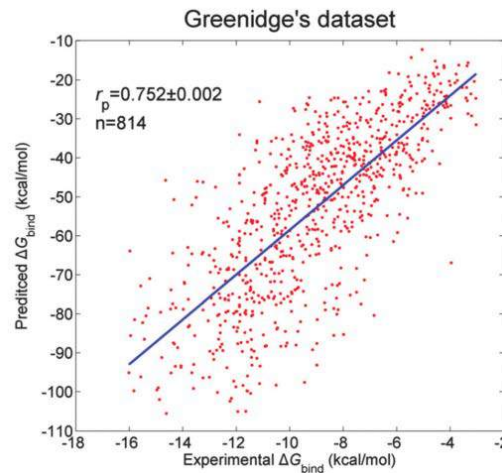
$L=0$ , could represent state X

$L=1$ , will represent state Y

$0 < L < 1$  will represent a series of intermediate states

And every potential term will be defined as a function of L:

- Bonds:  $k(\lambda) = \lambda k(Y) + (1-\lambda)k(X)$   
 $l_o(\lambda) = \lambda l_o(Y) + (1-\lambda)l_o(X)$
- Angles:  $k_\theta(\lambda) = \lambda k_\theta(Y) + (1-\lambda)k_\theta(X)$   
 $\theta_o(\lambda) = \lambda \theta_o(Y) + (1-\lambda)\theta_o(X)$
- Charges:  $q(\lambda) = \lambda q(Y) + (1-\lambda)q(X)$
- VDW:  $\epsilon(\lambda) = \lambda \epsilon(Y) + (1-\lambda)\epsilon(X)$   
 $\sigma(\lambda) = \lambda \sigma(Y) + (1-\lambda)\sigma(X)$
- etc.



representing hydrophobic core and red is water and brown, the interface

It is nothing but a step wise thermodynamic integration, by breaking