

# BGGN 213

## Structural Bioinformatics II

Lecture 12

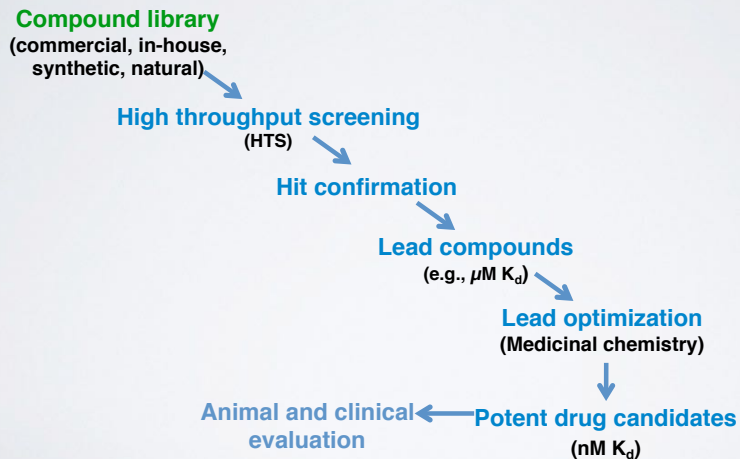
Barry Grant  
UC San Diego

<http://thegrantlab.org/bggn213>

## NEXT UP:

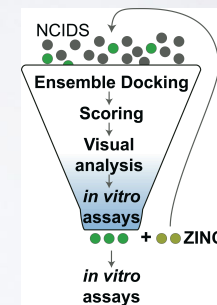
- ▶ **Overview of structural bioinformatics**
  - Major motivations, goals and challenges
- ▶ **Fundamentals of protein structure**
  - Composition, form, forces and dynamics
- ▶ **Representing and interpreting protein structure**
  - Modeling energy as a function of structure
- ▶ **Example application areas**
  - **drug discovery** & Predicting functional dynamics

## THE TRADITIONAL EMPIRICAL PATH TO DRUG DISCOVERY



## COMPUTER-AIDED LIGAND DESIGN

- Aims to reduce number of compounds synthesized and assayed
- Lower costs
- Reduce chemical waste
- Facilitate faster progress



Two main approaches:

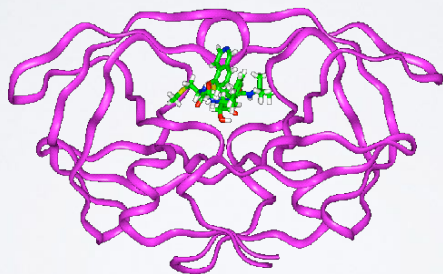
- (1). **Receptor/Target-Based**
- (2). **Ligand/Drug-Based**

Two main approaches:

- (1). **Receptor/Target-Based**
- (2). **Ligand/Drug-Based**

## SCENARIO I: RECEPTOR-BASED DRUG DISCOVERY

Structure of Targeted Protein Known: **Structure-Based Drug Discovery**

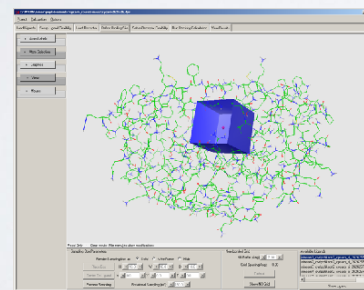


HIV Protease/KNI-272 complex

## PROTEIN-LIGAND DOCKING

Structure-Based Ligand Design

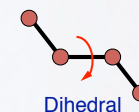
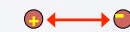
Docking software  
Search for structure of lowest energy



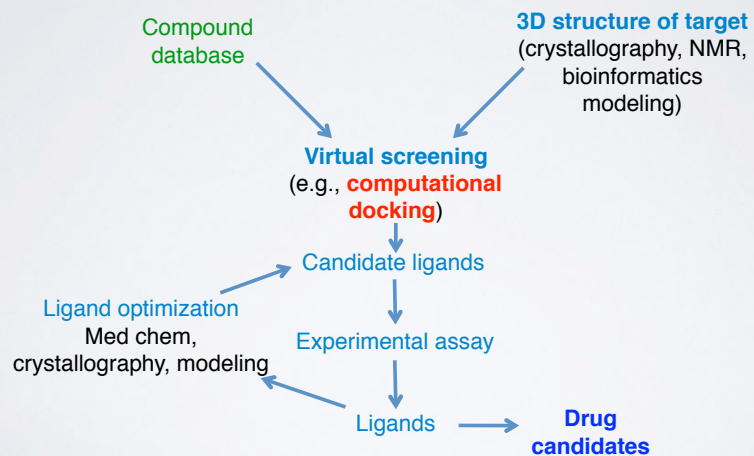
Potential function  
Energy as function of structure



Screened Coulombic



# STRUCTURE-BASED VIRTUAL SCREENING



# COMPOUND LIBRARIES

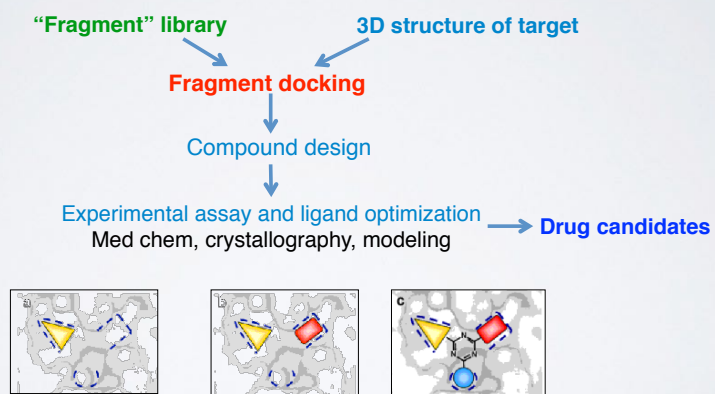


Commercial  
(in-house pharma)

Government (NIH)

Academia

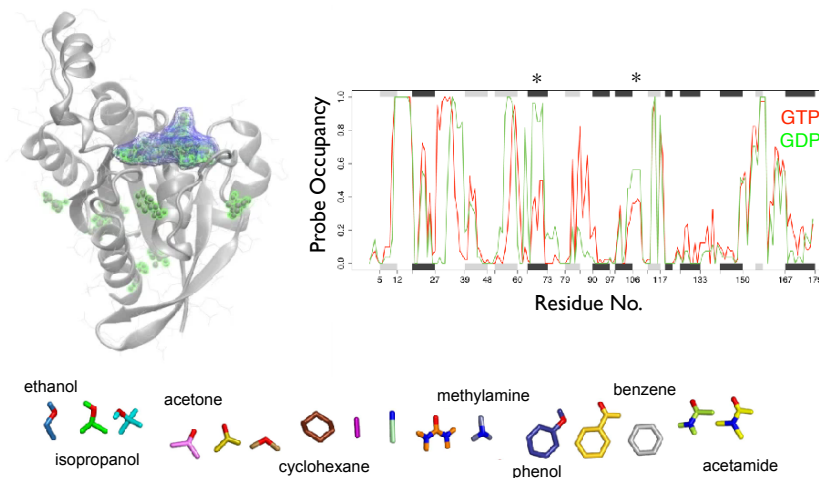
# FRAGMENTAL STRUCTURE-BASED SCREENING



<http://www.beilstein-institut.de/bozen2002/proceedings/Jhoti/Jhoti.html>

## Multiple non active-site pockets identified

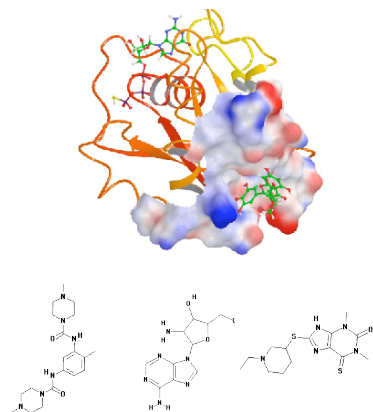
Small organic probe fragment affinities map multiple potential binding sites across the structural ensemble.



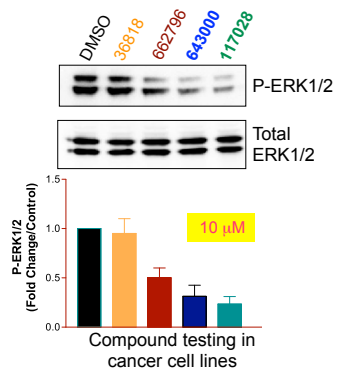
## Ensemble docking & candidate inhibitor testing

Top hits from ensemble docking against distal pockets were tested for inhibitory effects on basal ERK activity in glioblastoma cell lines.

Ensemble computational docking



Compound effect on U251 cell line



PLoS One (2011, 2012)

## COMMON SIMPLIFICATIONS USED IN PHYSICS-BASED DOCKING

Quantum effects approximated classically

Protein often held rigid

Configurational entropy neglected

Influence of water treated crudely

Two main approaches:

- (1). **Receptor/Target-Based**
- (2). **Ligand/Drug-Based**

Do it Yourself!

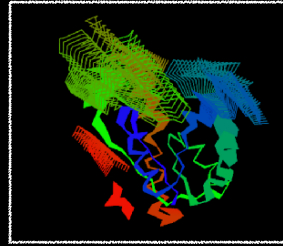
# Hand-on time!

[https://bioboot.github.io/bggn213\\_S18/lectures/#12](https://bioboot.github.io/bggn213_S18/lectures/#12)

You can use the classroom computers or your own laptops. If you are using your laptops then you will need to install **VMD** and **MGLTools**

# Bio3D view()

- If you want the 3D viewer in your R markdown you can install the development version of Bio3D



- For **MAC**:

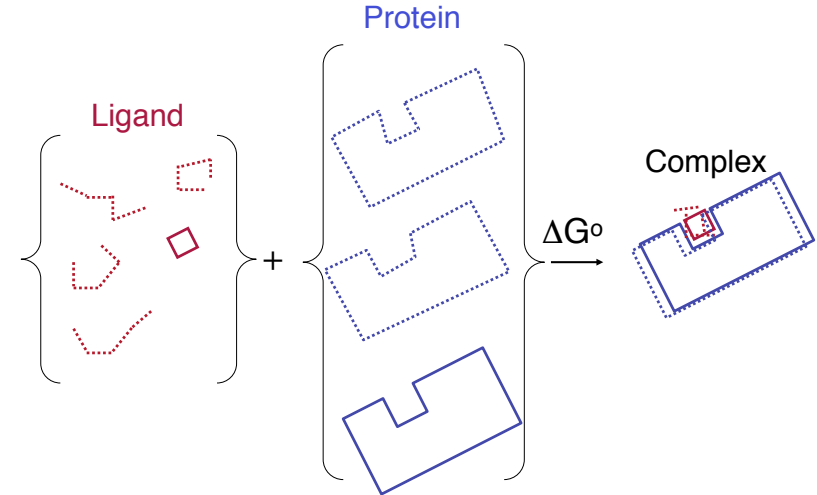
```
> download.file("https://tinyurl.com/bio3d-mac", "bio3d.tar.gz")  
> install.packages("bio3d.tar.gz", repos = NULL)
```

- For **Windows**:

```
> install.packages("https://bioboot.github.io/bggn213_S18/class-material/bio3d_2.3-4.9000.zip", repos = NULL)
```

[ See: Appendix I in Lab Sheet ]

## Proteins and Ligand are Flexible



[HTTP://129.177.232.111:3848/PCA-APP/](http://129.177.232.111:3848/PCA-APP/)

[HTTP://BIO3D.UCSD.EDU/PCA-APP/](http://BIO3D.UCSD.EDU/PCA-APP/)

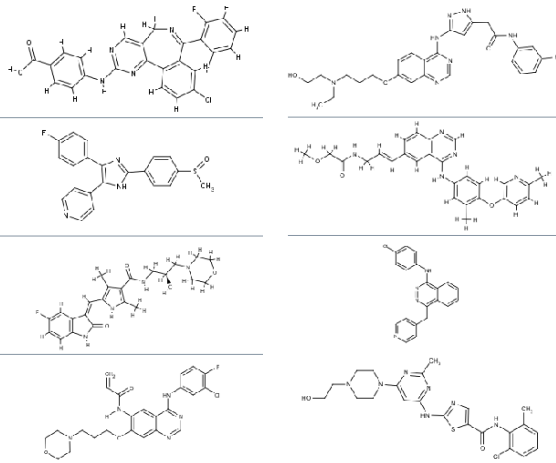
Two main approaches:

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## Scenario 2

### Structure of Targeted Protein Unknown: Ligand-Based Drug Discovery

e.g. MAP Kinase Inhibitors



Using knowledge of existing inhibitors to discover more

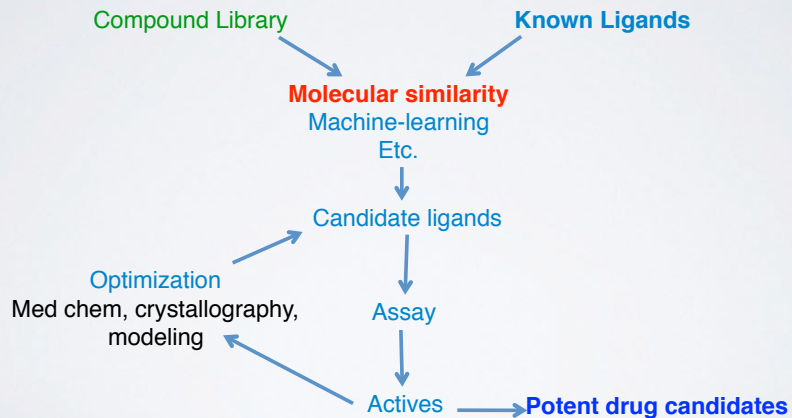
## Why Look for Another Ligand if You Already Have Some?

Experimental screening generated some ligands, but they don't bind tightly enough

A company wants to work around another company's chemical patents

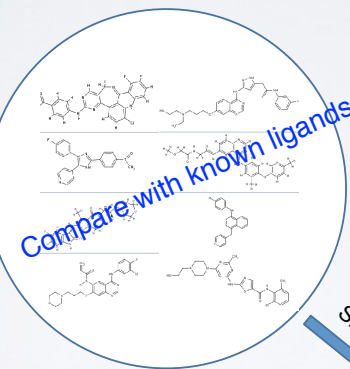
An high-affinity ligand is toxic, is not well-absorbed, difficult to synthesize etc.

## LIGAND-BASED VIRTUAL SCREENING



## CHEMICAL SIMILARITY LIGAND-BASED DRUG-DISCOVERY

Compounds  
(available/synthesizable)



Different

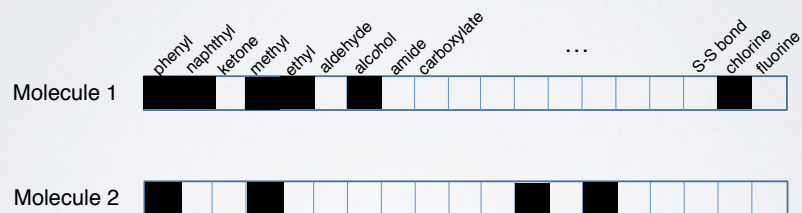
Don't bother

Similar

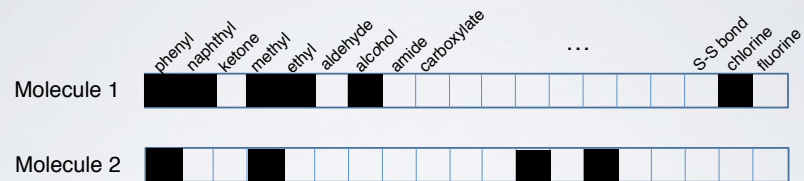
Test experimentally

# CHEMICAL FINGERPRINTS

## BINARY STRUCTURE KEYS

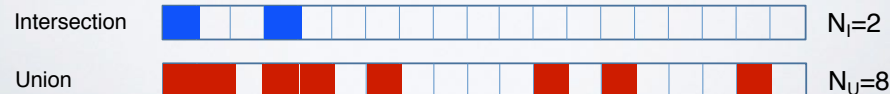


# CHEMICAL SIMILARITY FROM FINGERPRINTS



Tanimoto Similarity  
(or Jaccard Index),  $T$

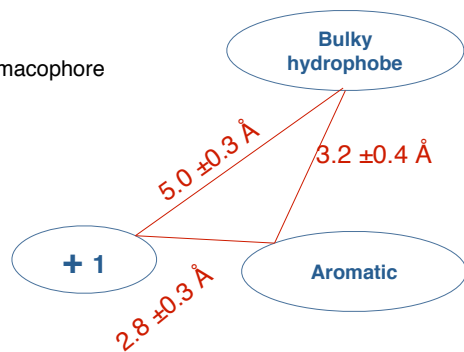
$$T \equiv \frac{N_I}{N_U} = 0.25$$



## Pharmacophore Models

Φάρμακο (drug) + Φορά (carry)

A 3-point pharmacophore

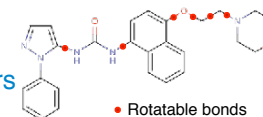


## Molecular Descriptors

More abstract than chemical fingerprints

### Physical descriptors

- molecular weight
- charge
- dipole moment
- number of H-bond donors/acceptors
- number of rotatable bonds
- hydrophobicity (log P and clogP)



### Topological

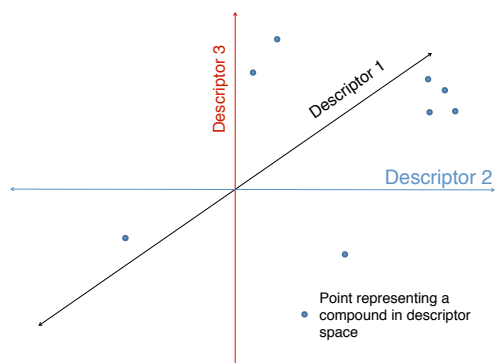
- branching index
- measures of linearity vs interconnectedness

Etc. etc.

## A High-Dimensional “Chemical Space”

Each compound is at a point in an n-dimensional space

Compounds with similar properties are near each other



Apply **multivariate statistics** and **machine learning** for descriptor-selection. (e.g. partial least squares, PCA, support vector machines, random forest, deep learning etc.)

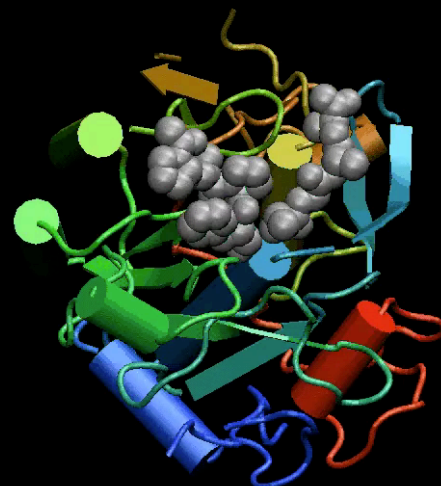
## Rules for drug discovery success

- Set of approved drugs or medicinal chemistry compounds and their targets can be used to derive rules for drug discovery success (or failure):
  - What features make a successful drug target?
  - What features make a protein druggable by small molecules?
  - What features of a compound contribute to good oral bioavailability?
  - What chemical groups may be associated with toxicity?

**Optional:**  
Stop here for Today!

[ [Muddy Point Assessment](#) ]

NMA models the protein as a network of elastic strings



Proteinase K



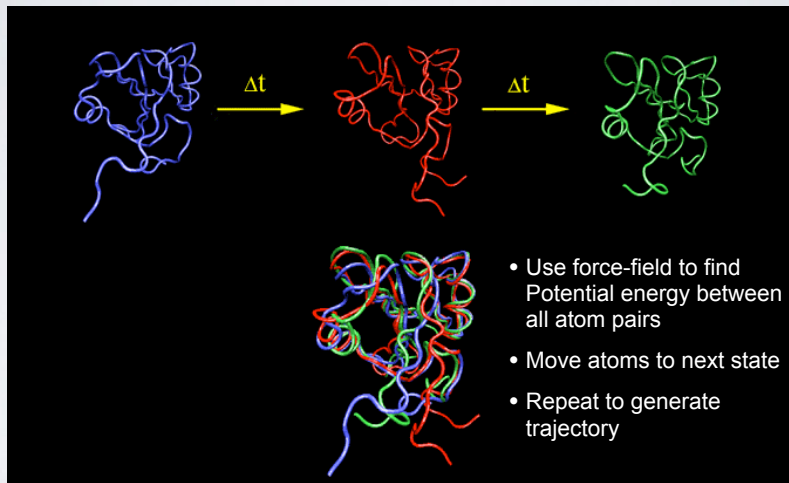
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- ▶ **Representing and interpreting protein structure**
  - Modeling energy as a function of structure
- ▶ **Example application areas**
  - Drug discovery & predicting **functional dynamics**

## PREDICTING FUNCTIONAL DYNAMICS

- Proteins are **intrinsically flexible** molecules with **internal motions that are often intimately coupled to their biochemical function**
  - E.g. ligand and substrate binding, conformational activation, allosteric regulation, etc.
- Thus knowledge of dynamics can provide a deeper understanding of the **mapping of structure to function**
  - **Molecular dynamics** (MD) and **normal mode analysis** (NMA) are two major methods for predicting and characterizing molecular motions and their properties

## MOLECULAR DYNAMICS SIMULATION



McCammon, Gelin & Karplus, *Nature* (1977)  
[ See: <https://www.youtube.com/watch?v=ui1ZysMFcKk> ]

**KEY CONCEPT:** POTENTIAL FUNCTIONS DESCRIBE A SYSTEMS **ENERGY** AS A FUNCTION OF ITS **STRUCTURE**

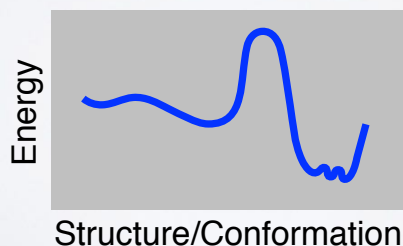
Two main approaches:

- (1). **Physics-Based**
- (2). **Knowledge-Based**

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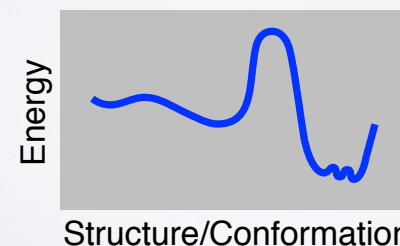
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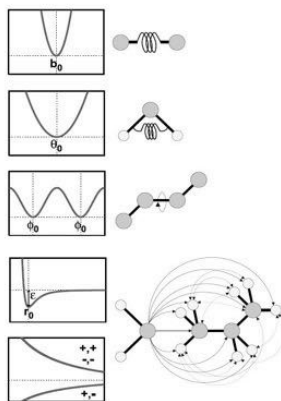
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**PHYSICS-BASED** POTENTIALS  
**ENERGY TERMS FROM PHYSICAL THEORY**

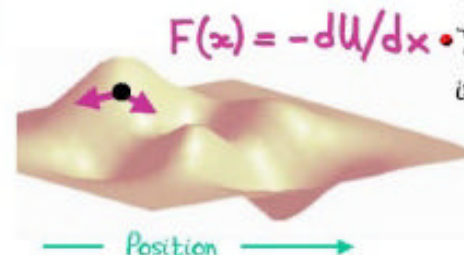
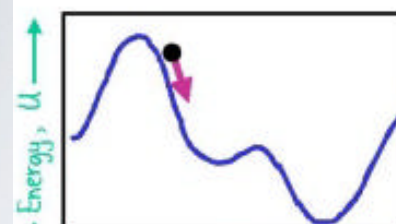
$$U(\vec{R}) = \underbrace{\sum_{\text{bonds}} k_i^{\text{bond}} (r_i - r_0)^2}_{U_{\text{bond}}} + \underbrace{\sum_{\text{angles}} k_i^{\text{angle}} (\theta_i - \theta_0)^2}_{U_{\text{angle}}} + \underbrace{\sum_{\text{dihedrals}} k_i^{\text{dihedral}} [1 + \cos(n_i \phi_i + \delta_i)]}_{U_{\text{dihedral}}} + \underbrace{\sum_i \sum_{j \neq i} A \epsilon_{ij} \left[ \left( \frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left( \frac{\sigma_{ij}}{r_{ij}} \right)^6 \right]}_{U_{\text{nonbond}}} + \sum_i \sum_{j \neq i} \frac{q_i q_j}{\epsilon r_{ij}}$$

- $U_{\text{bond}}$  = oscillations about the equilibrium bond length
- $U_{\text{angle}}$  = oscillations of 3 atoms about an equilibrium bond angle
- $U_{\text{dihedral}}$  = torsional rotation of 4 atoms about a central bond
- $U_{\text{nonbond}}$  = non-bonded energy terms (electrostatics and Lenard-Jones)



CHARMM PE. function, see: <http://www.charmm.org/>

TOTAL POTENTIAL ENERGY

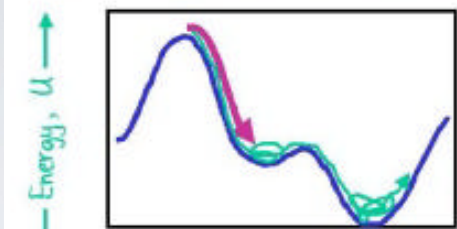


$$F(x) = -dU/dx$$

- The total potential energy or enthalpy fully defines the system,  $U$ .
- The forces are the gradients of the energy.
- The energy is a sum of independent terms for: Bond, Bond angles, Torsion angles and non-bonded atom pairs.

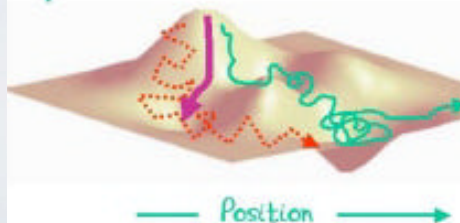
Slide Credit: Michael Levitt

## MOVING OVER THE ENERGY SURFACE



- Energy Minimization drops into local minimum.

- Molecular Dynamics uses thermal energy to move smoothly over surface.



- Monte Carlo Moves are random. Accept with probability  $\exp(-\Delta U/kT)$ .

Slide Credit: Michael Levitt

## PHYSICS-ORIENTED APPROACHES

### Weaknesses

Fully physical detail becomes computationally intractable  
 Approximations are unavoidable  
 (Quantum effects approximated classically, water may be treated crudely)  
 Parameterization still required

### Strengths

Interpretable, provides guides to design  
 Broadly applicable, in principle at least  
 Clear pathways to improving accuracy

### Status

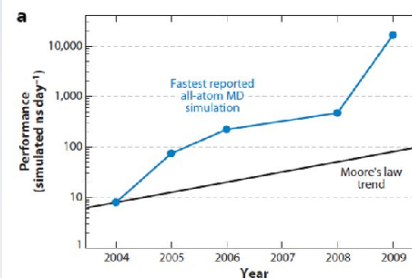
Useful, widely adopted but far from perfect  
 Multiple groups working on fewer, better approx  
 Force fields, quantum entropy, water effects  
 Moore's law: hardware improving

## HOW COMPUTERS HAVE CHANGED

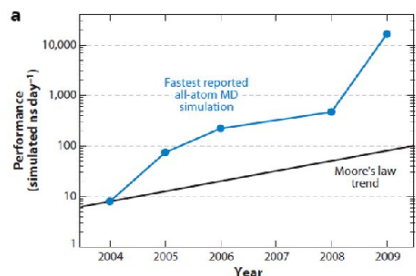
| DATE   | COST    | SPEED   | MEMORY | SIZE   |
|--------|---------|---------|--------|--------|
| 1967   | \$40M   | 0.1 MHz | 1 MB   | HALL   |
| 2013   | \$4,000 | 1 GHz   | 10 GB  | LAPTOP |
| CHANGE | 10,000  | 10,000  | 10,000 | 10,000 |

If cars were like computers then a new Volvo would cost \$3, would have a top speed of 1,000,000 km/hr, would carry 50,000 adults and would park in a shoebox

## SIDE-NOTE: GPUS AND ANTON SUPERCOMPUTER



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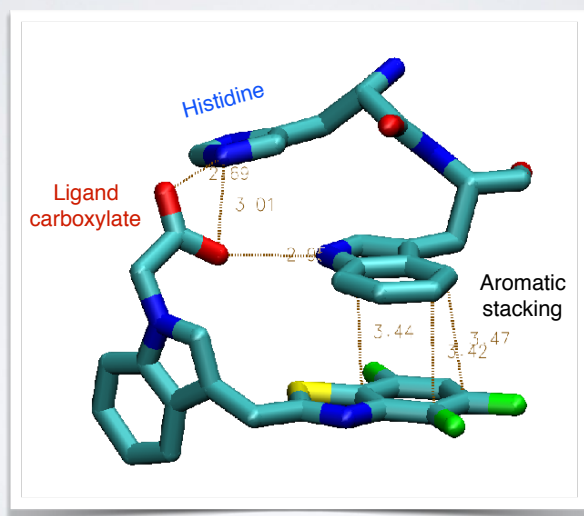


**KEY CONCEPT:** POTENTIAL FUNCTIONS DESCRIBE A SYSTEMS **ENERGY** AS A FUNCTION OF ITS **STRUCTURE**

Two main approaches:

- (1). **Physics-Based**
- (2). **Knowledge-Based**

## KNOWLEDGE-BASED DOCKING POTENTIALS



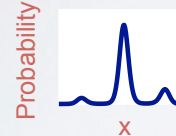
## ENERGY DETERMINES **PROBABILITY** (STABILITY)

Basic idea: Use probability as a proxy for energy



Boltzmann:  

$$p(r) \propto e^{-E(r)/RT}$$



Inverse Boltzmann:  

$$E(r) = -RT \ln[p(r)]$$

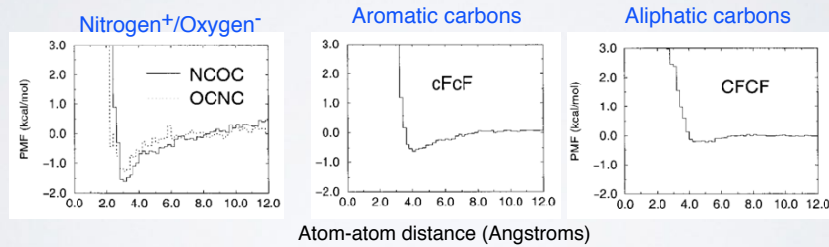
Example: ligand carboxylate O to protein histidine N

- Find all protein-ligand structures in the PDB with a ligand carboxylate O
1. For each structure, histogram the distances from O to every histidine N
  2. Sum the histograms over all structures to obtain  $p(r_{O-N})$
  3. Compute  $E(r_{O-N})$  from  $p(r_{O-N})$

# KNOWLEDGE-BASED DOCKING POTENTIALS

“PMF”, Muegge & Martin, J. Med. Chem. (1999) 42:791

A few types of atom pairs, out of several hundred total



$$E_{\text{prot-lig}} = E_{\text{vdw}} + \sum_{\text{pairs } (ij)} E_{\text{type}(ij)}(r_{ij})$$

# KNOWLEDGE-BASED POTENTIALS

## Weaknesses

Accuracy limited by availability of data

## Strengths

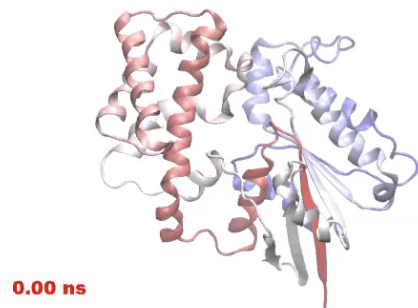
Relatively easy to implement  
Computationally fast

## Status

Useful, far from perfect  
May be at point of diminishing returns  
(not always clear how to make improvements)

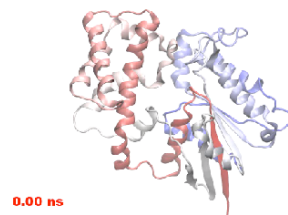
# MD Prediction of Functional Motions

Accelerated MD simulation of nucleotide-free transducin alpha subunit

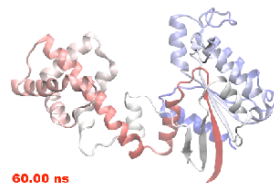


Yao and Grant, Biophys J. (2013)

“close”

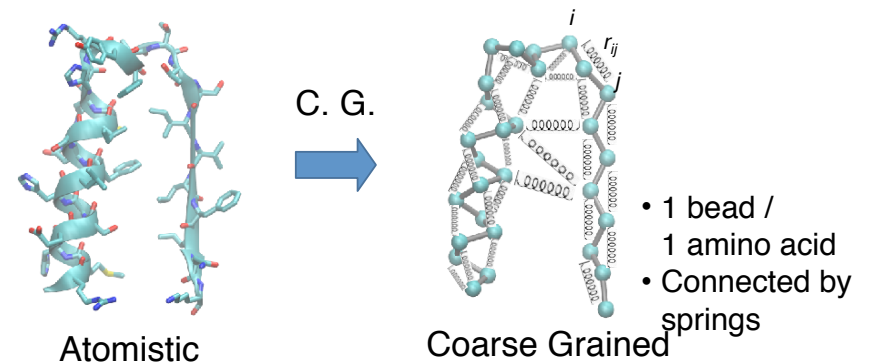


“open”

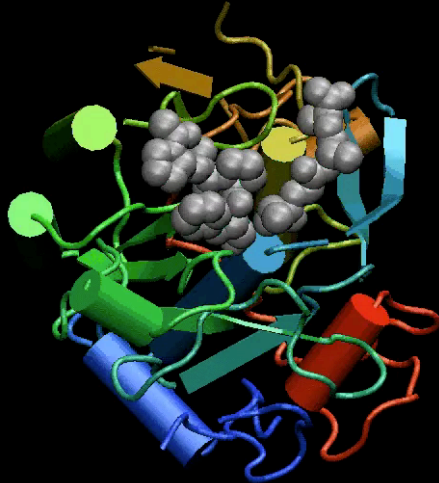


# COARSE GRAINING: **NORMAL MODE ANALYSIS** (NMA)

- MD is still time-consuming for large systems
- Elastic network model NMA (ENM-NMA) is an example of a lower resolution approach that finishes in seconds even for large systems.



NMA models the protein as a network of elastic strings



Proteinase K

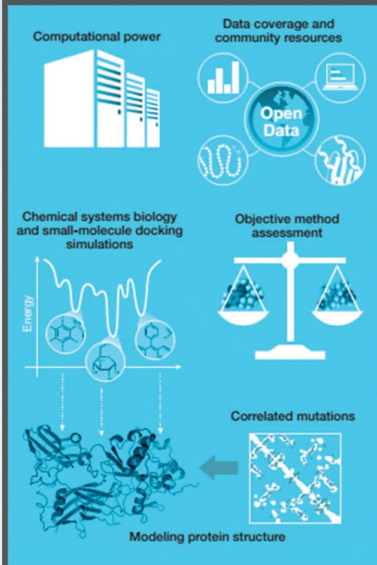
Do it Yourself!

# Hand-on time!

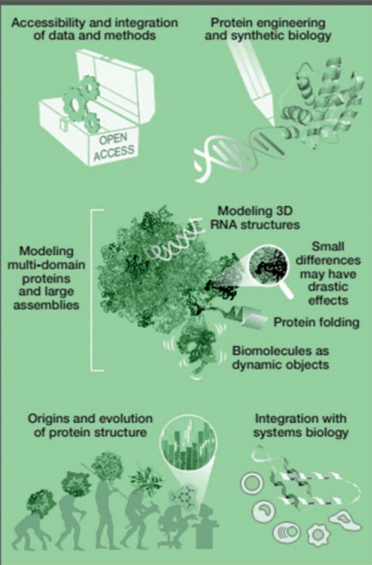
[https://bioboot.github.io/bggn213\\_S18/lectures/#12](https://bioboot.github.io/bggn213_S18/lectures/#12)

Focus on **section 3 & 4** exploring **PCA** and **NMA apps**

## ACHIEVEMENTS

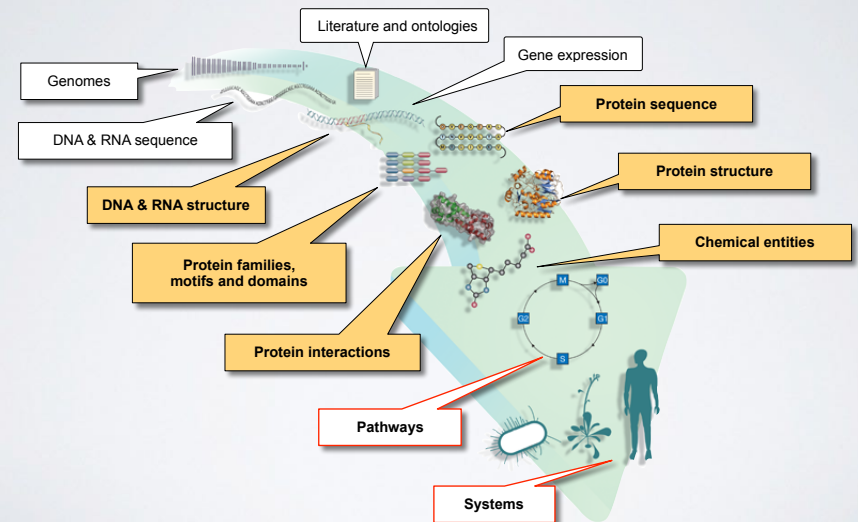


## CHALLENGES



Ilan Samish et al. *Bioinformatics* 2015;31:146-150

## INFORMING SYSTEMS BIOLOGY?



## SUMMARY

- Structural bioinformatics is computer aided structural biology
- Described major motivations, goals and challenges of structural bioinformatics
- Reviewed the fundamentals of protein structure
- Introduced both physics and knowledge based modeling approaches for describing the structure, energetics and dynamics of proteins computationally
- Introduced both structure and ligand based bioinformatics approaches for drug discovery and design