# Small Molecule and Protein Docking

### Introduction

A significant portion of biology is built on the paradigm

sequence → structure → function

 As we sequence more genomes and get more structural information, the next challenge will be to predict interactions and binding for two or more biomolecules (nucleic acids, proteins, peptides, drugs or other small molecules)

### Introduction

- The questions we are interested in are:
  - Do two biomolecules bind each other?
  - If so, how do they bind?
  - What is the binding free energy or affinity?
- The goals we have are:
  - Searching for lead compounds
  - Estimating effect of modifications
  - General understanding of binding

— ...

### Rationale

- The ability to predict the binding site and binding affinity of a drug or compound is immensely valuable in the area or pharmaceutical design
- Most (if not all) drug companies use computational methods as one of the first methods of screening or development
- Computer-aided drug design is a more daunting task, but there are several examples of drugs developed with a significant contribution from computational methods

### **Examples**

- Tacrine inhibits acetylcholinesterase and boost acetylcholine levels (for treating Alzheimer's disease)
- Relenza targets influenza
- Invirase, Norvir, Crixivan Various HIV protease inhibitors
- Celebrex inhibits Cox-2 enzyme which causes inflammation (not our fault)

## Docking

- Docking refers to a computational scheme that tries to find the best binding orientation between two biomolecules where the starting point is the atomic coordinates of the two molecules
- Additional data may be provided (biochemical, mutational, conservation, etc.) and this can significantly improve the performance, however this extra information is not required

### Bound vs. Unbound Docking

- The simplest problem is the "bound" docking problem. The goal in this case is to reproduce a known complex where the starting point is atomic structures from a co-crystal
- The "unbound" docking problem is significantly more difficult task. Here the starting point is structure in their unbound conformation, perhaps the native structure, perhaps a modeled structure, etc.

### Approach to the Problem

- One of the first suggestions on how to tackle this problem was given by Crick who postulated that "knobs" could be fitted into "holes" on the protein surface
- Lee & Richards had already described how to treat the van der Waals surface of a protein, however the van der Waals surface was not appropriate for docking purposes (too many crevices)
- Connolly showed how to calculate the molecular surface (and also the solvent accessible surface) which allowed for many docking programs to start being developed in the mid 1980's

### A Matter of Size

- For two proteins, the docking problem is very difficult since the search space (all possible relative conformations) is extremely large
- In the case of a small molecule (drug, peptide or ligand) binding to a protein, we have a chance of exploring the conformational space, at least for the small molecule
- Now we want to consider the case where we have limited or no a priori knowledge and the mode of binding

## **Docking Methodology**

- All small molecule docking programs have three main components
  - Representation of system
  - The search algorithm
  - The scoring or energy evaluation routine
- In order to do a good job of docking, you need to search efficiently and evaluate the energy accurately

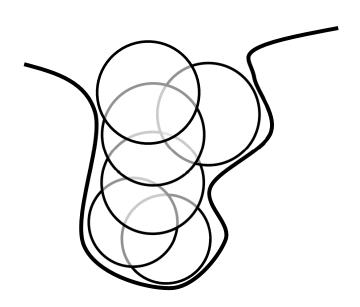
# Representation of the System or Site Characterization

### Representation of Phase Space

- The type or choice of representation for a system is really a reflection of the type of energy evaluation or scoring function that will be used
- If we would choose the most straightforward or logical representation of the atomic coordinates of the of the two biomolecules, we would simply use a molecular mechanics force field such as Amber, Charmm or OPLS
- However, this may not be the best choice in terms of computational efficiency or practicality

### **DOCK**

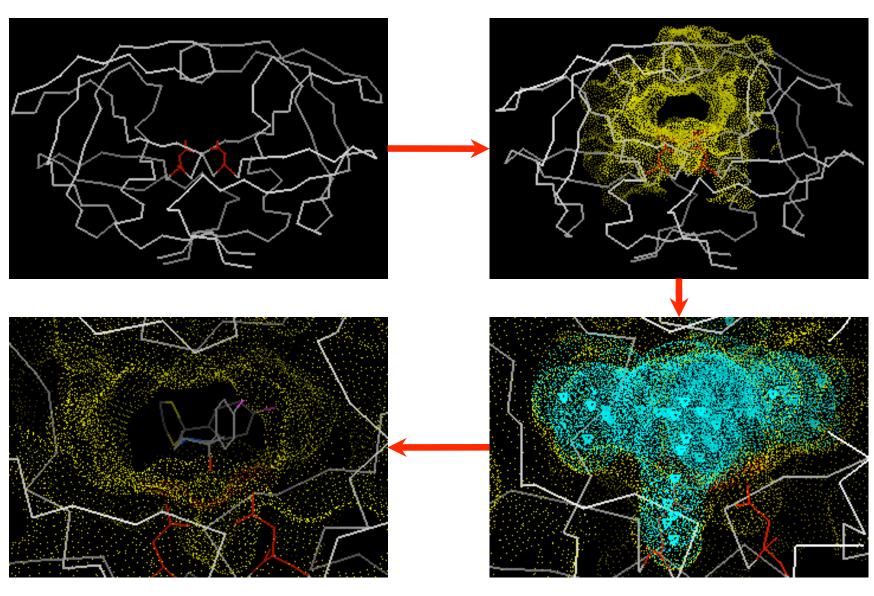
- The DOCK program is from the Kuntz group at UCSF
- It was the first docking program developed in 1982
- It represents the (negative image of the) binding site as a collection of overlapping spheres



### **DOCK**

- This method of a negative image is targeted at finding complexes with a high degree of shape complementarity
- The ligand is fit into the image by a least squares fitting of the atomic positions to the sphere centers
- Creating the negative image is obviously not a problem with a unique solution. Hence, factors such as the sphere radius and center-to-center distance of the spheres must be carefully controlled.

### **DOCK**



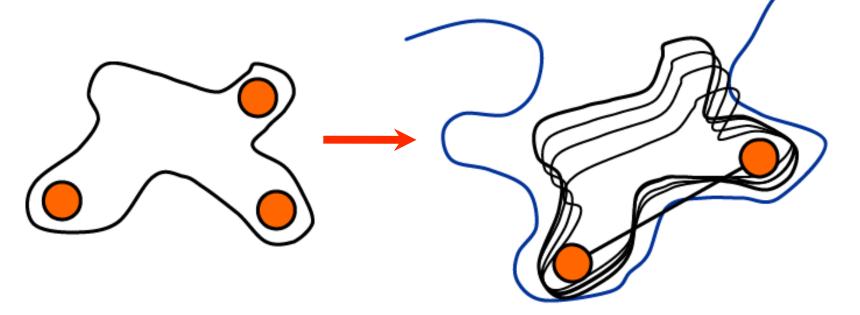
http://dock.compbio.ucsf.edu

### **DOCK Extensions**

- Dock 4.0 has recently been extended to include:
  - Several different scoring schemes
  - Ligand flexibility (via incremental construction and fragment joining)
  - Chemical properties of receptor (each sphere assume a chemical characteristic)

### CLIX

 CLIX uses a chemical description of the receptor and distance constraints on the atoms

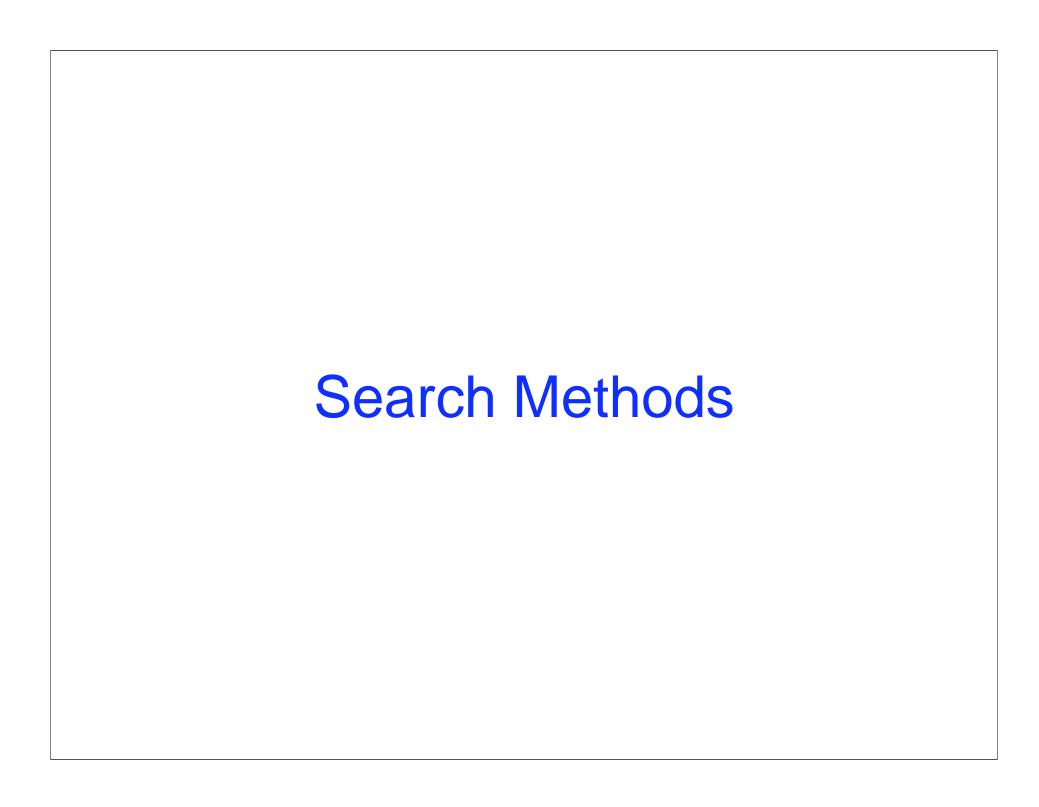


Lawrence et al., Proteins: Struct. Func. And Gen. 12:31 (1992)

### **ESCHER**

- ESCHER uses the solvent accessible surface from a Connolly algorithm
- This surface is cut into 1.5 Å slabs that are transformed into polygons and used for rigid docking (again image matching)

G. Ausiello, G. Cesareni, M. Helmer Citterich, "ESCHER: a new docking procedure applied to the reconstruction of protein tertiary structure", *Proteins*, **28**:556-567 (1997)



### **Search Basics**

- One of the more difficult tasks in computational docking is simply enumerating the number of ways two molecules can be put together (3 translational plus 3 rotational degrees of freedom)
- The size of this search space grows exponentially as we increase the size of the molecules, however this is still only for rigid structures

### Search Difficulties

- If we allow flexibility of one or both of the binding partners, this quickly becomes an intractable problem
- If we manage to solve/circumvent the problem of flexibility, we then want to be able to screen large databases of structures or drugs, so our troubles are again compounded

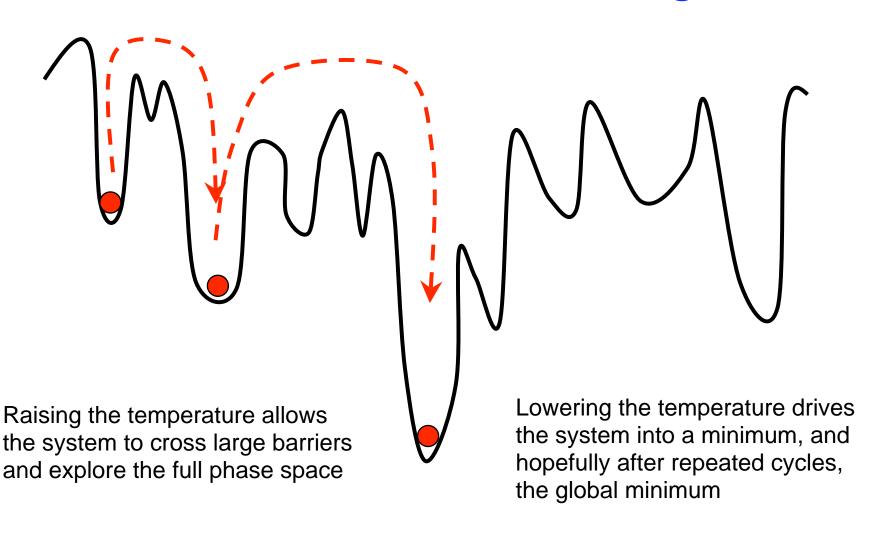
### Monte Carlo Simulated Annealing

- Also known as Metropolis Monte Carlo
- The basic steps are
  - 1. The ligand performs a random walk around the protein
  - 2. At each step, a random displacement, rotation, etc. is applied and the energy is evaluated and compared to the previous energy
  - 3. If the new energy is lower, the step is accepted
  - 4. If the energy is higher, the step is accepted with a probability  $\exp(\Delta E/kT)$

## Simulated Annealing

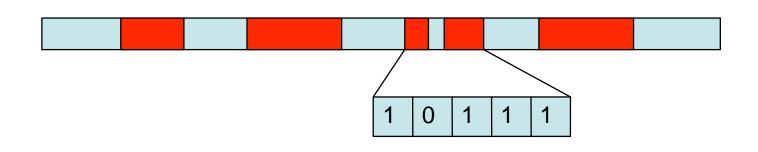
- If we were to perform this at a constant temperature, it would be basic Monte Carlo
- In this case, after a specified number of steps, the temperature is lowered and the search repeated
- As the temperature continues to go down, steps which increase the energy become less likely and the system moves to the minimum (or minima)

### Simulated Annealing



## Genetic Algorithms

- Genetic algorithms belong to a class of stochastic search methods, but rather than operating on a single solution, they function on a population
- As implied by the name, you must encode your solution, structure or problem as a genome or chromosome
- The translation, orientation and conformation of the ligand is encoded (the state variables)



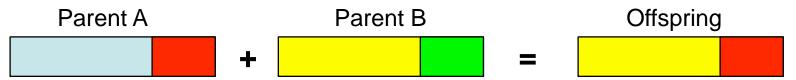
## Genetic Algorithms

- The algorithm starts by generating a population of genomes and then applies crossover and mutation to the individuals to create a new population
- The "fitness" of each structure has to be evaluated, in our case by our estimate of the binding free energy
- The best member or members survive to the next generation
- This procedure is repeated for some number of generations or energy evaluations

### **Crossover and Mutation**

There are several possible methods of crossover

### **Single Point Crossover**

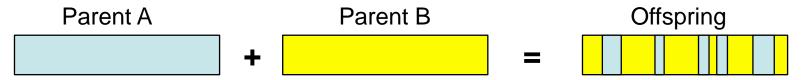


### **Two Point Crossover**



### **Crossover and Mutation**

### **Uniform Crossover**



For mutation, selected bits are simply inverted



### Efficiency

- Random mutations are also possible (perhaps the orientation or position is shifted by some random amount)
- Binary crossover and mutation can introduce inefficiencies into the algorithm since they can easily drive the system away from the region of interest
- For this reason, genetic searches are often combined with local searches (producing a Lamarckian Genetic Algorithm)

### **Local Searches**

- In a Lamarckian GA, the genetic representations of the ligands starting point is replaced with the results of the local search
- The mapping of the local search back on to the genetic representation is a biologically unrealistic task, but it works well and makes the algorithm more efficient
- One of the more common local search methods is the Solis and Wets algorithm

# Solis and Wets Algorithm

```
Starting point x
Set bias vector b to 0
Initialize \rho
```

While max iterations not exceeded:

Add deviate d to each dimension (from distribution with width  $\rho$ )

If new solution is better:

success++ b = 0.4 d + 1.2 b

else

failures++ b = b - 0.4 d

If too much success: increase  $\rho$ 

If too many failures: decrease  $\rho$ 

Adaptive step size which biases the search in the direction of success

# LGA Scheme Start Generate Local Search **Population** Mutation and Pick Top Member(s) Crossover Final Structure

### Approaches to Flexibility

- A relatively simple molecule with 10 rotatable bonds has more than 10<sup>9</sup> possible conformation if we only consider 6 possible positions for each bond
- Monte Carlo, Simulated Annealing and Genetic Algorithm can help navigate this vast space
- Other methods have been developed to again circumvent this problem

## **Flexibility**

- Some algorithms (call Place & Join algorithms) break the ligand up into pieces, dock the individual pieces, and try and reconnect the bound conformations
- FlexX uses a library of precomputed, minimized geometries from the Cambridge database with up to 12 minima per bond. Sets of alternative fragments are selected by choosing single or multiple pieces in combination
- Flexible docking via molecular dynamics with minimization can handle arbitrary flexibility, however it is extremely slow

## **Hybrid Methods**

- Some of the newer methods use a hybrid scheme of a quick docking using a GA or other scheme followed by molecular dynamics to refine the prediction
- These methods are likely to be some of the best once they are correctly parameterized, etc.

# **Scoring Functions**

# Scoring

- We covered some of the most common representations of the system as well as the most commonly used search methods
- All of these search methods involve evaluating the "fitness" or "energy" of a given binding conformation
- In order to do this effectively, we must have a good scoring function that can give an accurate estimate of the binding free energy (or relative free energy)

#### **Practical Considerations**

- If we have developed an efficient search algorithm, it may produce 10<sup>9</sup> or more potential solutions
- Many solutions can be immediately eliminated due to atomic clashes or other obvious problems, but we still must evaluate the fitness of a large number of structures
- For this reason, our scoring function must not only accurate, but it must be fast and efficient

# **Scoring Accuracy**

- If we were scoring a single ligand-protein complex, we could adopt much more sophisticated methods to arrive at an accurate value for the binding free energy
- Requiring true accuracy in a scoring function is not a realistic expectation, however there are two features that a good scoring function should possess
- When docking a database of compounds, a good scoring function should
  - Give the best rank to the "true" bound structure
  - Give the correct relative rank of each ligand in the database
  - And again, it must be able to do these things relatively quickly

### Types of Scoring Functions

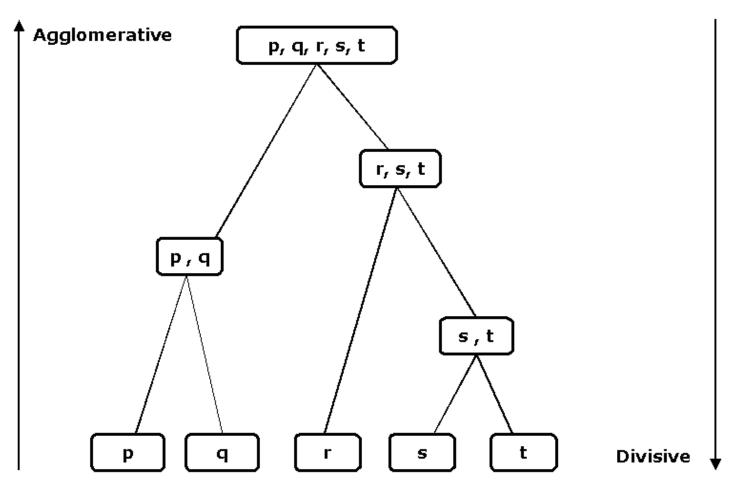
- There are several types of scoring functions that we have discussed previously
  - First Principles Methods
  - Semiempirical methods
  - Empirical methods
  - Knowledge based potentials
- All of these are used by various docking programs

# Clustering

- No docking algorithm can produce a single, trustworthy structure for the bound complex, but instead they produce an ensemble of predictions
- Each predicted structure has an associated energy, but we need to consider both the binding free energy (or enthalpy as the case may be) as well as the relative population
- By clustering our data based on some "distance" criteria, we can gain some sense of similarity between predictions
- The distance can be any of a number of similarity measures, but for 3D structures, RMSD is the standard choice

#### Clustering

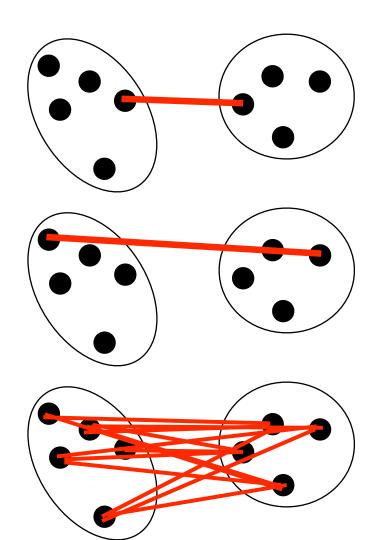
 One of the more commonly used methods is hierarchical clustering



# Agglomerative Clustering

- In the agglomerative hierarchical clustering method, all structures begin as individual clusters, and the closest clusters are subsequently (iteratively) merged together
- There are again several different methods of measuring the distance between clusters
  - Simple linkage
  - Complete linkage
  - Group average

### Linkage Criterion



Simple linkage selects the distance as the gives the minimum distance between any two members

Complete linkage selects the maximum distance between any two members

Group average approaches vary in effectiveness depending on the nature of the data

#### **Example: Autodock**

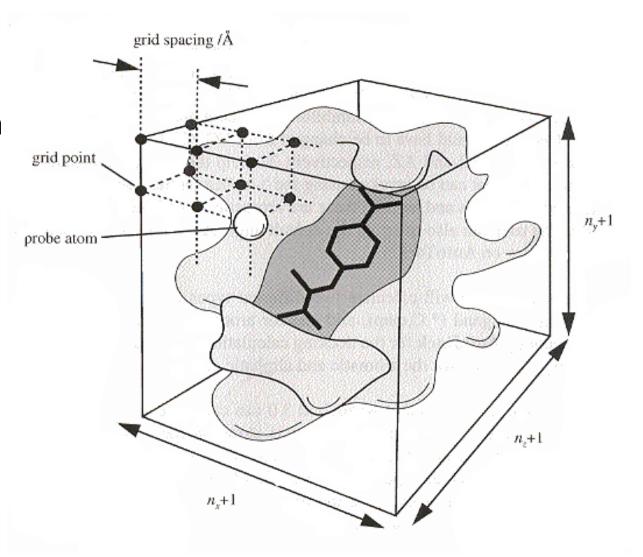
- Autodock uses pre-calculated affinity maps for each atom type in the substrate molecule, usually C, N, O and H, plus an electrostatic map
- These grids include energetic contributions from all the usual sources

$$\Delta G = G_1 \sum_{i,j} \left( \frac{A_{ij}}{r_{ij}^{12}} - \frac{B_{ij}}{r_{ij}^{6}} \right) + G_2 \sum_{i,j} \left( \frac{C_{ij}}{r_{ij}^{12}} - \frac{D_{ij}}{r_{ij}^{10}} + E_{hbond} \right)$$

$$+ G_3 \sum_{i,j} \frac{q_i q_j}{\epsilon(r) r_{ij}} + G_4 \Delta G_{tor} + G_5 \sum_{i,j} S_i V_j e^{-r_{ij}^2/2\sigma^2}$$

### **Grid Maps**

Each type of atom is placed at each individual grid point and the change in free energy is calculated



## Other Energetic Notes

- Also includes torsional energy (if the ligand is flexible)
- Entropic contributions due to rotatable bonds
- Desolvation is calculated, but only for heavy atoms on the marcromolecule

#### Search Methods

- Autodock has three search methods
  - Monte Carlo simulated annealing
  - A global genetic algorithm search
  - A local Solis and Wets search
- Combining the last two search methods gives the Lamarckian Genetic Algorithm which is what we typically use

### Methodology

- Require structures for both ligand and macromolecule
- Add charges to both structures (create a pdbq file)
  - For proteins or polypeptides we use the Kollman united atom model
  - For drugs and other molecule, we use Gasteiger charges

### Methodology

- For the macromolecule, solvation needs to be calculated (producing a pdbqs file)
- Based on the atom types in the ligand, the appropriate maps need to be calculated for the macromolecule
- Flexible bonds (if any) need to be assigned to the ligand

#### Protein-Protein Docking Methods

- The approaches to protein-protein docking have a lot in common with small molecule docking
- The methods are still based on the combination of search algorithm coupled to a scoring function
- The scoring functions are essentially the same (since we are still dealing with atomic interactions), however the major problem is that the conformational space we now need to search is massive

#### **FFT Methods**

- A clever method developed by Katchalski-Katzir et al. uses FFT methods to search the translational and rotational degrees of freedom
- H. A. Gabb, R. M Jackson and M. Sternberg "Modelling protein docking using shape complementarity, electrostatics and biochemical information"
  - J. Mol Biol. (1997) 272:106-120

#### **CAPRI**

- Just like the CASP competition in the protein folding field, there is a bi-annual competition capped CAPRI: the Critical Assessment of PRedicted Interactions
- J. Janin et al. "CAPRI: a Critical Assessment of Predicted Interactions" Proteins (2003) 52:2-9
- Mendez et al. "Assessment of blind predictions of protein-protein interactions: Current status of docking methods" Proteins (2003) 52:51-67