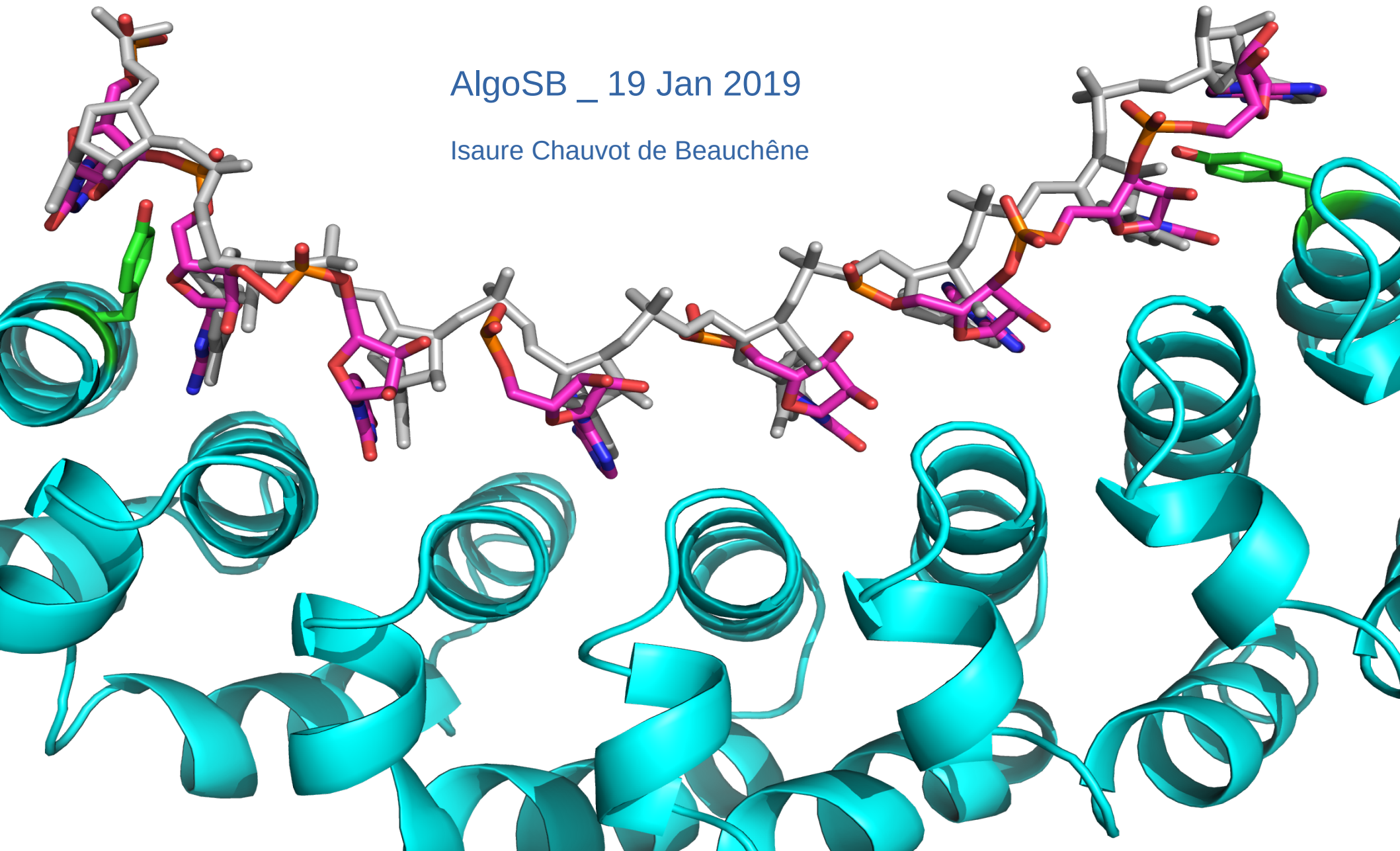


Modeling Protein- RNA complexes

AlgoSB _ 19 Jan 2019

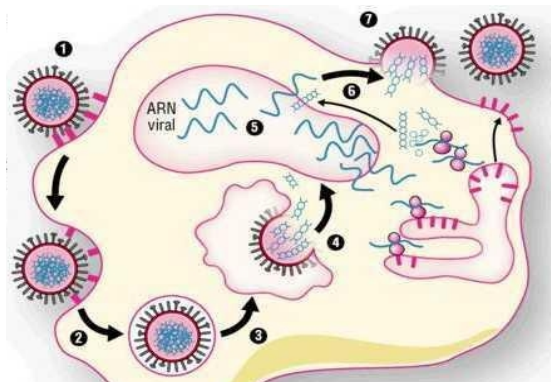
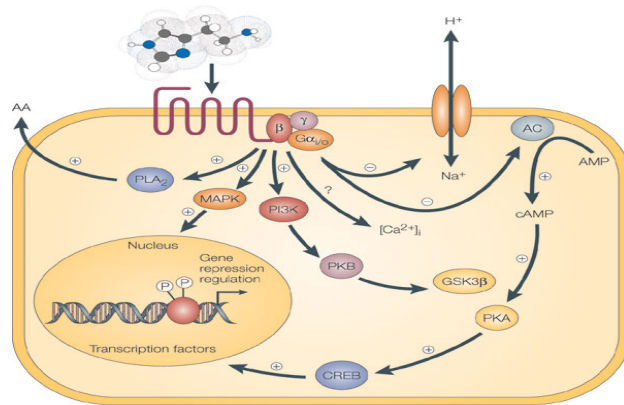
Isaure Chauvot de Beauchêne



Protein – RNA interactions

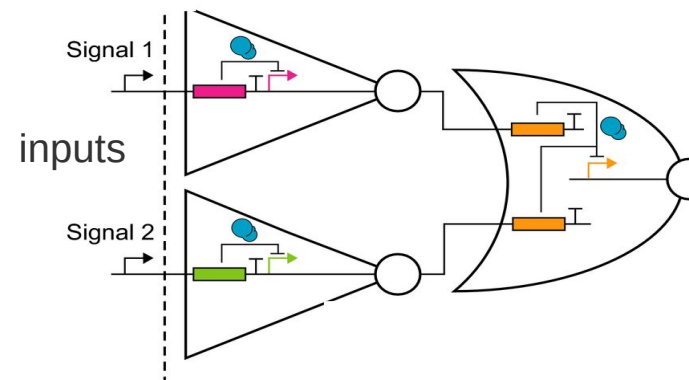
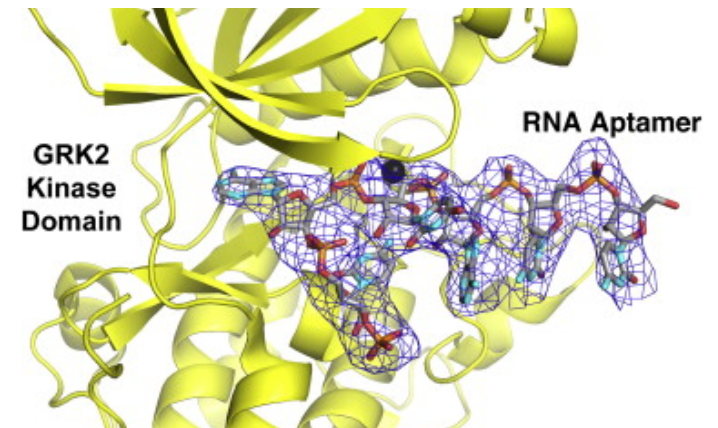
Many RNA functions require interactions with proteins:

- Genes regulation
- Signal transduction
- Protein synthesis
- Replication of RNA viruses



A 3D representation of the interface allows to:

- Understand protein/RNA function
- Predict mutations effect
- Compute binding energy
- Design drugs to target the interface
- Design a protein to target an RNA
- Design an RNA to target a protein („aptamer“)



Protein – RNA interfaces : Experimental data

Main features

The docking problem: Sampling
Evaluation of docking models
Scoring

Flexibility: Flexible docking
Fragment-based docking

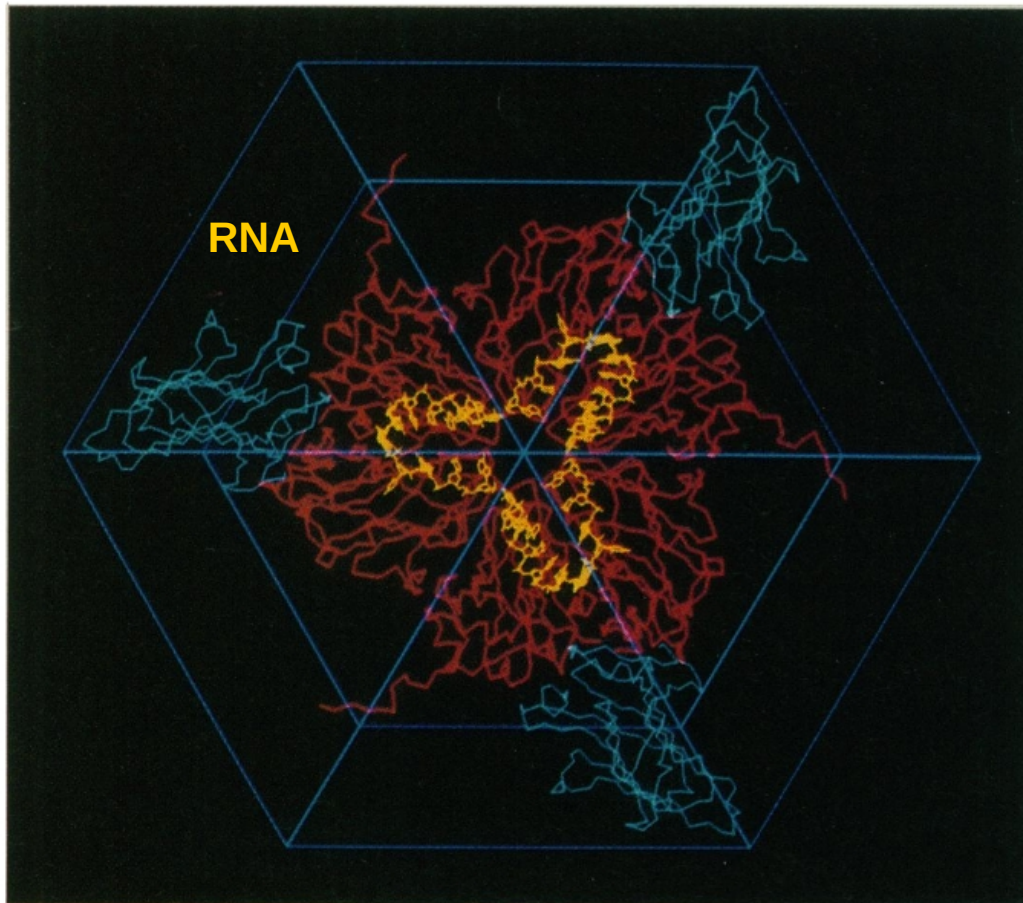
Data-driven docking: Contact/interface -driven
Fitting in 3D shape

Other docking paradigms

Modeling pipeline

Protein-RNA Interactions in an Icosahedral Virus at 3.0 Å Resolution

ZHONGGUO CHEN, CYNTHIA STAUFFACHER, YUNGE LI, TIM SCHMIDT, WU BOMU,
GREG KAMER, MICHAEL SHANKS, GEORGE LOMONOSSOFF, JOHN E. JOHNSON*

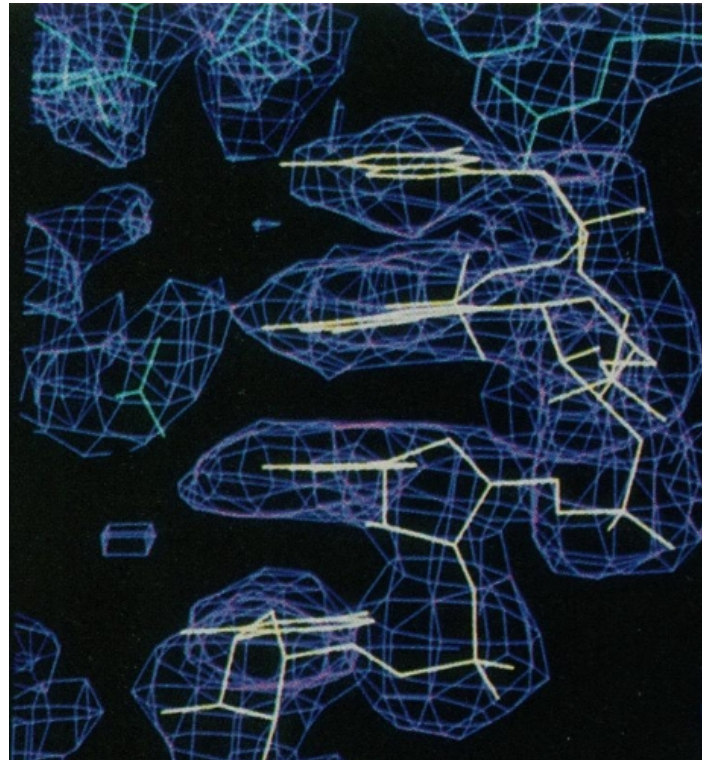
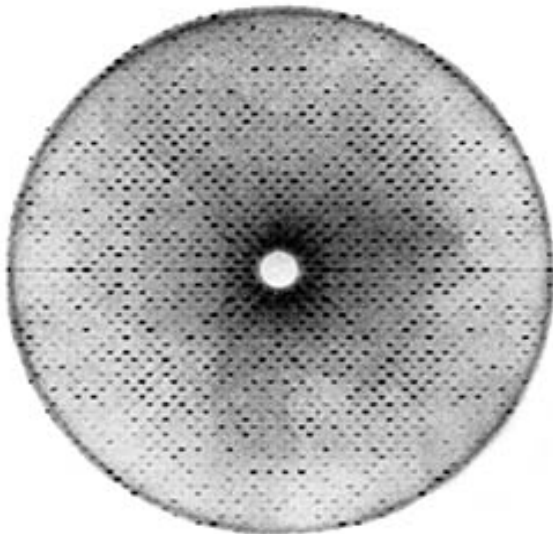
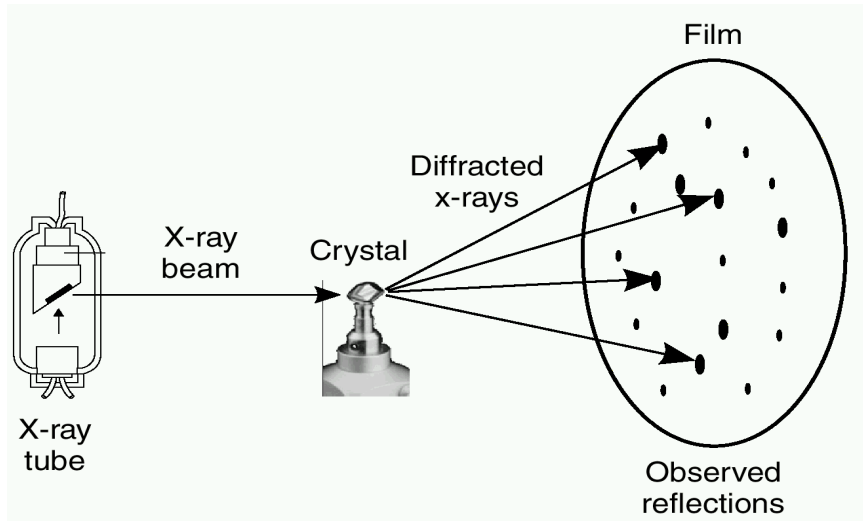


Science, Jul 1989



Experimental structures

X-ray crystallography



1 Å = 10 nm



Very high resolution (~ 1 - 3 Å)
Most heavy atoms visible



One static view, no dynamics
Most flexible parts not visible
Artefacts from crystallisation
(incl. false interfaces !)

Experimental structures

Nuclear magnetic resonance



Multiple conformations
=> info on dynamics

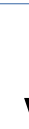


Limited to small systems:
< 50 kDa (~150 nucleotides)

weak oscillating magnetic field

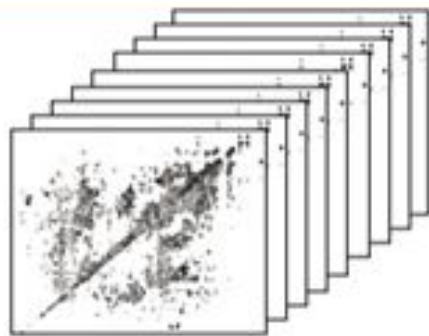


Nucleus in strong
static magnetic field

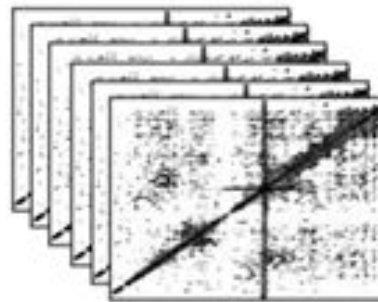


Electromagnetic signal

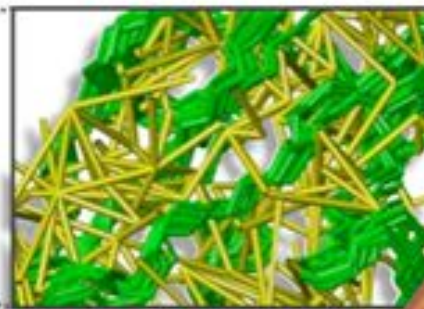
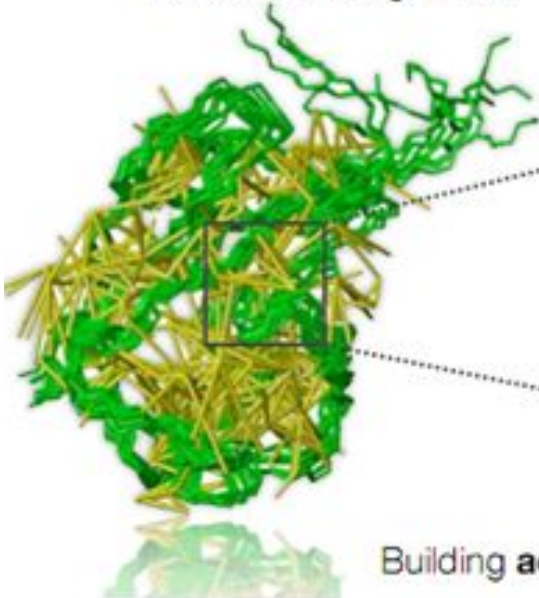
Frequency = $f(\text{atomic environment})$



Resonance assignment



Conformational restraints



Building **accurate** 3D protein model

Protein – RNA interfaces : Experimental data

Main features

The docking problem: Sampling
Evaluation of docking models
Scoring

Flexibility: Flexible docking
Fragment-based docking

Data-driven docking: Contact/interface -driven
Fitting in 3D shape

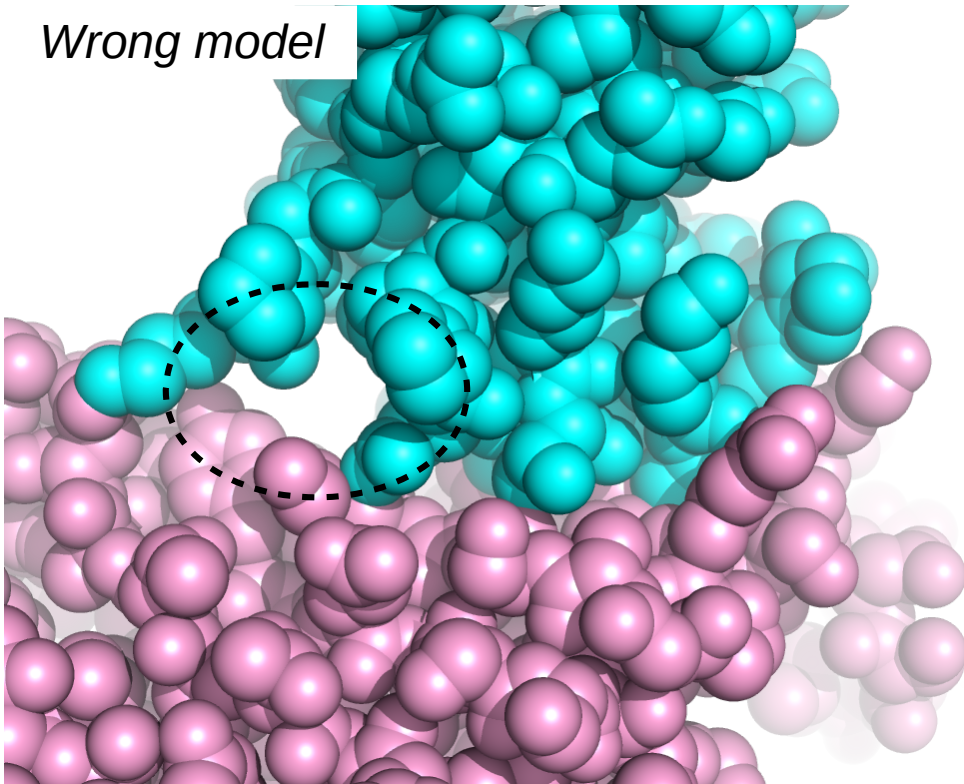
Other docking paradigms

Modeling pipeline

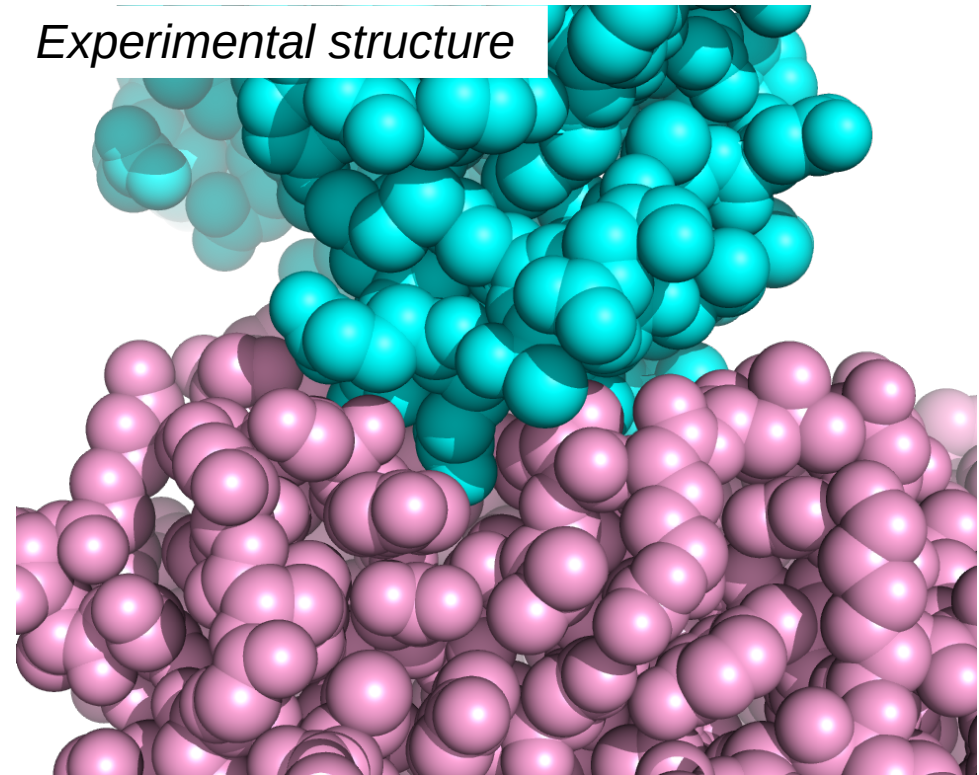
Protein – RNA interfaces

Main characteristics of macro-molecular interfaces

Wrong model



Experimental structure



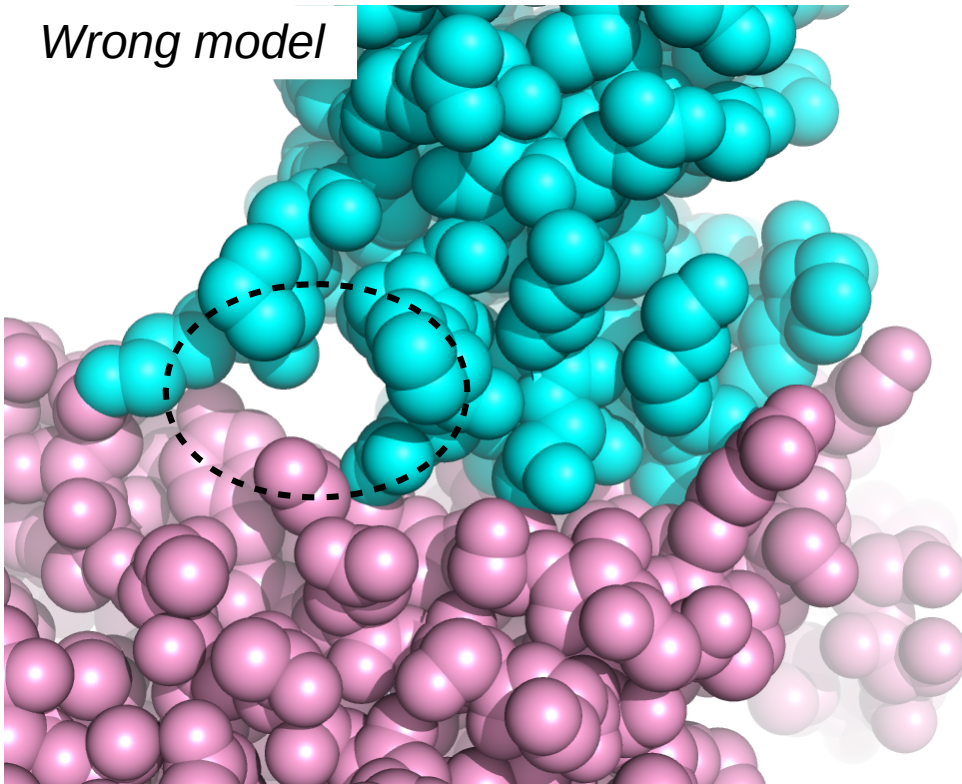
Protein – RNA interfaces

Main characteristics of macro-molecular interfaces

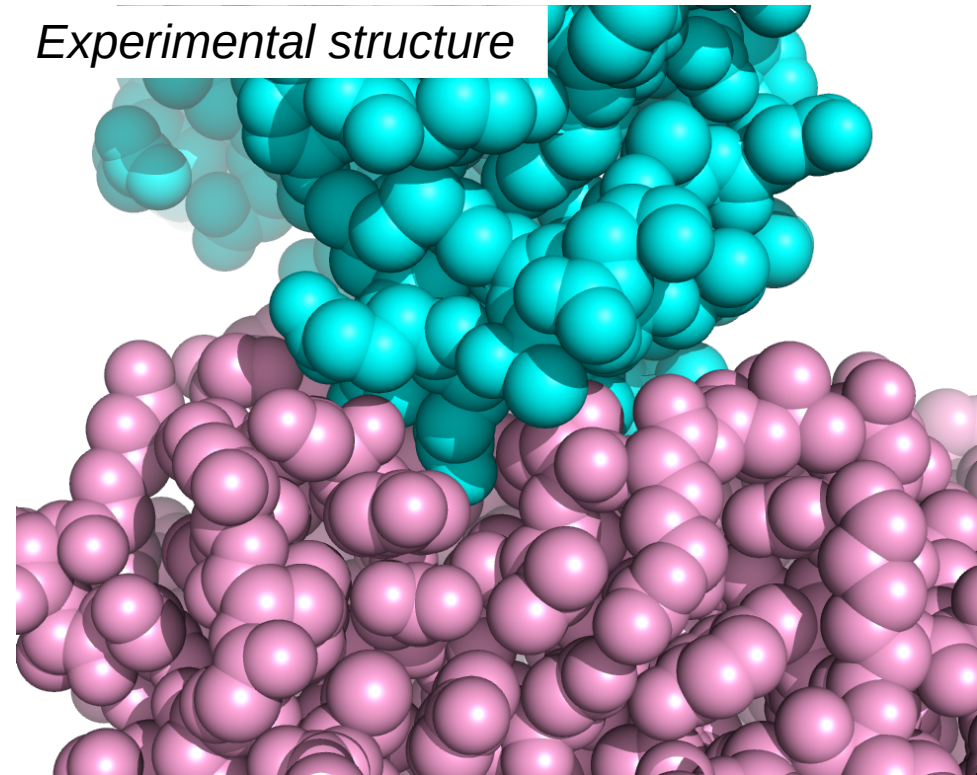
- Shape complementarity

	Large	}	Desolvation effect
	Compact		

Wrong model



Experimental structure



Protein – RNA interfaces

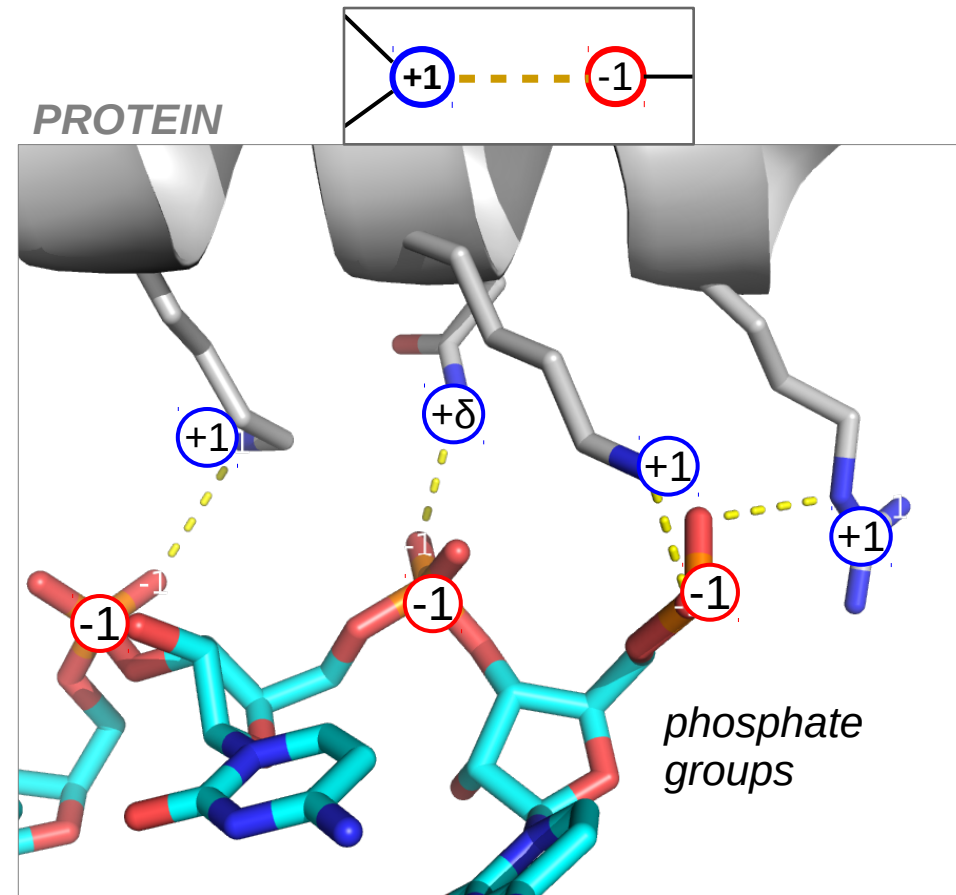
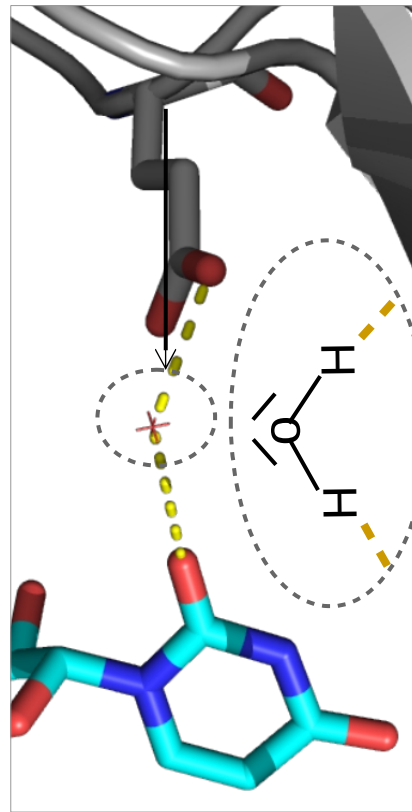
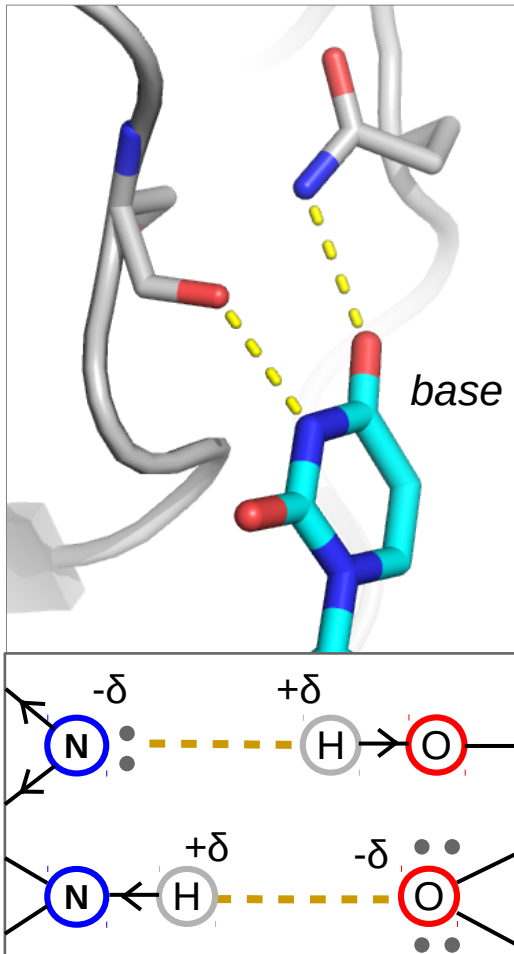
Main characteristics of macro-molecular interfaces

- Shape complementarity

} Desolvation effect	Large
	Compact

 Favourable contacts
- Electrostatic complementarity

} Ionic bridges	}	}	Ionic bridges
			hydrogen bonds
			Water bridges



Protein – RNA interfaces

Main characteristics of macro-molecular interfaces

- Shape complementarity

	Large	}	Desolvation effect
	Compact		Favourable contacts
- Electrostatic complementarity

	Ionic bridges
	hydrogen bonds
	Water bridges

Examples of protein – RNA interfaces

- sequence (non-)specificity *pymol : toxin - antitoxin*
- electrostatic-driven (*Phosphate groups*)
- stacking *pymol : RRM – poly U*

Protein – RNA interfaces

Dec. 2018:

137,000 protein structures

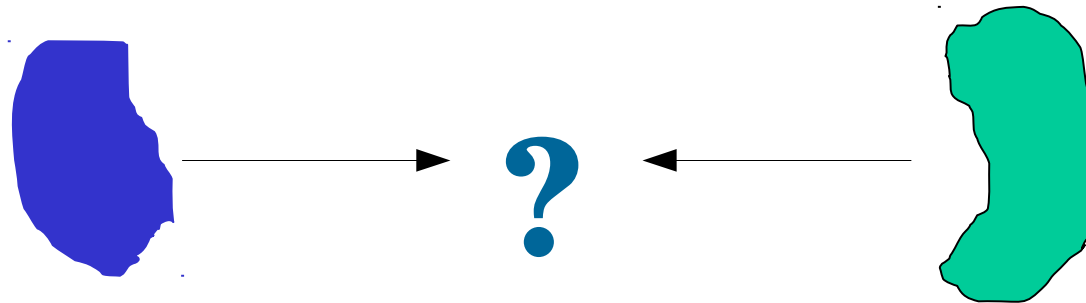
1,400 RNA structures

2,400 protein-RNA structures

*Much lower than what has been
experimentally proved to exist in vivo*

The structure of a complex is more difficult to solve experimentally than a single protein or RNA

=> We often need to model the complex from the structures of the single molecules.



obligatory vs non-obligatory complex

Experimental data on protein – RNA interfaces

The docking problem: Sampling

Evaluation of docking models

Scoring

Flexibility: Flexible docking

Fragment-based docking

Data-driven docking: Contact/interface -driven

Fitting in 3D shape

Other docking paradigms

Modeling pipeline

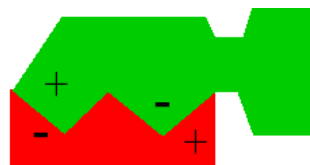
The Docking Problem

Predicting *in silico* the most probable 3D structures of a complex at the thermodynamic equilibrium, using the 3D structures of its individual components.

! It is assumed that the 2 molecules do bind !

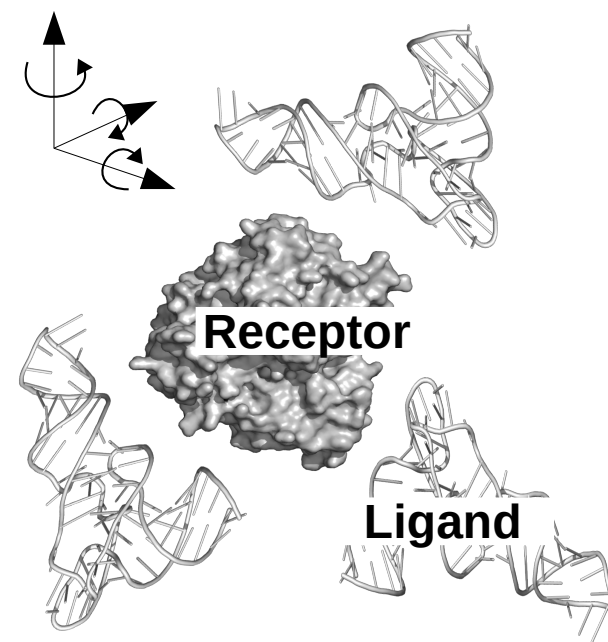
The thermodynamically **most stable conformation** is called **native conformation**. It is assumed to be the one observed experimentally (and carrying the biological function).

The docking is mainly based on geometrical and chemical properties.



The rigid approximation:

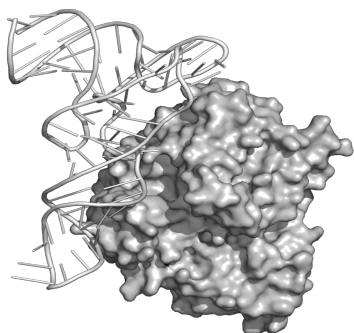
- The 2 molecules are kept (mostly) **rigid**
- One molecule being fixed, the other has **6 degrees of freedom**
- We neglect intra-molecular energies
- We minimize the **intermolecular interaction energy**



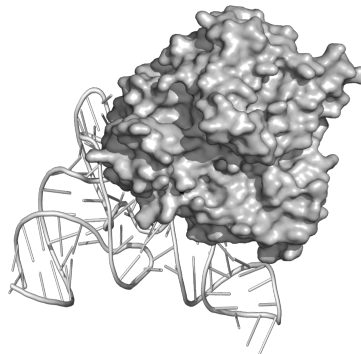
The Docking Problem

Sampling →

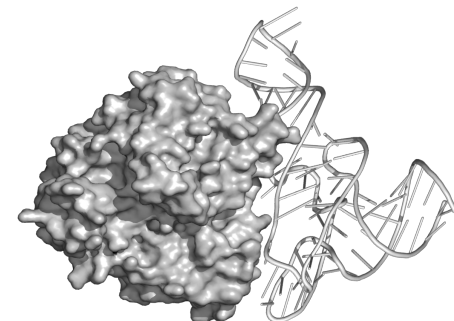
Model 1



Model 2



Model 3



Scoring →

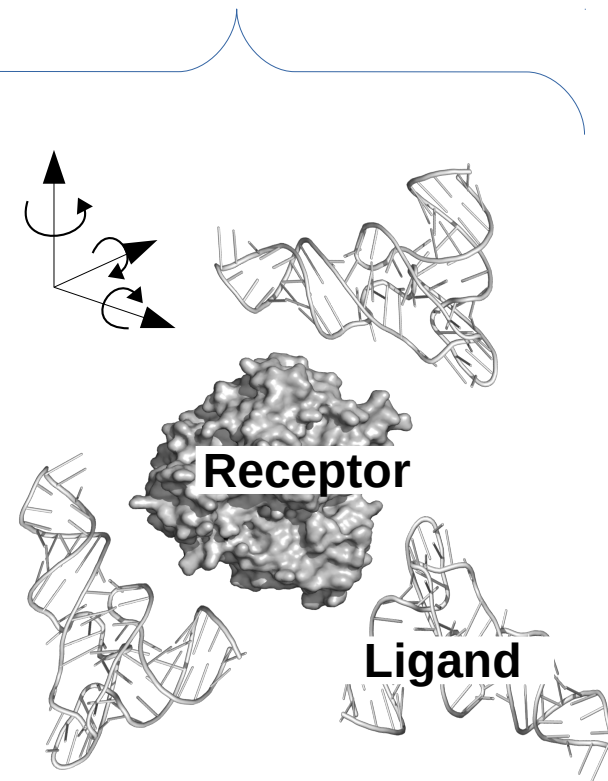
$E = -40$ kcal/mol

$E = -35$ kcal/mol

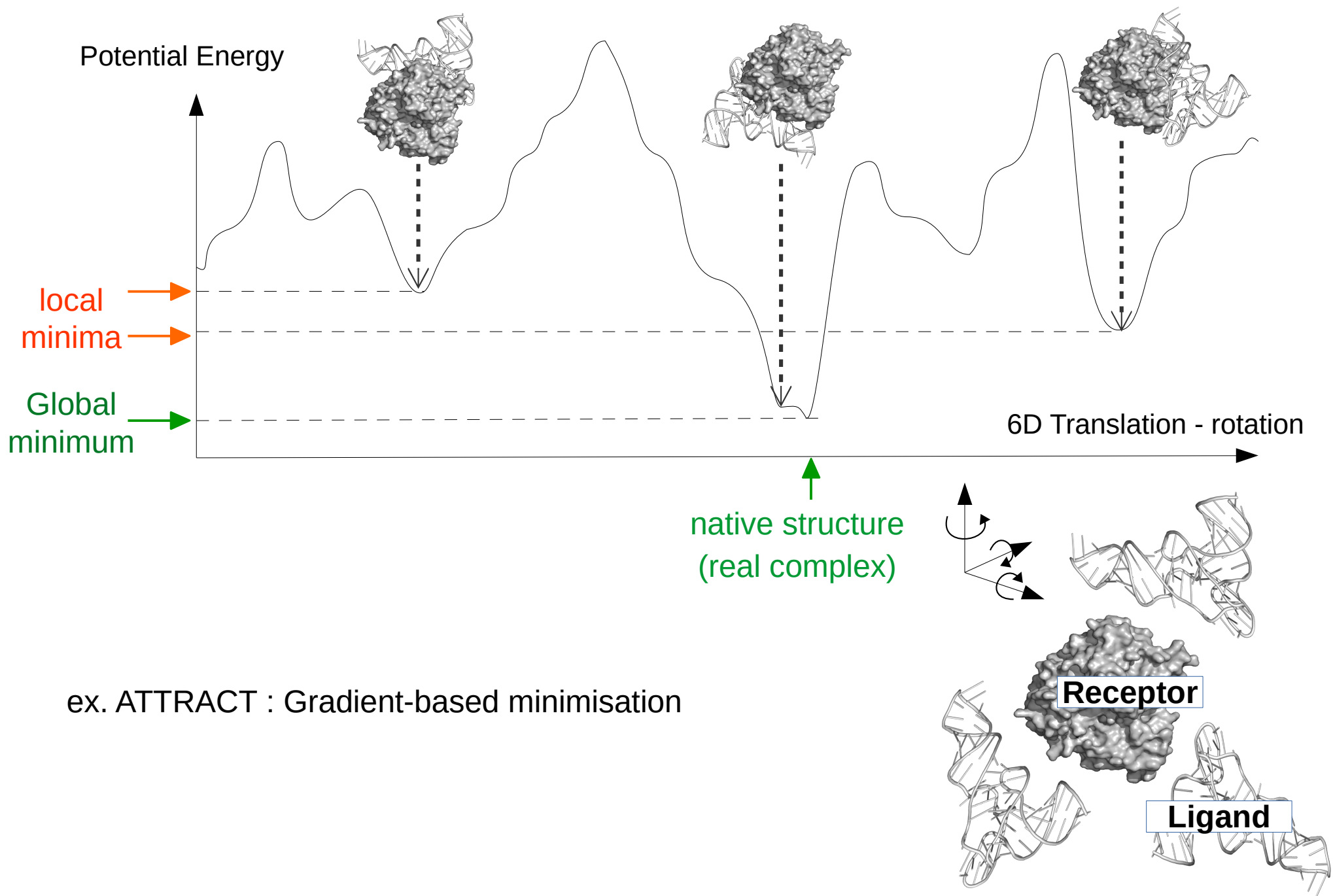
$E = -3$ kcal/mol

The rigid approximation:

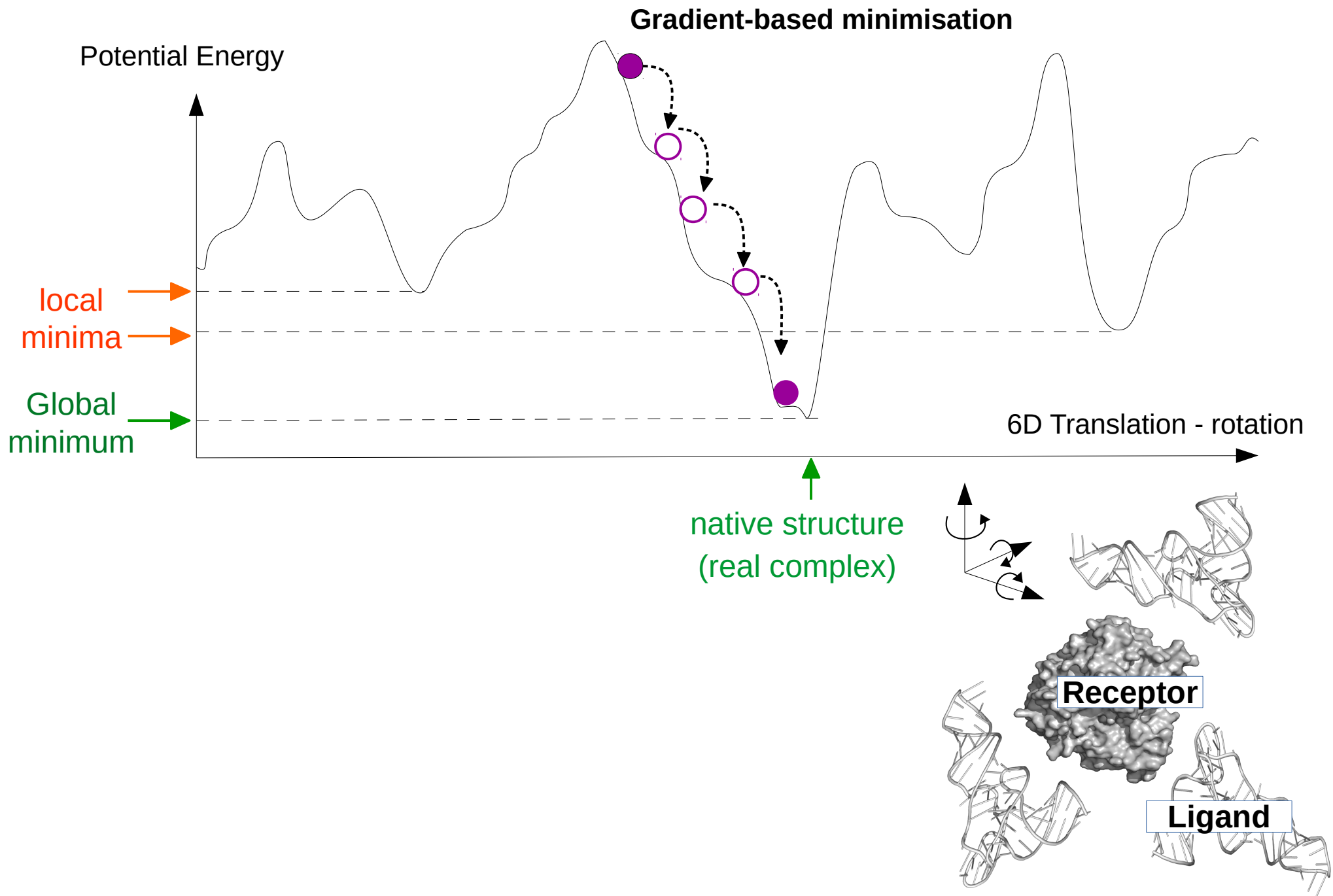
- The 2 molecules are kept (mostly) **rigid**
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Sampling of docking models



Sampling of docking models



Sampling of docking models

Place the ligand at random position

Compute the inter-molecular energy

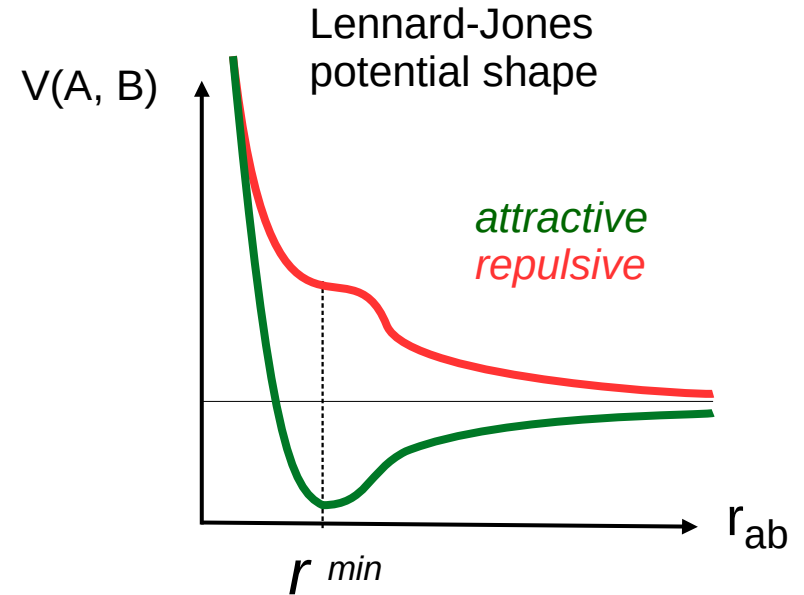
Assumption: pairwise additive interactions

$$V_{(lig, rec)} = \sum_{(A \in Rec ; B \in Lig)} V_{(A, B)}$$

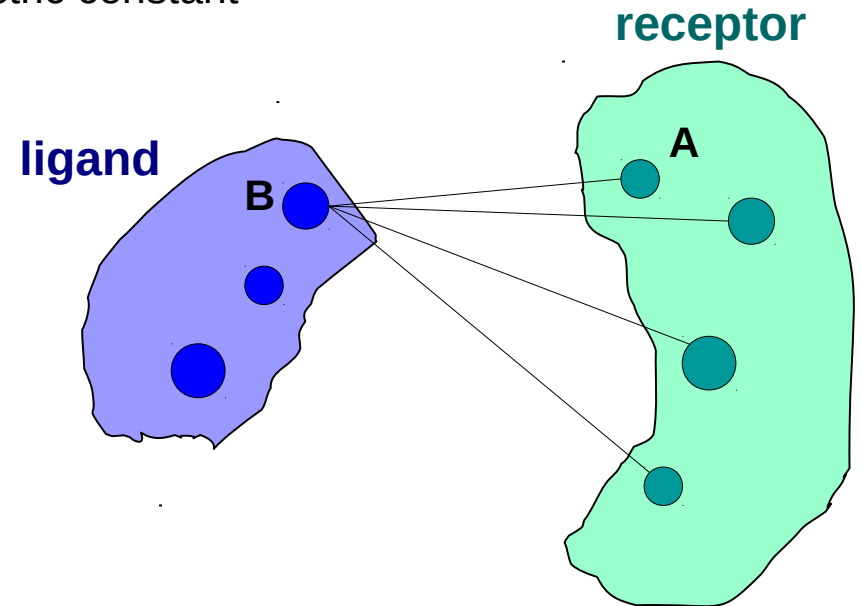
$$V_{(A, B)} = \left(\frac{R_{AB}^{12}}{r_{ab}^{12}} - \frac{R_{AB}^6}{r_{ab}^6} \right) + \frac{q_a q_b}{\epsilon r_{ab}}$$

Van der Waals *coulomb*

VdW : short range
Electrostatics: long range
(esp. RNA negative charges)



q = electrostatic charge
 R = mean atomic radius
 r = inter-atomic distance
 ϵ = dielectric constant



Sampling of docking models

Place the ligand at random position

Compute the inter-molecular energy
Assumption: pairwise additive interactions

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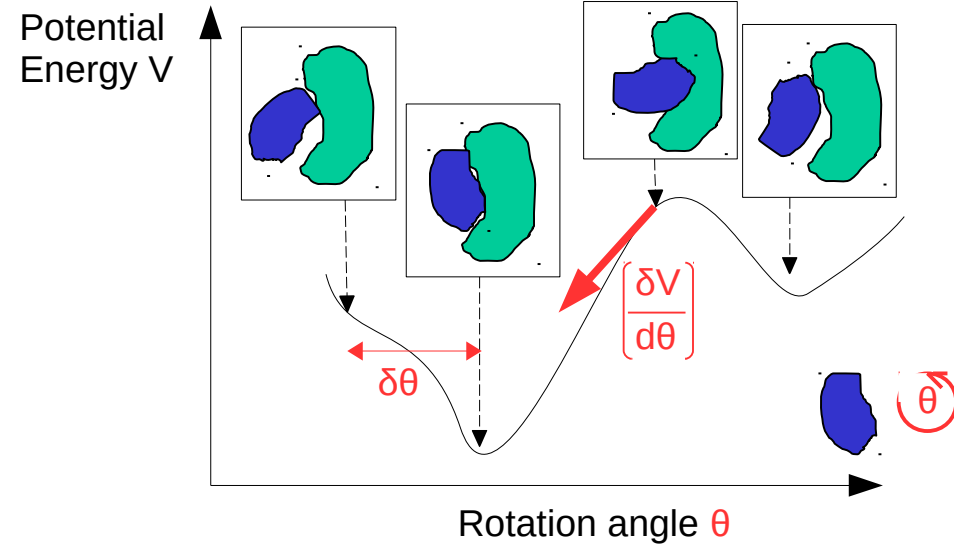
$$V_{(A, B)} = \left(\frac{R_{AB}}{r_{ab}} \right)^{12} - \left(\frac{R_{AB}}{r_{ab}} \right)^6 + \frac{q_a q_b}{\epsilon r_{ab}}$$

Van der Waals *coulomb*

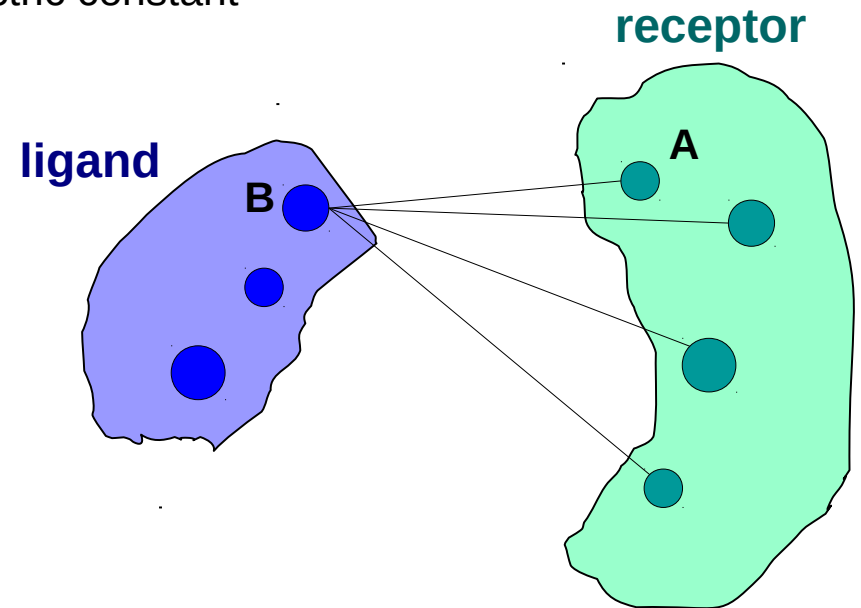
Compute the derivative (gradient)

$$\frac{\delta V}{dx \, dy \, dz \, d\theta \, d\phi \, d\omega}$$

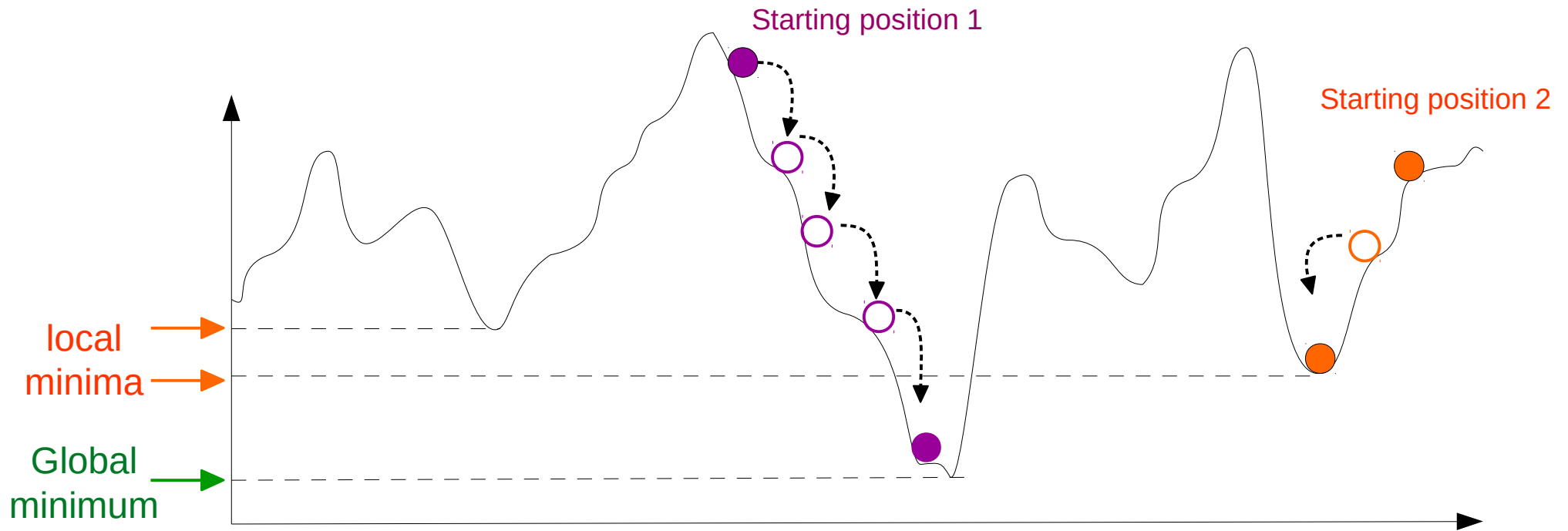
Displace and rotate the ligand in that direction from a small distance $\delta\theta$



q = electrostatic charge
 R = mean atomic radius
 r = inter-atomic distance
 ϵ = dielectric constant

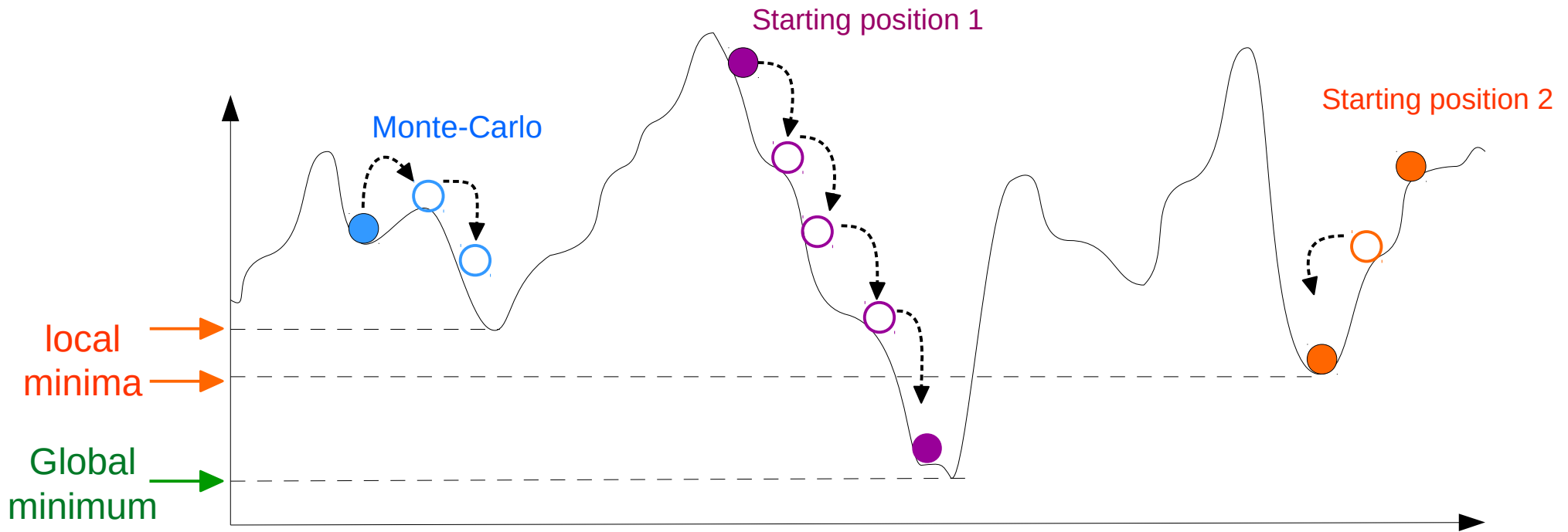


Sampling of docking models



How can we increase the probability to end-up in the global minimum?

Sampling of docking models



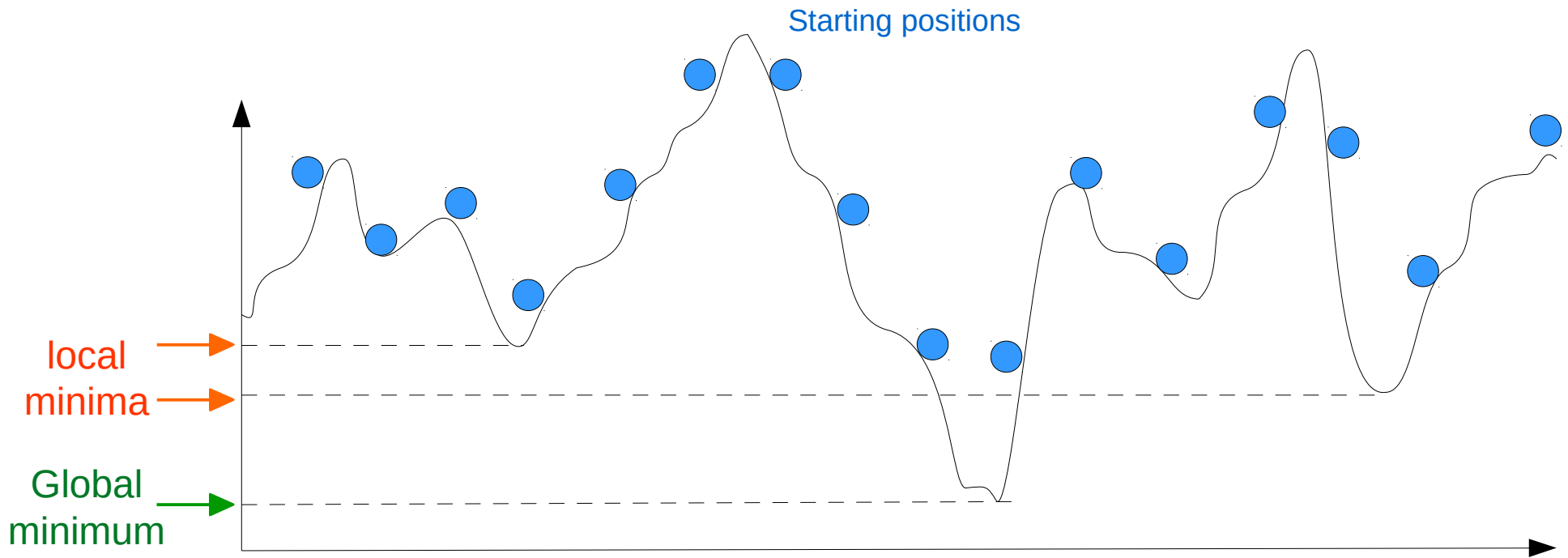
How can we increase the probability to end-up in the global minimum?

- Allow some increases of energy : **Monte-Carlo** procedure
☹️ Computationally demanding, Stochastic results

pos2 = pos1 + random($\delta\theta$)
Accept move with proba P

$$P = e^{-\frac{E_2 - E_1}{k_b T}}$$

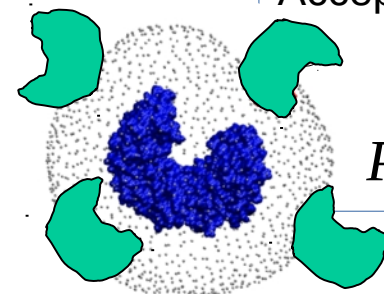
Sampling of docking models



How can we increase the probability to end-up in the global minimum?

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- Start from many **multiple starting positions**

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Sampling of docking models

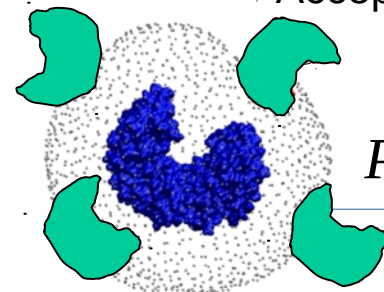


How can we increase the probability to end-up in the global minimum?

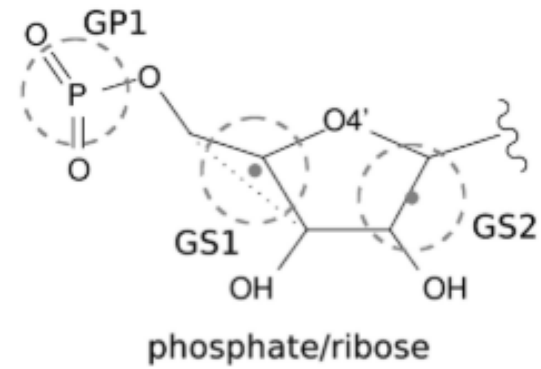
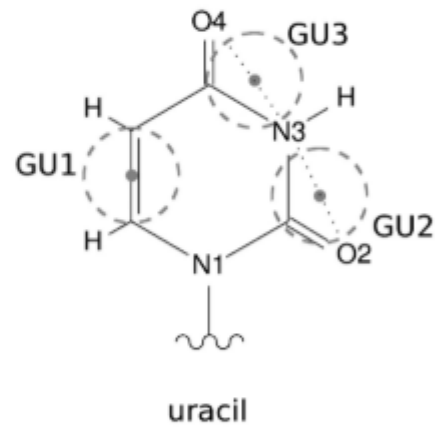
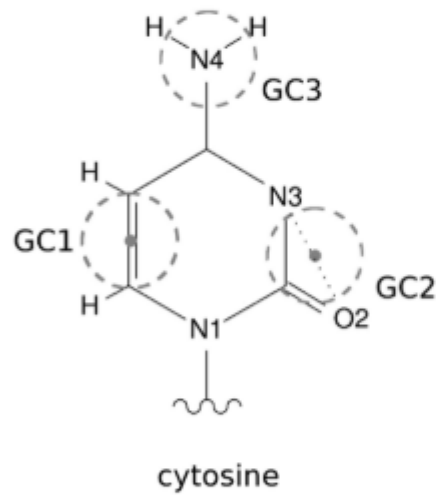
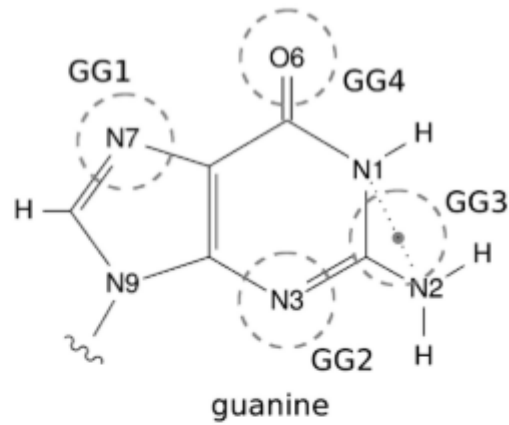
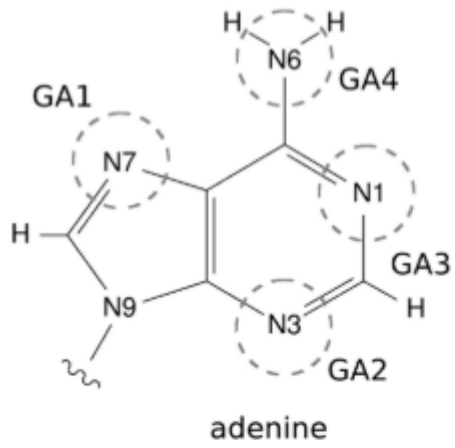
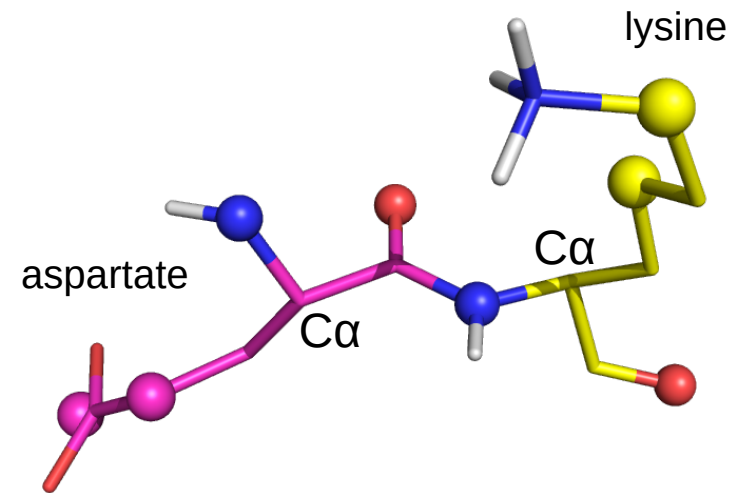
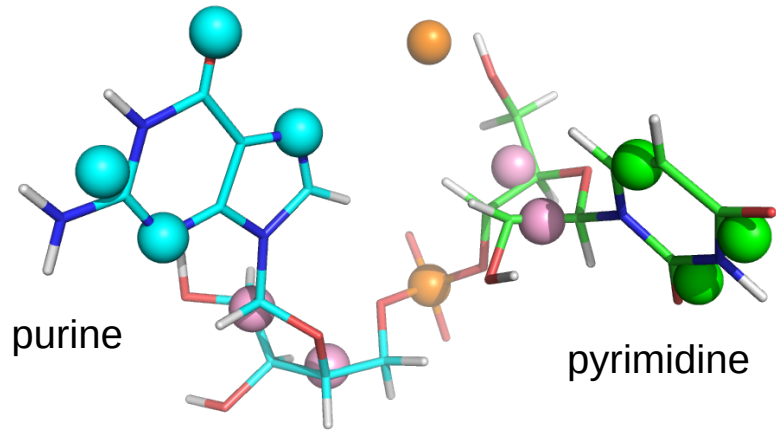
- Allow some increases of energy : **Monte-Carlo** procedure
☹️ Computationally demanding, Stochastic results
- Start from many **multiple starting positions**
- Reduce the number of **local minima**

pos2 = pos1 + random($\delta\theta$)
Accept move with proba P

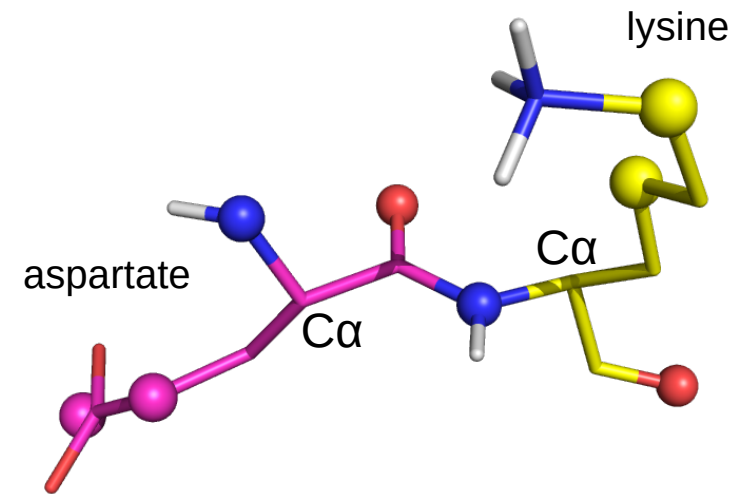
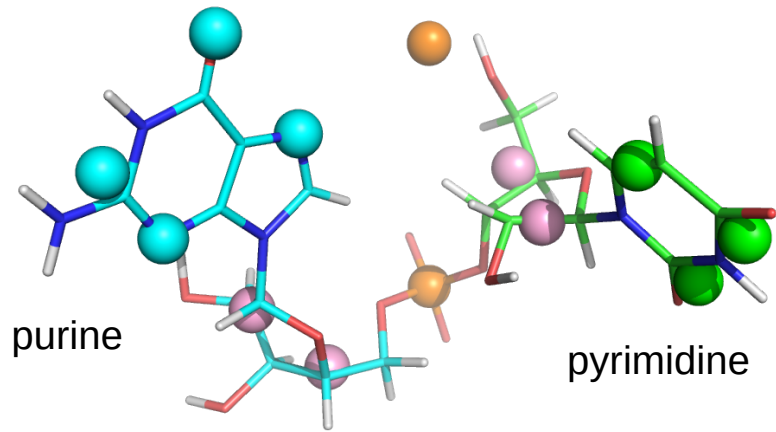
$$P = e^{-\frac{E_2 - E_1}{k_b T}}$$



Sampling of docking models

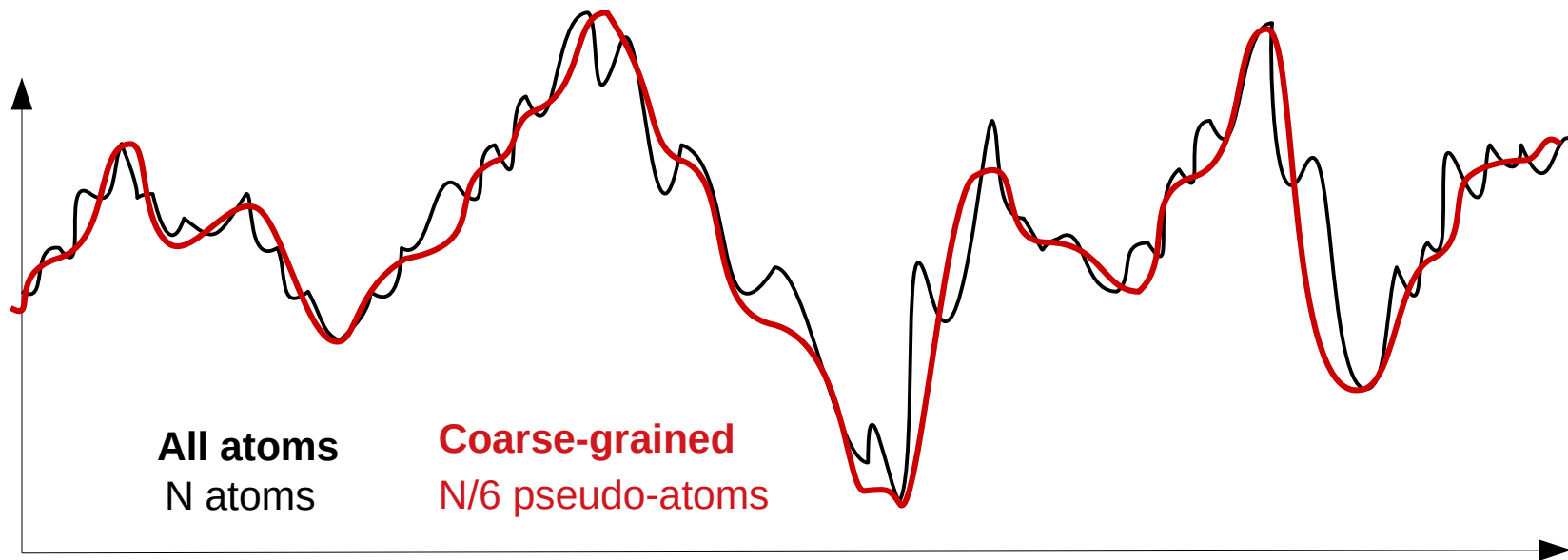


Sampling of docking models

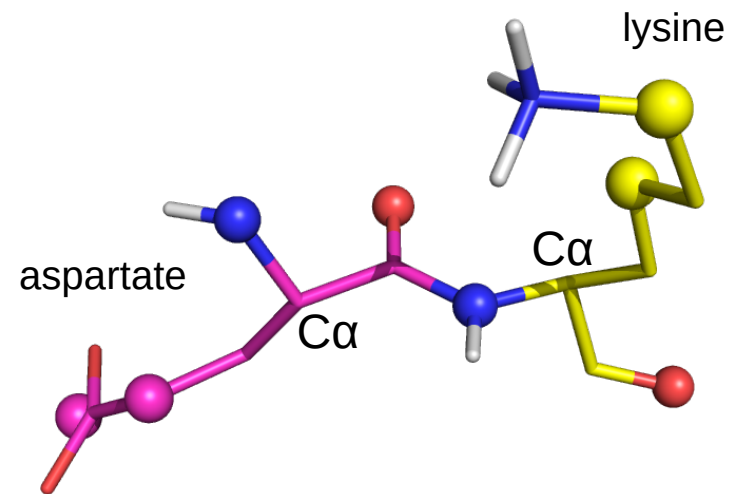
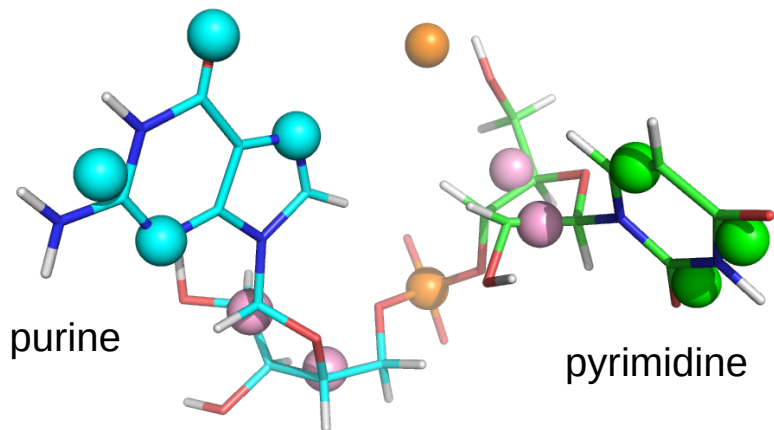


Coarse-grained model : replace few atoms by one pseudo-atom

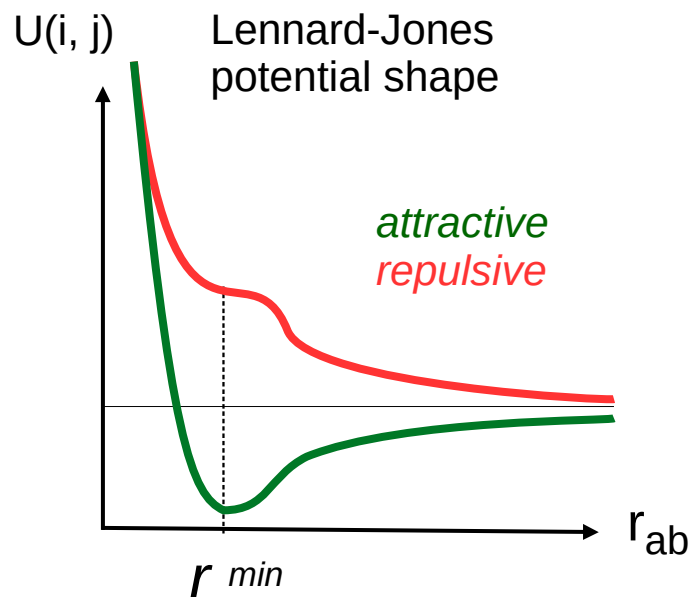
- Faster pairwise computations: $O(N^2) \Rightarrow 36x$ faster
- Smoother energy landscape \Rightarrow less local minima
- Account for some inaccuracies in the structures



Sampling of docking models



Parametrization



2 parameters x 31 amino-acid beads x 17 RNA beads

attraction

$$U_{ij}^{attr}(r) = \epsilon_{ij} \left(\frac{\sigma_{ij}}{r^8} - \frac{\sigma_{ij}}{r^6} \right)$$

repulsion

$$U_{ij}^{rep}(r) = \begin{cases} U_{ij}^{attr}(r) + 2U_{ij}^m & \text{for } r \leq r_{ij}^m \\ -U_{ij}^{attr}(r) & \text{for } r > r_{ij}^m \end{cases}$$

Adjust parameters based on known structure, so as to give better score to models close to known structures

Experimental data on protein – RNA interfaces

The docking problem: Sampling

Evaluation of docking models

Scoring

Flexibility: Flexible docking

Fragment-based docking

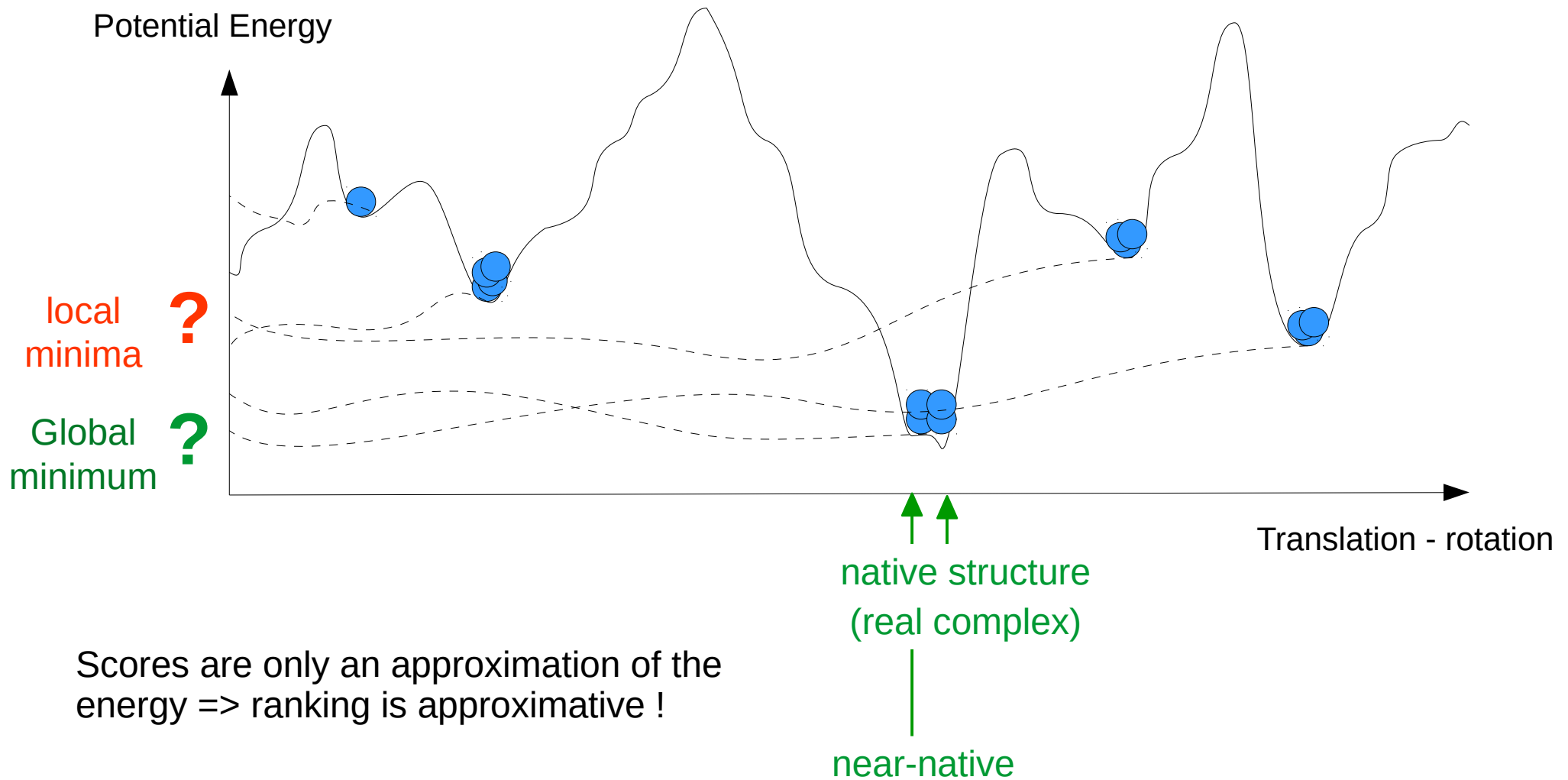
Data-driven docking: Contact/interface -driven

Fitting in 3D shape

Other docking paradigms

Modeling pipeline

Evaluation of docking models



Scores are only an approximation of the energy => ranking is approximative !

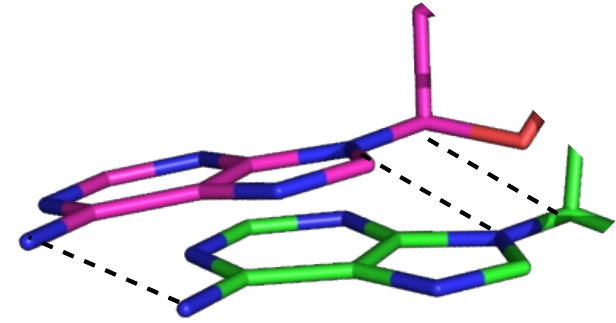
What shall we aim for, in sampling?



We put ourselves in the **artificial case** where the real structure of the complex is known

To assess the quality of a solution, we compute the **Root mean squared deviation (RMSD)**:
Average deviation of the atoms in the **docking solution** vs in the **experimental structure**.

$$RMSD = \sqrt{\frac{\sum_{i=1}^N \left((x_i^a - x_i^b)^2 + (y_i^a - y_i^b)^2 + (z_i^a - z_i^b)^2 \right)}{N}}$$



Evaluation of docking models



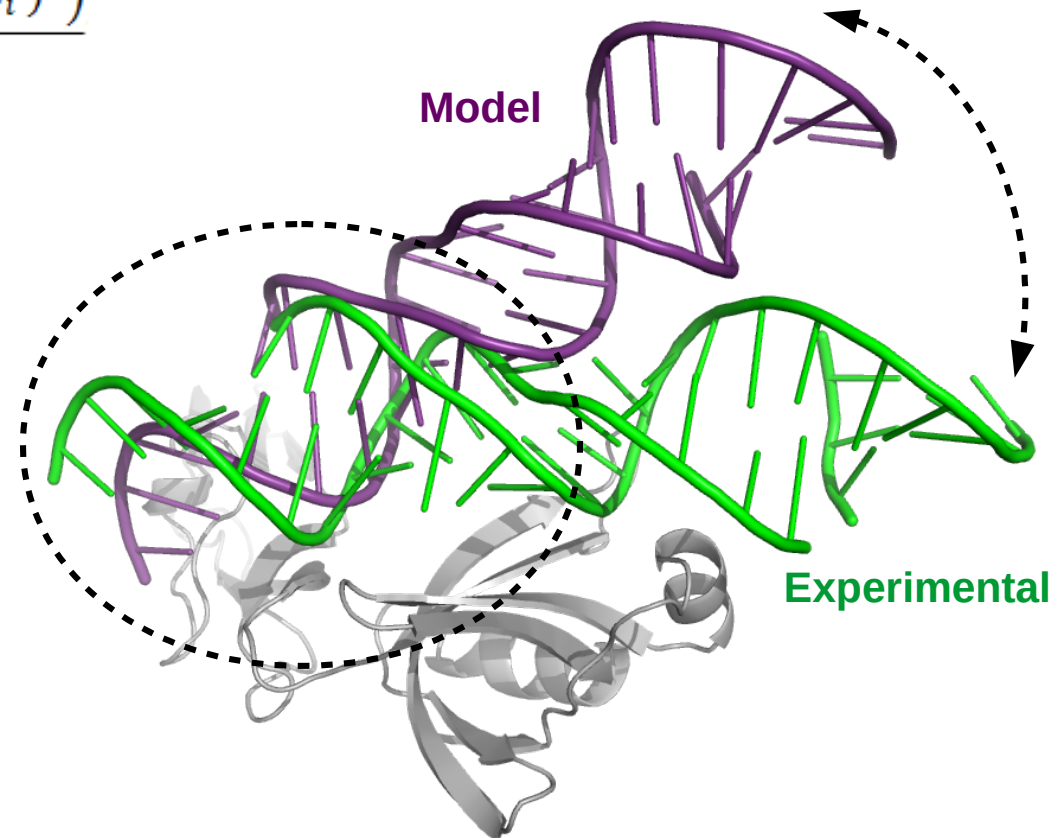
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Ligand-RMSD (L-RMSD)

RMSD of docked vs bound ligand, after fitting the model on the bound receptor





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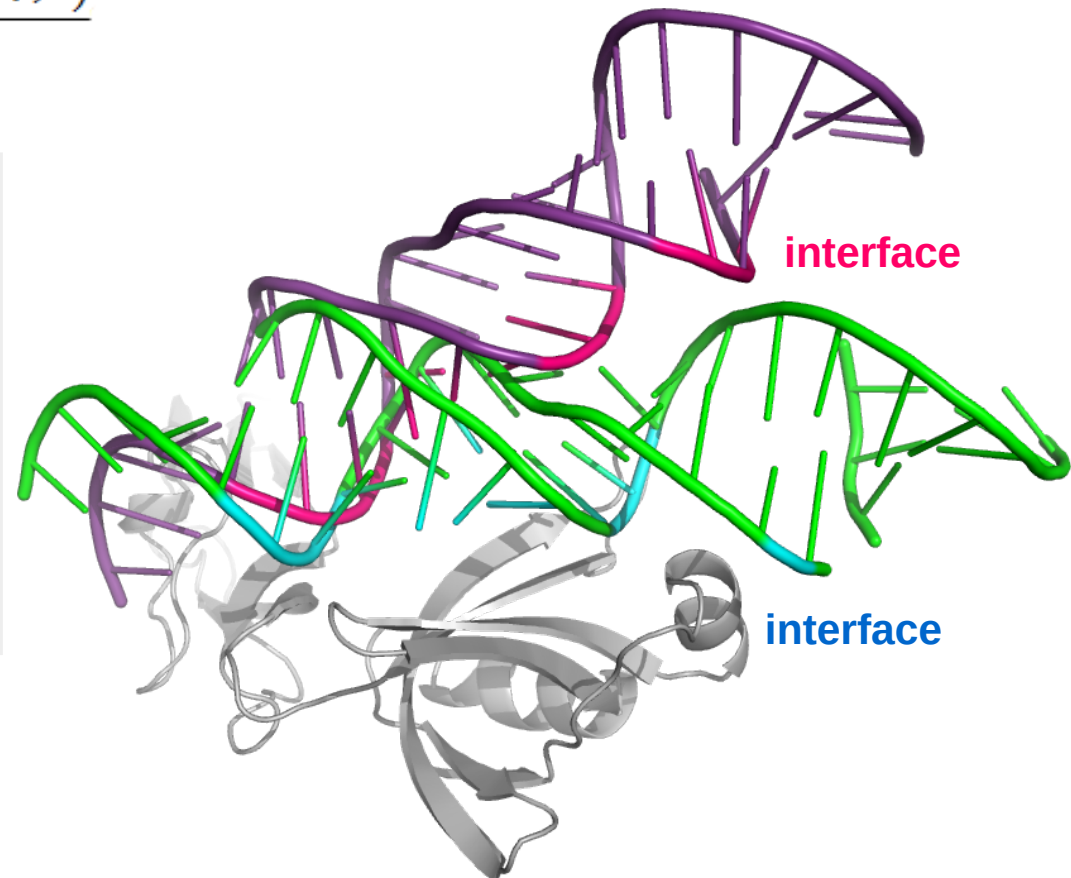
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Ligand-RMSD (L-RMSD)

RMSD of docked vs bound ligand, after fitting the model on the bound receptor

interface-RMSD (i-RMSD)

RMSD unbound vs bound of only the interface (receptor and ligand atoms at < 10Å distance from each-other), after fitting on the bound interface the equivalent atoms of the model.





We put ourselves in the **artificial case** where the real structure of the complex is known

To assess the quality of a solution, we compute the **Root mean squared deviation (RMSD)**: Average deviation of the atoms in the **docking solution** vs in the **experimental structure**.

$$RMSD = \sqrt{\frac{\sum_{i=1}^N \left((x_i^a - x_i^b)^2 + (y_i^a - y_i^b)^2 + (z_i^a - z_i^b)^2 \right)}{N}}$$

Ligand-RMSD (L-RMSD)

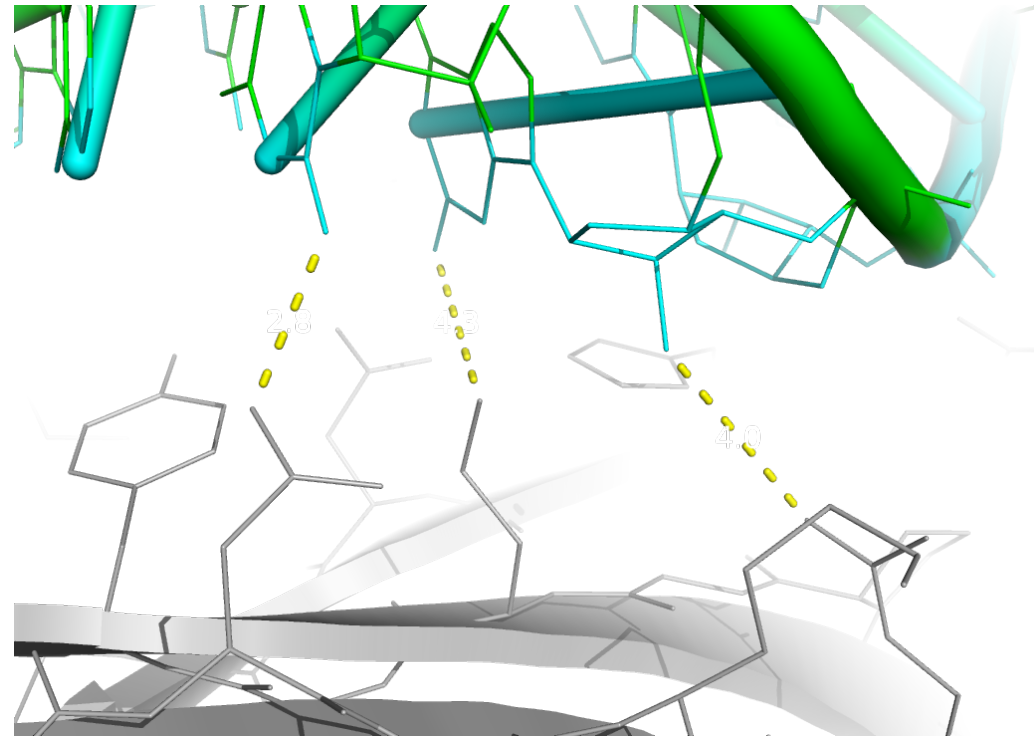
RMSD of docked vs bound ligand, after fitting the model on the bound receptor

interface-RMSD (i-RMSD)

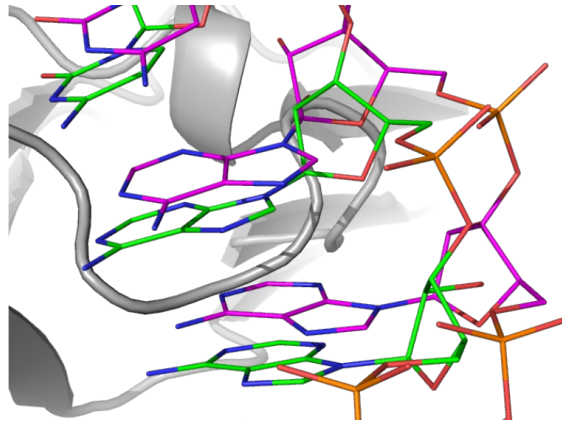
RMSD unbound vs bound of only the interface (receptor and ligand atoms at < 10Å distance from each-other), after fitting on the bound interface the equivalent atoms of the model.

Fraction of native contacts (Fnat)

A contact is a pair of RNA-protein atoms at <5 Å from each other. Fnat is the % of contacts recovered by the model.



Evaluation of docking models

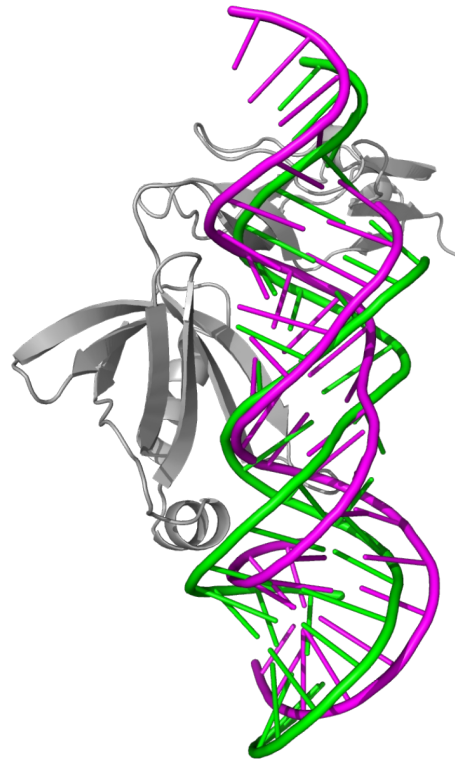


What should we aim for?

iRMSD: 0.5 Å

iRMSD: 1.2 Å
**

iRMSD: 3.5 Å
*



Experimental data on protein – RNA interfaces

The docking problem: Sampling
Evaluation of docking models

Scoring

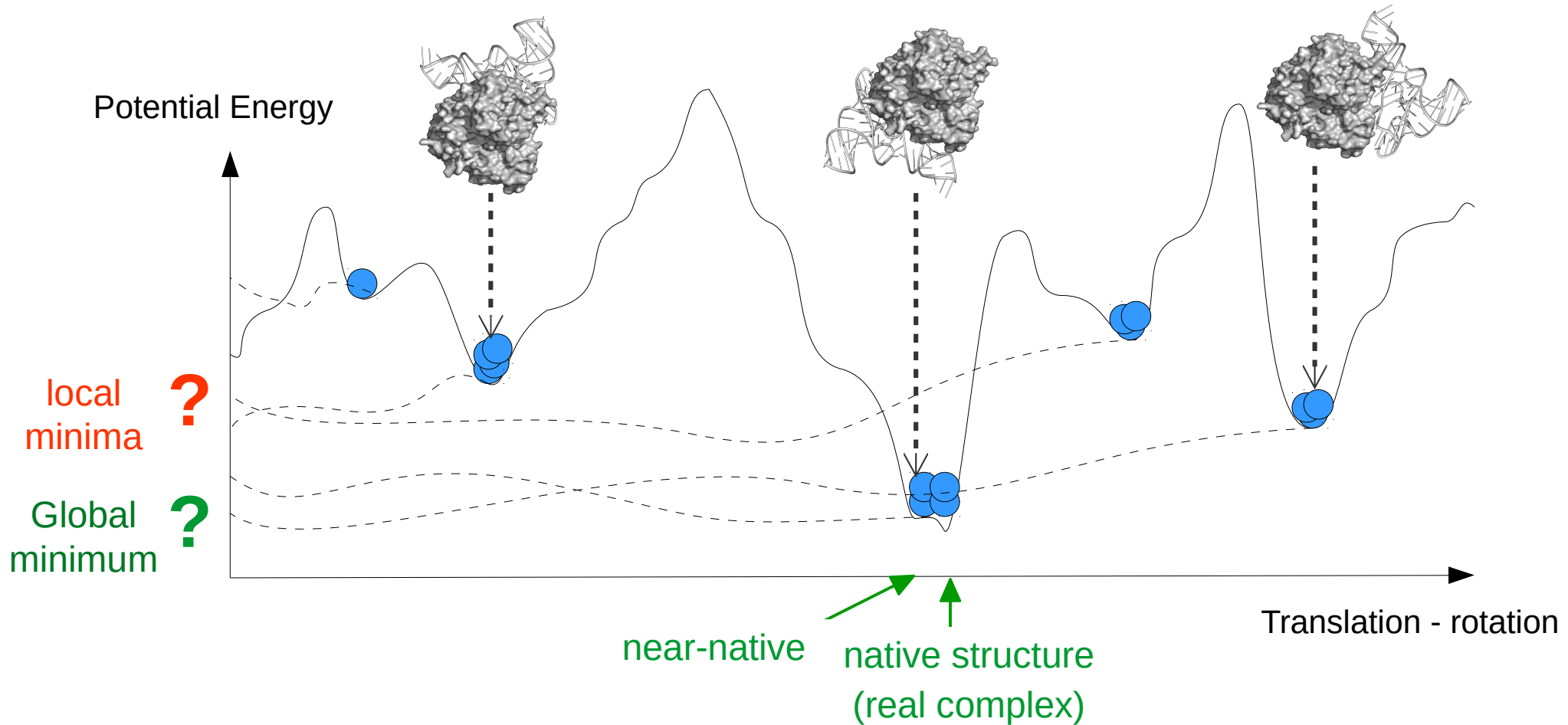
Flexibility: Flexible docking
Fragment-based docking

Data-driven docking: Contact/interface -driven
Fitting in 3D shape

Other docking paradigms

Modeling pipeline

Scoring

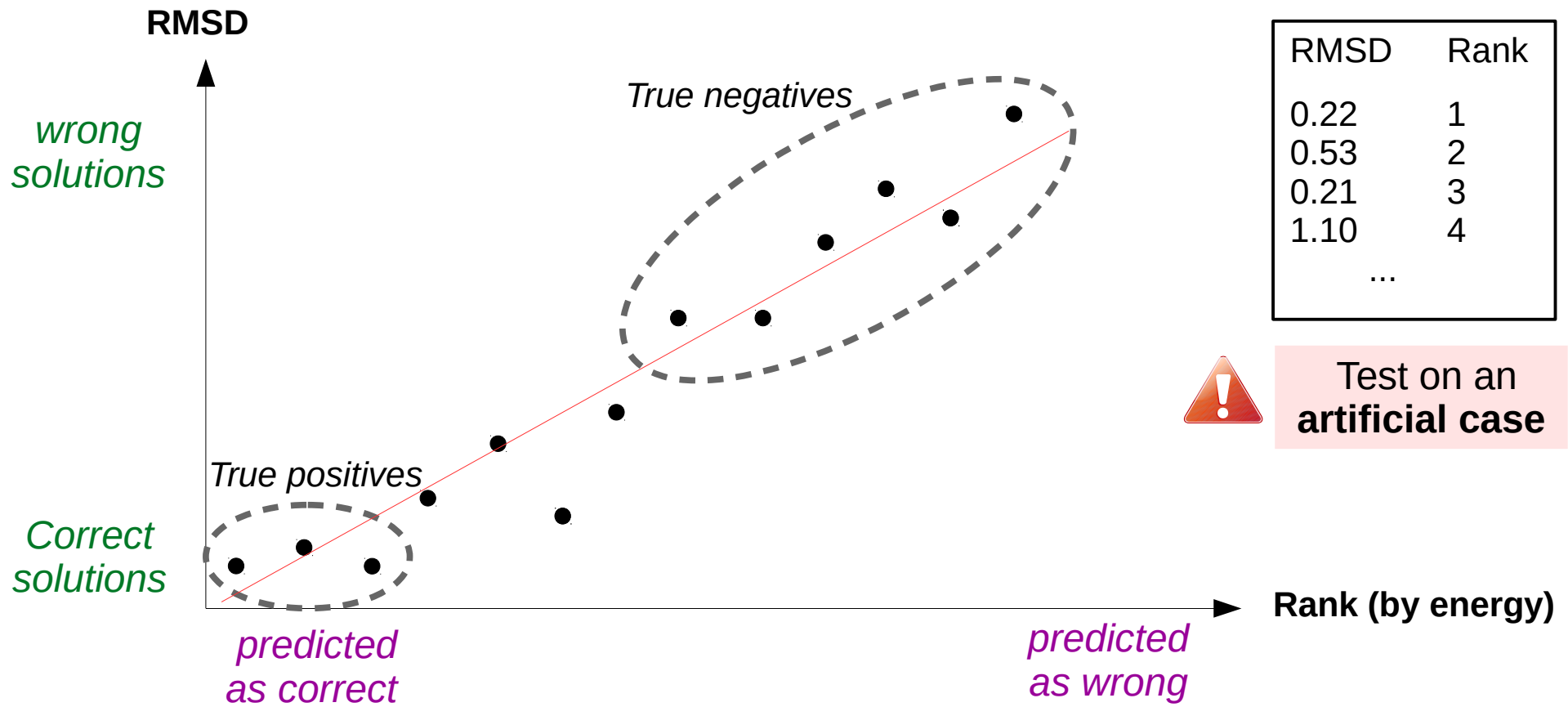


Sampling : Producing models as close as possible from the native structure

Scoring : Discriminating good („near-native“) solutions from wrong solutions („decoys“)

To discriminate the wrong solution from the good ones, all solutions are scored (approximation of the energy) and **ranked** from the smallest to highest score.

Scoring

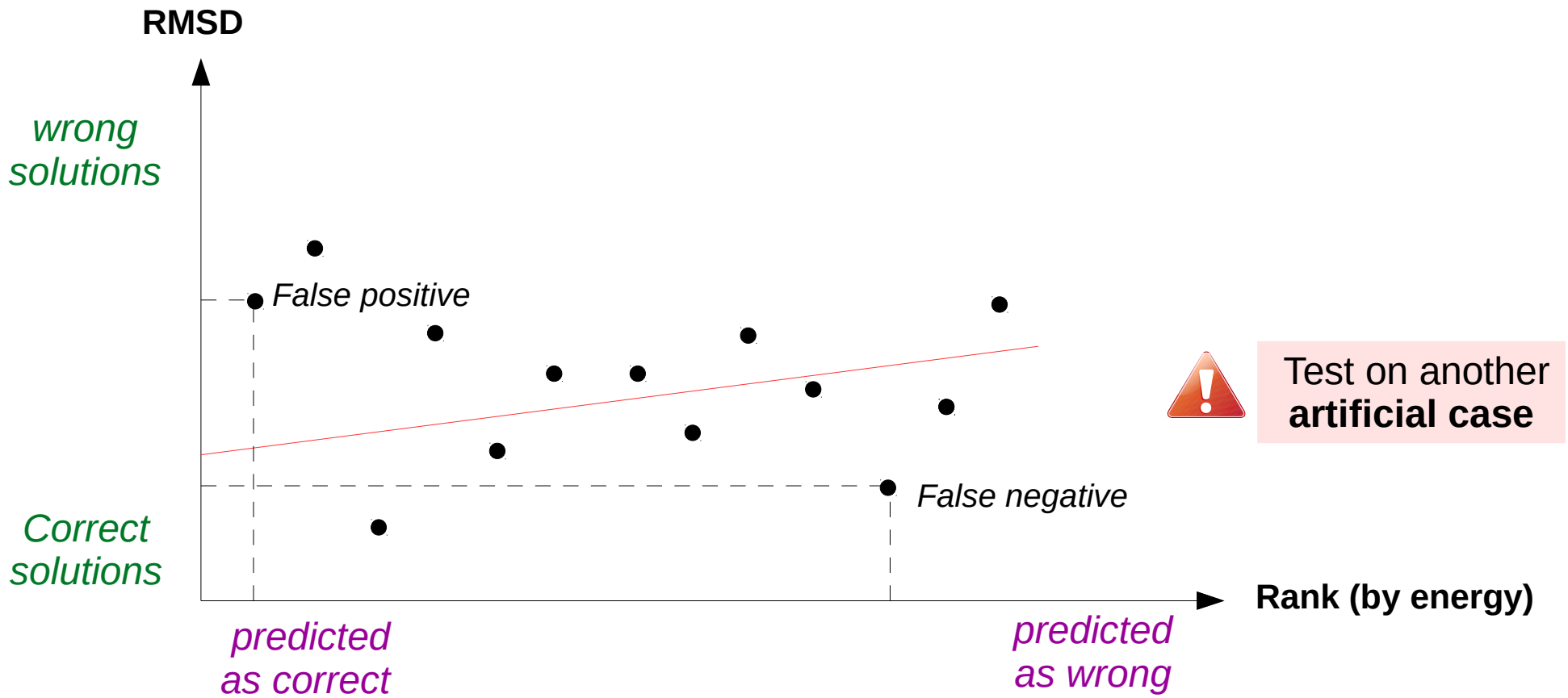


Sampling : Producing models as close as possible from the native structure

Scoring : Discriminating good („near-native“) solutions from wrong solutions („decoys“)

To discriminate the wrong solution from the good ones, all solutions are scored (approximation of the energy) and **ranked** from the smallest to highest score.

Scoring



The first good model is typically in the ~10-100 top-ranked model

=> You have to consider as many models to not throw away all good ones!

How to check if any “predicted good structure” is really good ?



The first good model is typically in the ~10-100 top-ranked model

=> You have to consider as many models to not throw away all good ones!

=> What about a **real case**, where we don't have access to the RMSD ?
(there is no reference structure)

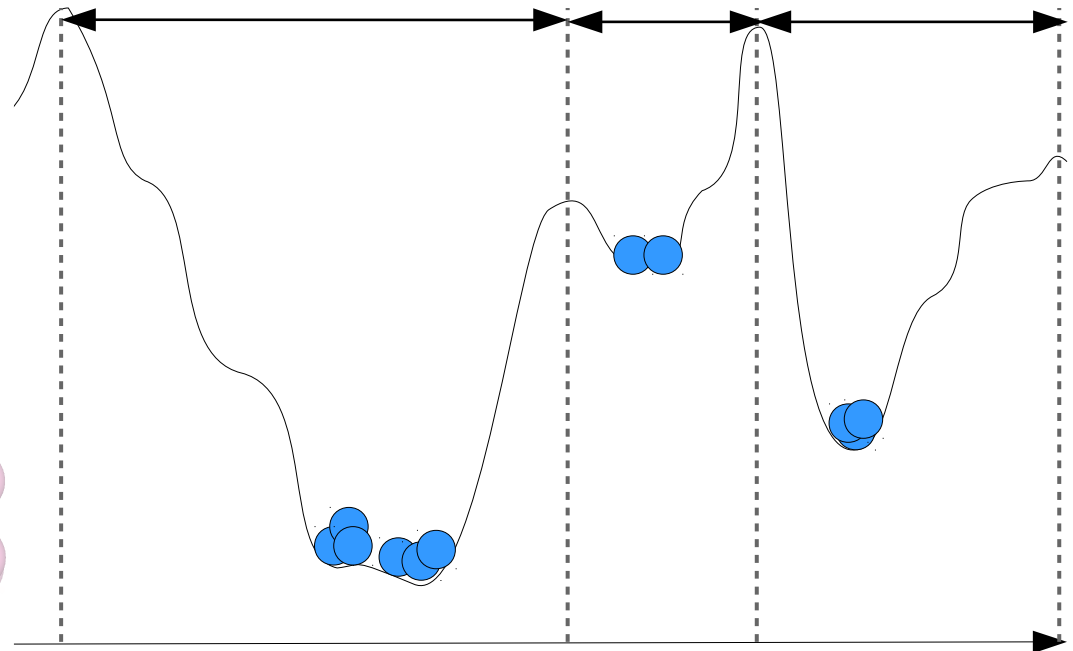
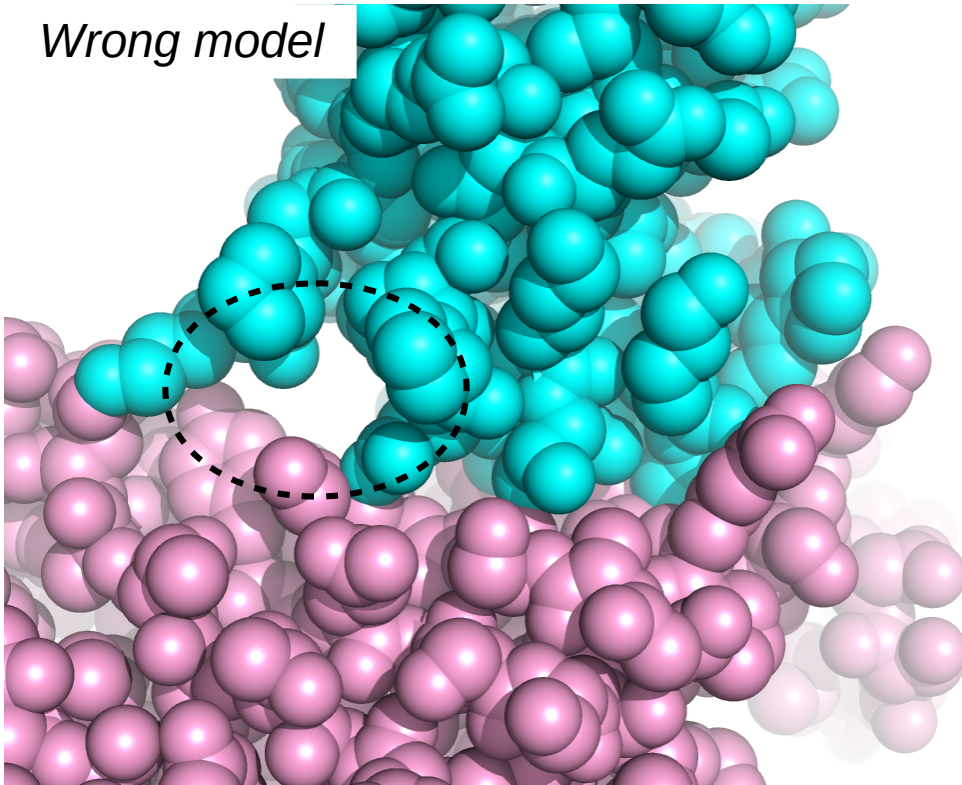
How to check if any “predicted good structure” is really good ?

- > Visualization: *large interface*
good shape complementarity (no holes at interface)
clustering at same area



real case

Wrong model

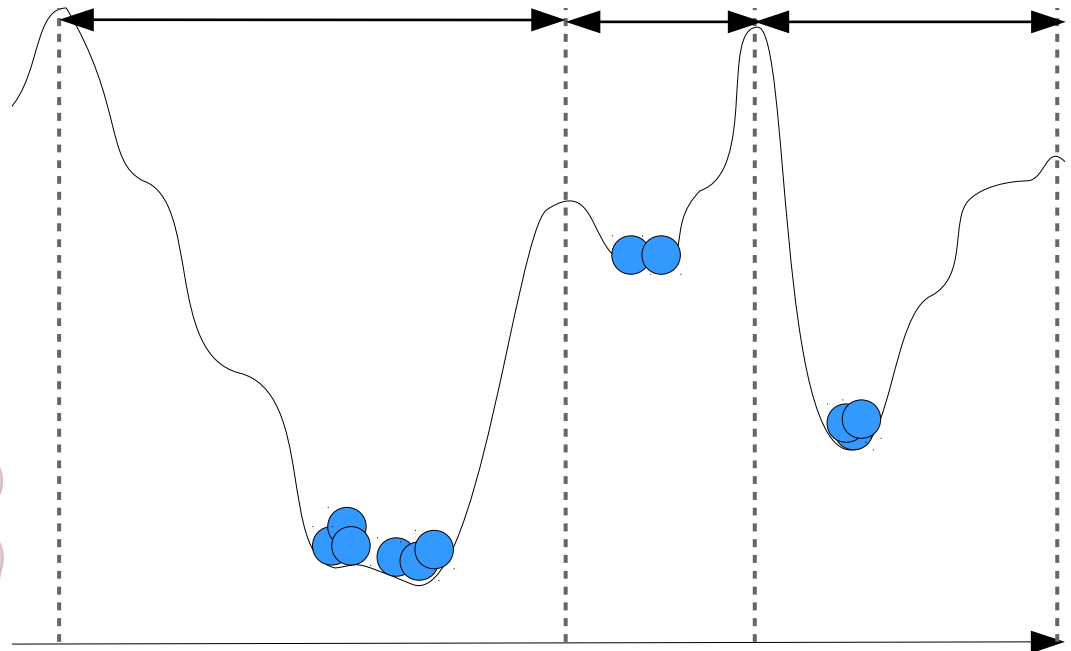
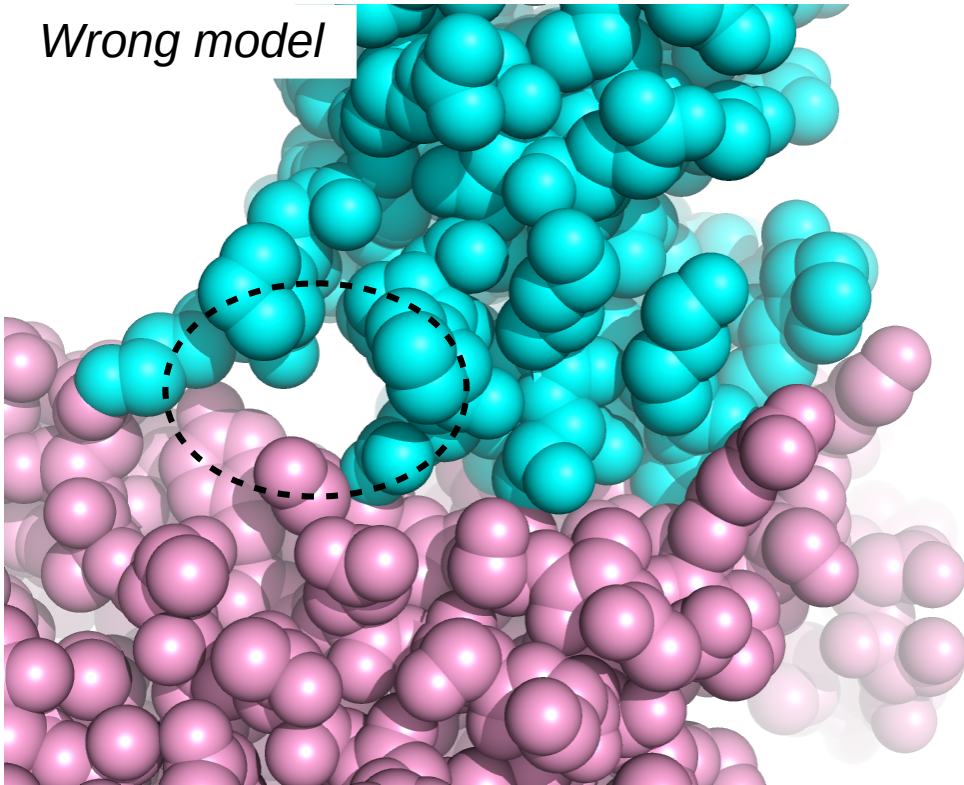


How to check if any “predicted good structure” is really good ?

Antibodies are large proteins synthesized by the immune system to identify and neutralize pathogens such as bacteria and viruses. PDB code 1E6J is a complex between the HIV capsid protein and a large antibody that binds to it.

pymol : HIV capsid - antibody

Wrong model



How to check if any “predicted good structure” is really good ?

- > Visualization: *large interface*
good shape complementarity (no holes at interface)
clustering at same area
- > Correspondence to biological function

G protein-coupled receptor kinase 2 (GRK2). However, during chronic heart failure GRK2 is upregulated and believed to contribute to disease progression. We have determined crystallographic structures of GRK2 bound to an RNA aptamer that potently and selectively inhibits kinase activity. Key to the mechanism of inhibition is the positioning of an adenine nucleotide into the ATP-binding pocket and interactions with the basic α F- α G loop region of the GRK2 kinase domain.

Tesmer, Lennarz, Mayer, Tesmer (2012) Structure 20(8):1300-9.

pymol : GPCR + aptamer

How to check if any “predicted good structure” is really good ?

- > Visualization: *large interface*
good shape complementarity (no holes at interface)
clustering at same area
- > Correspondence to biological function
- > Mutagenesis on amino-acids at the interface
 - should change the binding affinity
 - should not disrupt the individual structures !!

Experimental data on protein – RNA interfaces

The docking problem: Sampling
Evaluation of docking models
Scoring

Flexibility: Flexible docking
Fragment-based docking

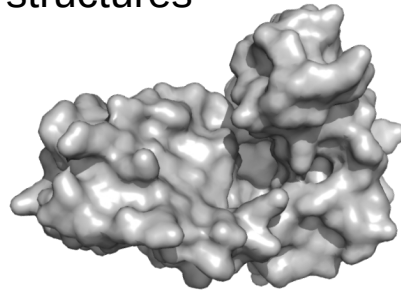
Data-driven docking: Contact/interface -driven
Fitting in 3D shape

Other docking paradigms

Modeling pipeline

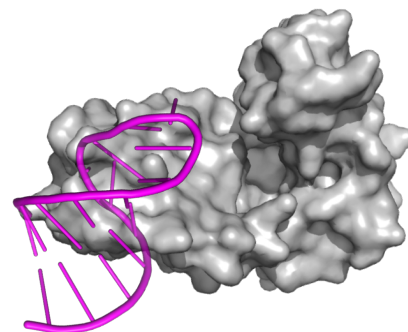
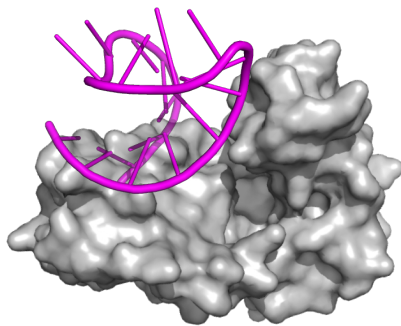
Known protein – RNA interaction

Experimental unbound structures



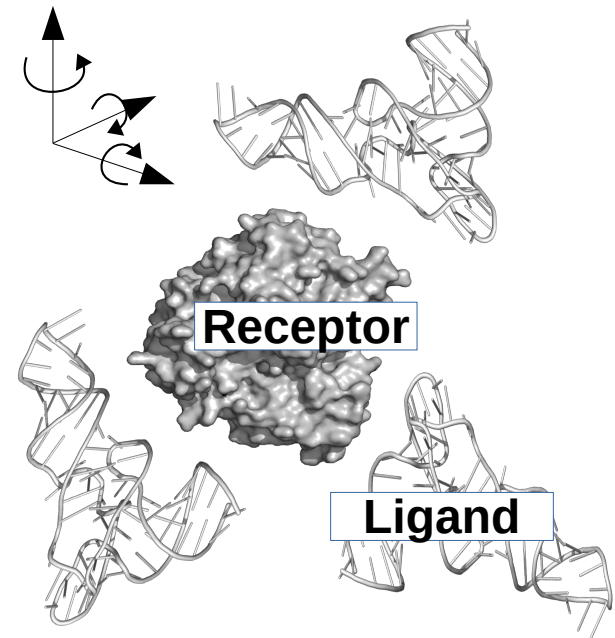
Rigid docking

models



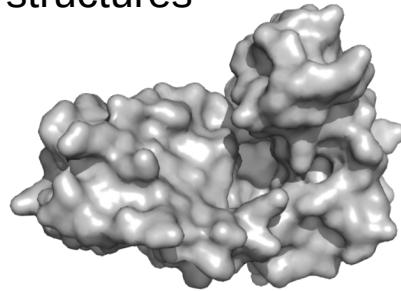
The rigid approximation:

- The molecules are kept **rigid**
- receptor is fixed, ligand has **6 DOF**
- We neglect intra-molecular energies
- We minimize the **interaction energy**



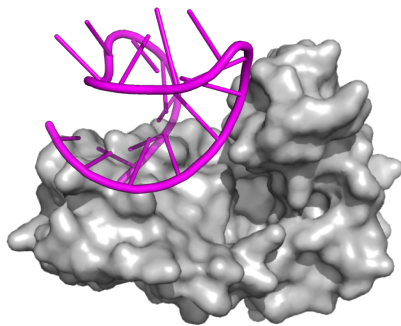
Known protein – RNA interaction

Experimental unbound structures



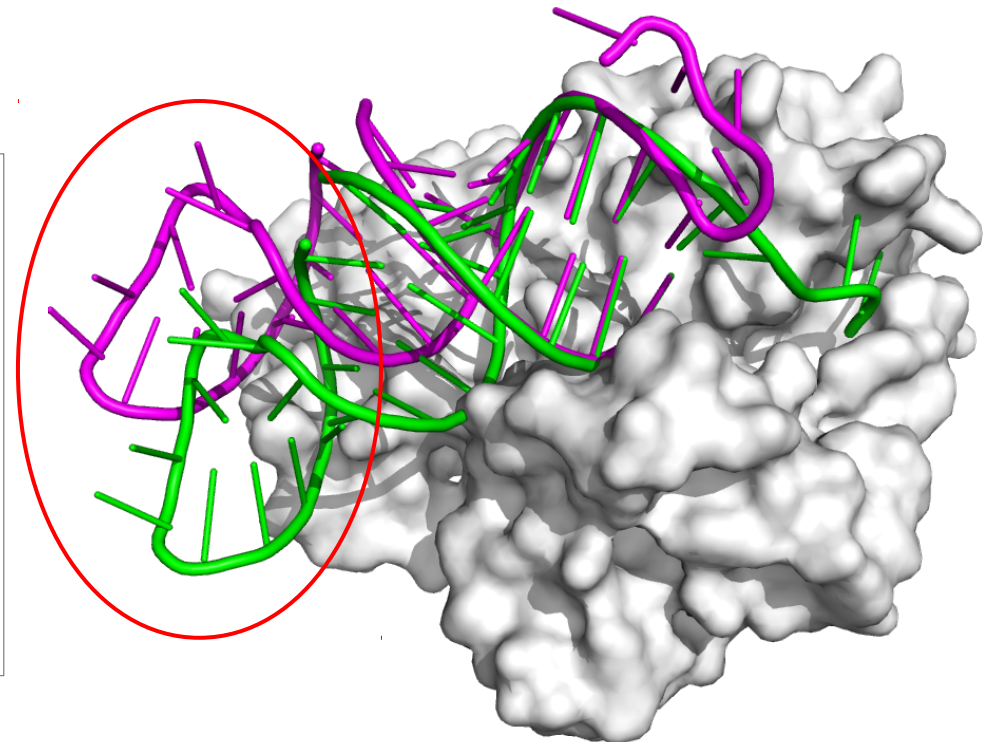
Rigid docking

models



~~The rigid approximation:~~

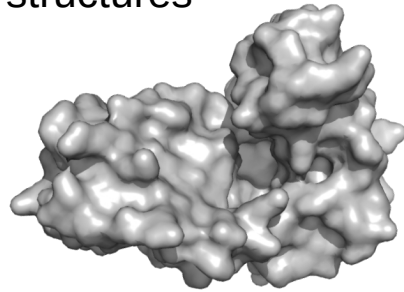
- ~~• The molecules are kept rigid~~
- ~~• receptor is fixed, ligand has 6 DOF~~
- ~~• We neglect intra-molecular energies~~
- ~~• We minimize the interaction energy~~



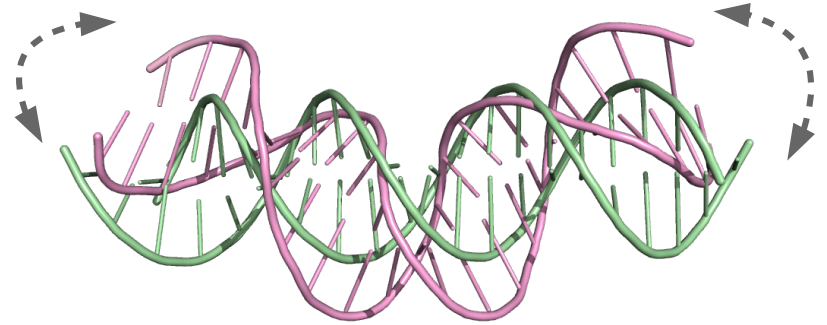
Flexibility

Known protein – RNA interaction

Experimental unbound structures

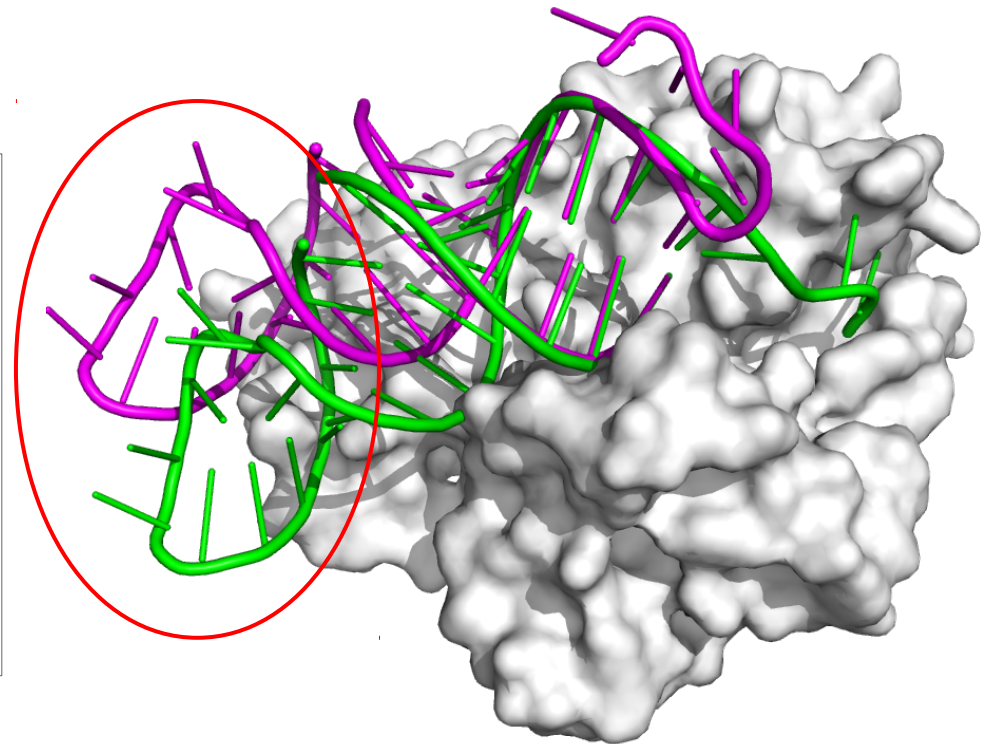
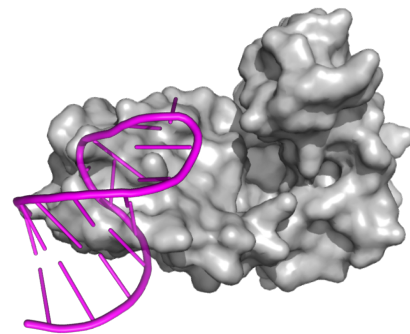
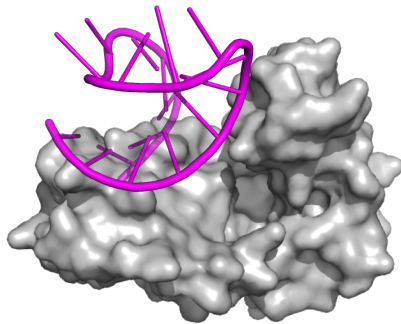


Flexible docking



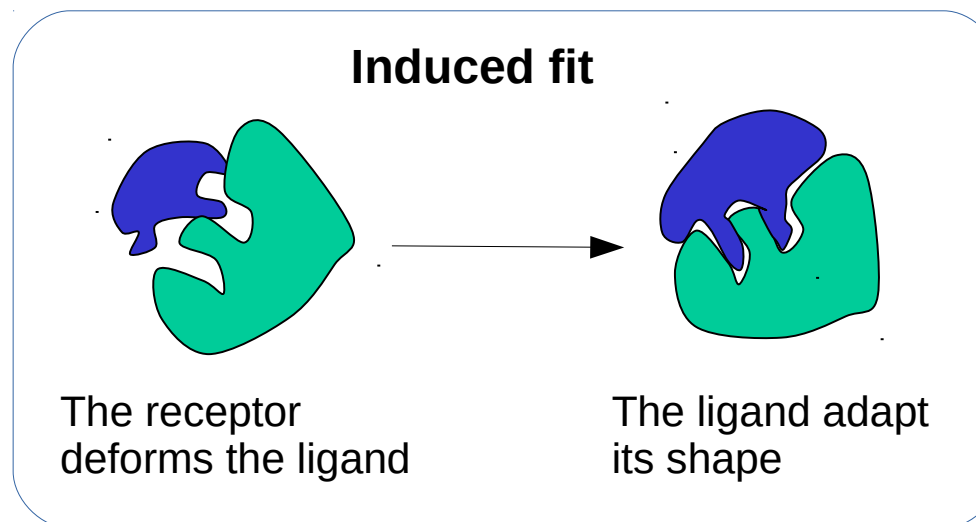
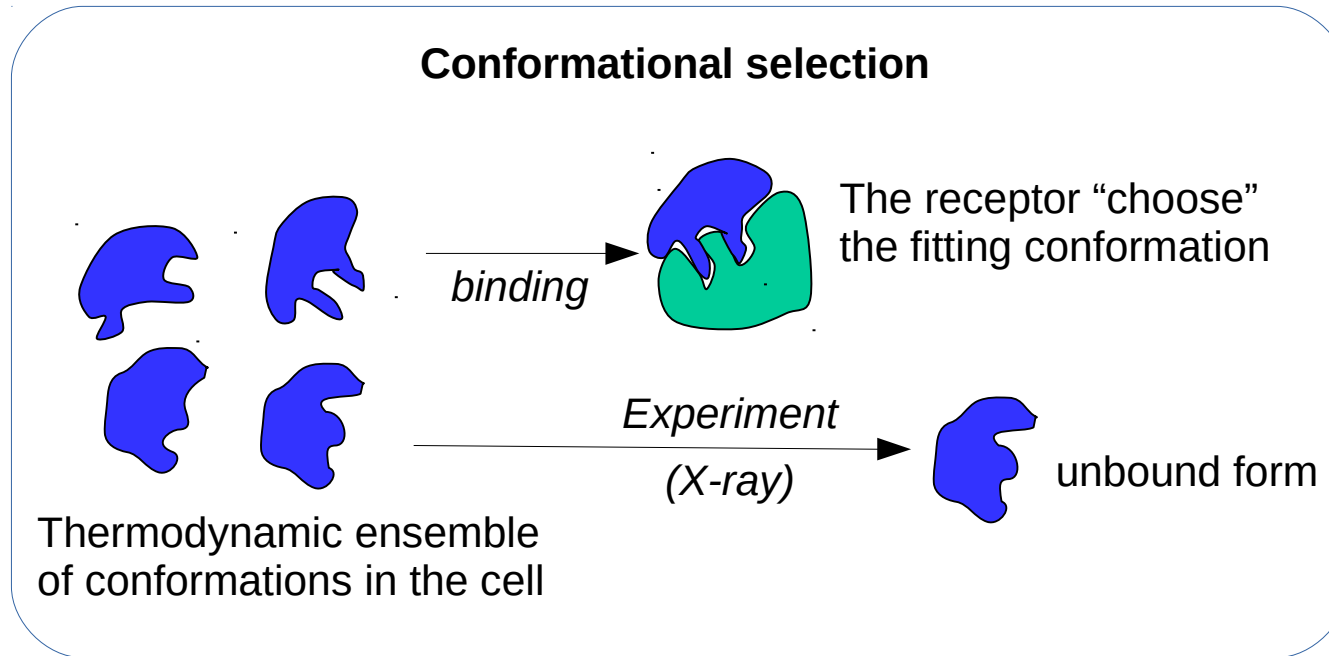
Flexible docking

models



Flexibility

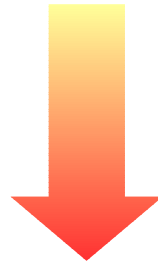
Conformational changes upon binding occurs by (a combination of) 2 ways :



Flexibility

Proteins

Side chain (protein surface)
Backbone in loop regions
Secondary structures
Domains



amplitude

RNA

Base flipping
Backbone in loop regions
Secondary structures
Unstructured regions (single-strand)

Flexibility

Proteins

Side chain (protein surface)

Backbone in loop regions

Secondary structures

Domains

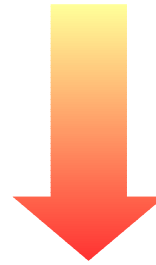
RNA

Base flipping

Backbone in loop regions

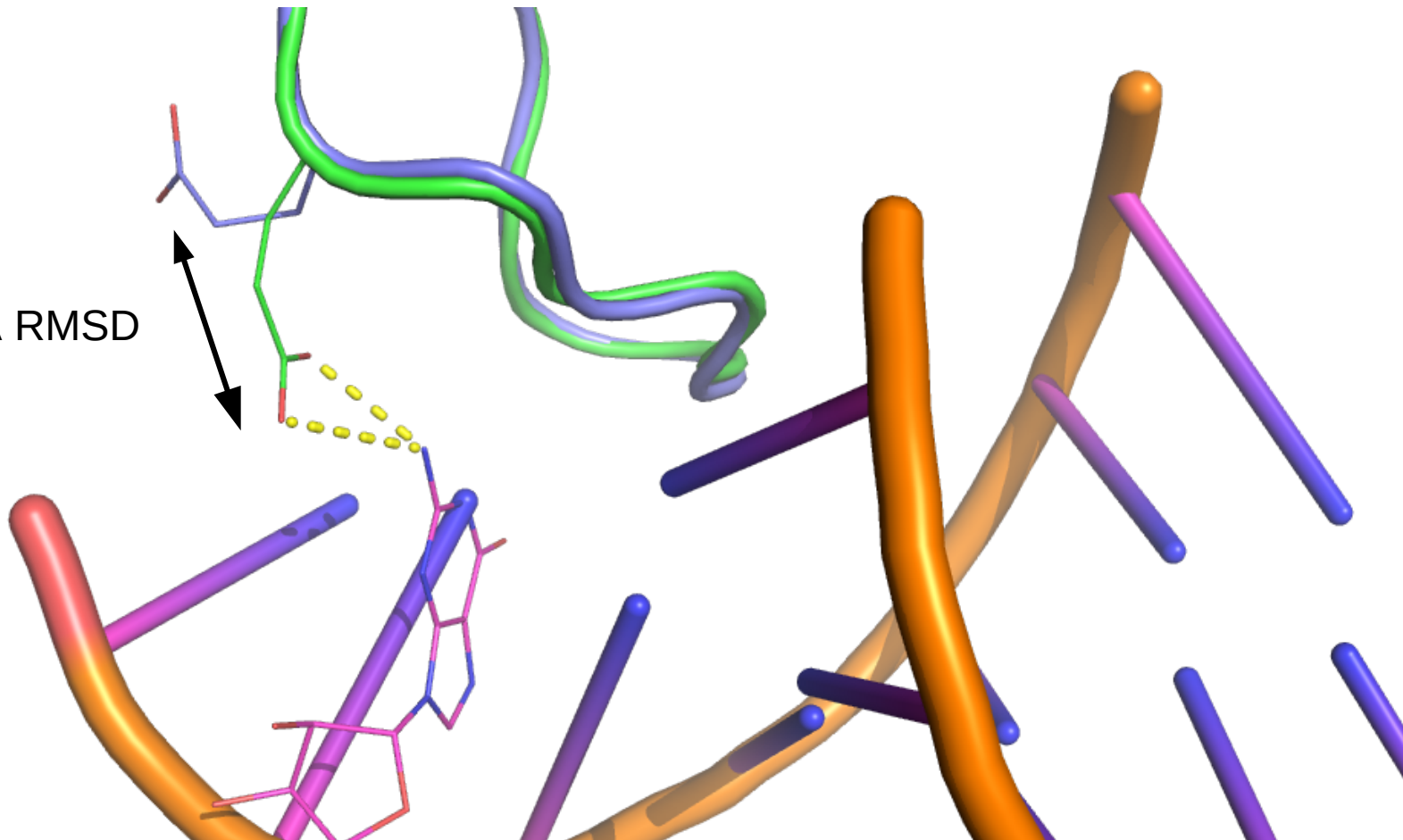
Helices bend / twist

Unstructured regions (single-strand)



amplitude

4 Å RMSD



Flexibility

Proteins

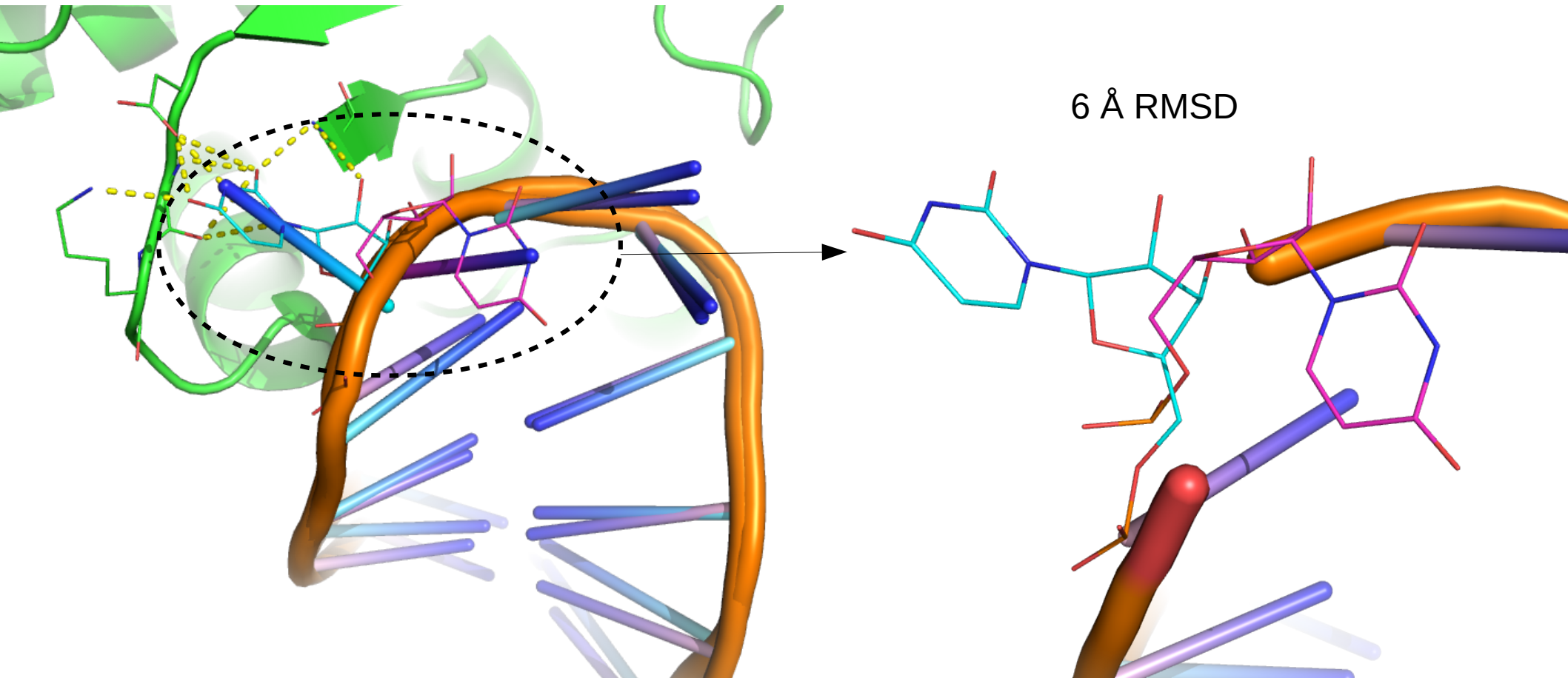
Side chain (protein surface)
Backbone in loop regions
Secondary structures
Domains

RNA

Base flipping
Backbone in loop regions
Helices bend / twist
Unstructured regions (single-strand)



amplitude



Flexibility

Proteins

Side chain (protein surface)

Backbone in loop regions

Secondary structures

Domains

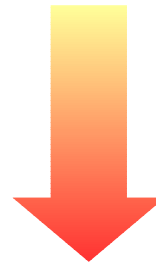
RNA

Base flipping

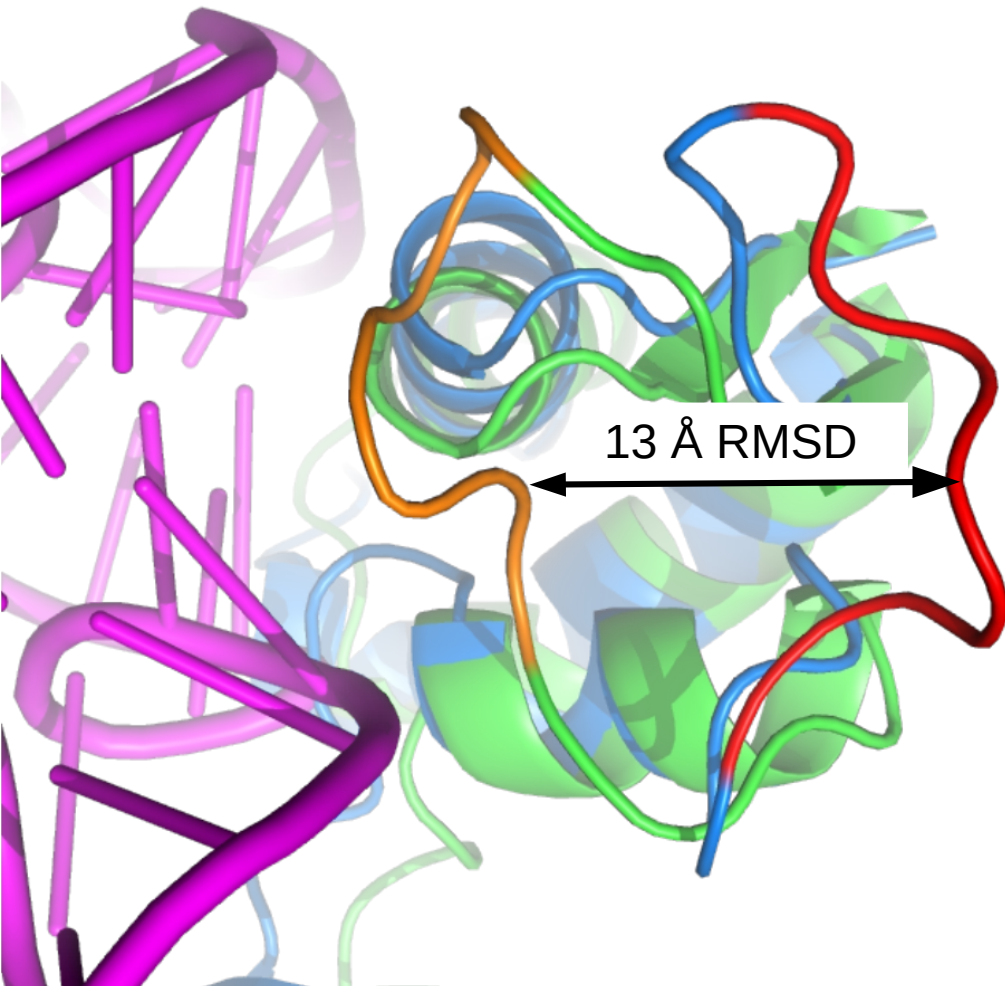
Backbone in loop regions

Helices bend / twist



Unstructured regions (single-strand)



amplitude



13 Å RMSD

 unbound
 bound

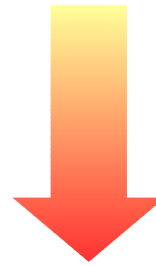
Flexibility

Proteins

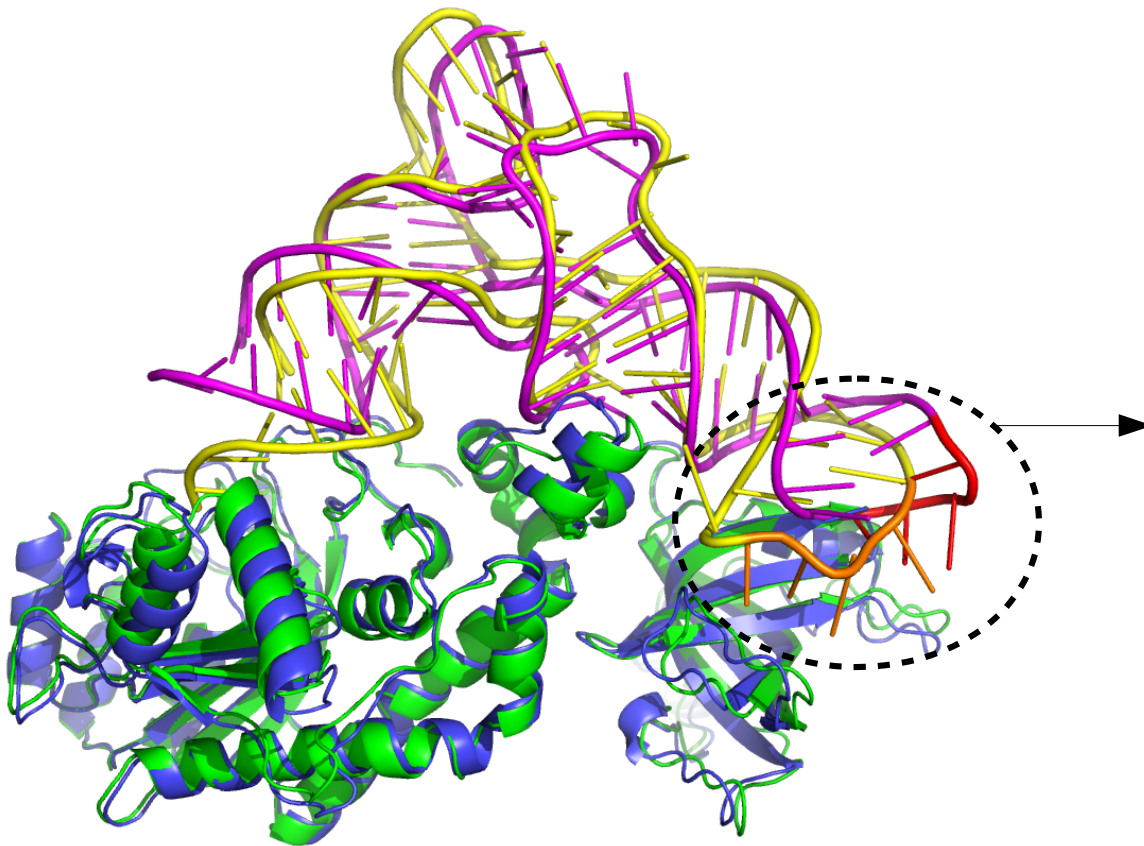
Side chain (protein surface)
Backbone in loop regions
Secondary structures
Domains

RNA

Base flipping
Backbone in loop regions
Helices bend / twist
Unstructured regions (single-strand)



amplitude



13 Å RMSD



Flexibility

Proteins

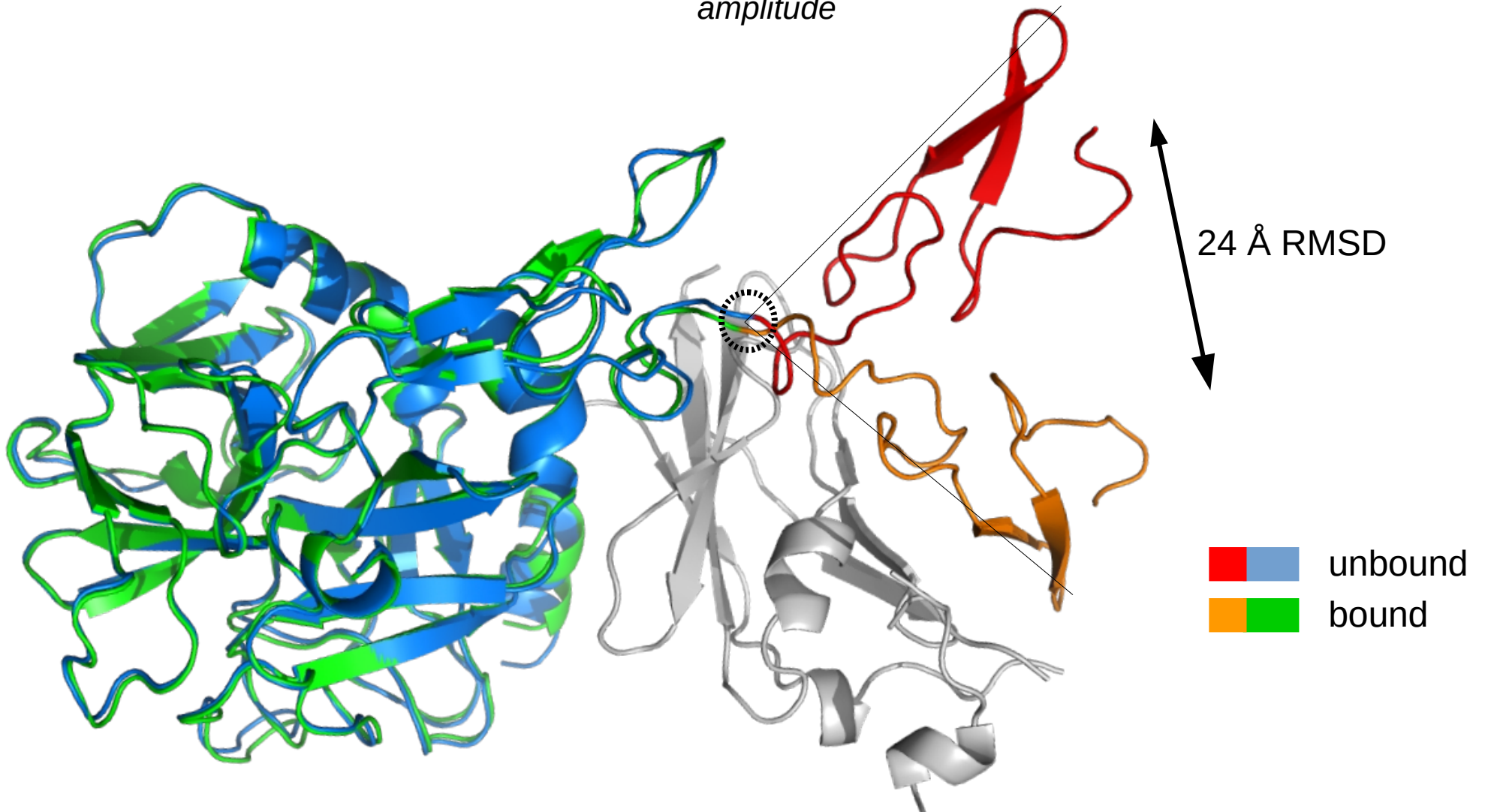
Side chain (protein surface)
Backbone in loop regions
Secondary structures
Domains

RNA

Base flipping
Backbone in loop regions
Helices bend / twist
Unstructured regions (single-strand)



amplitude



Flexibility

Proteins

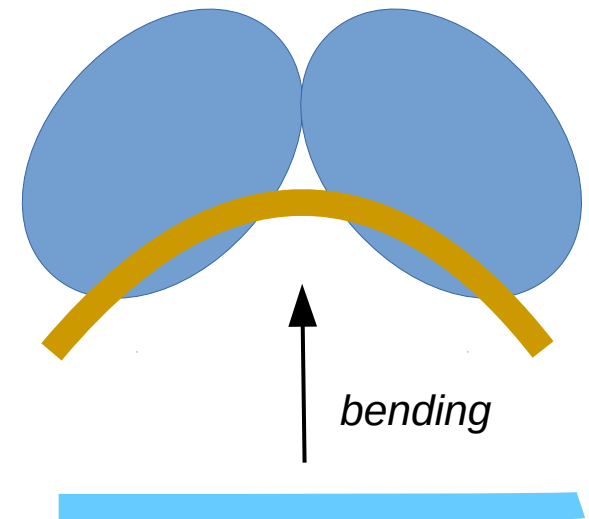
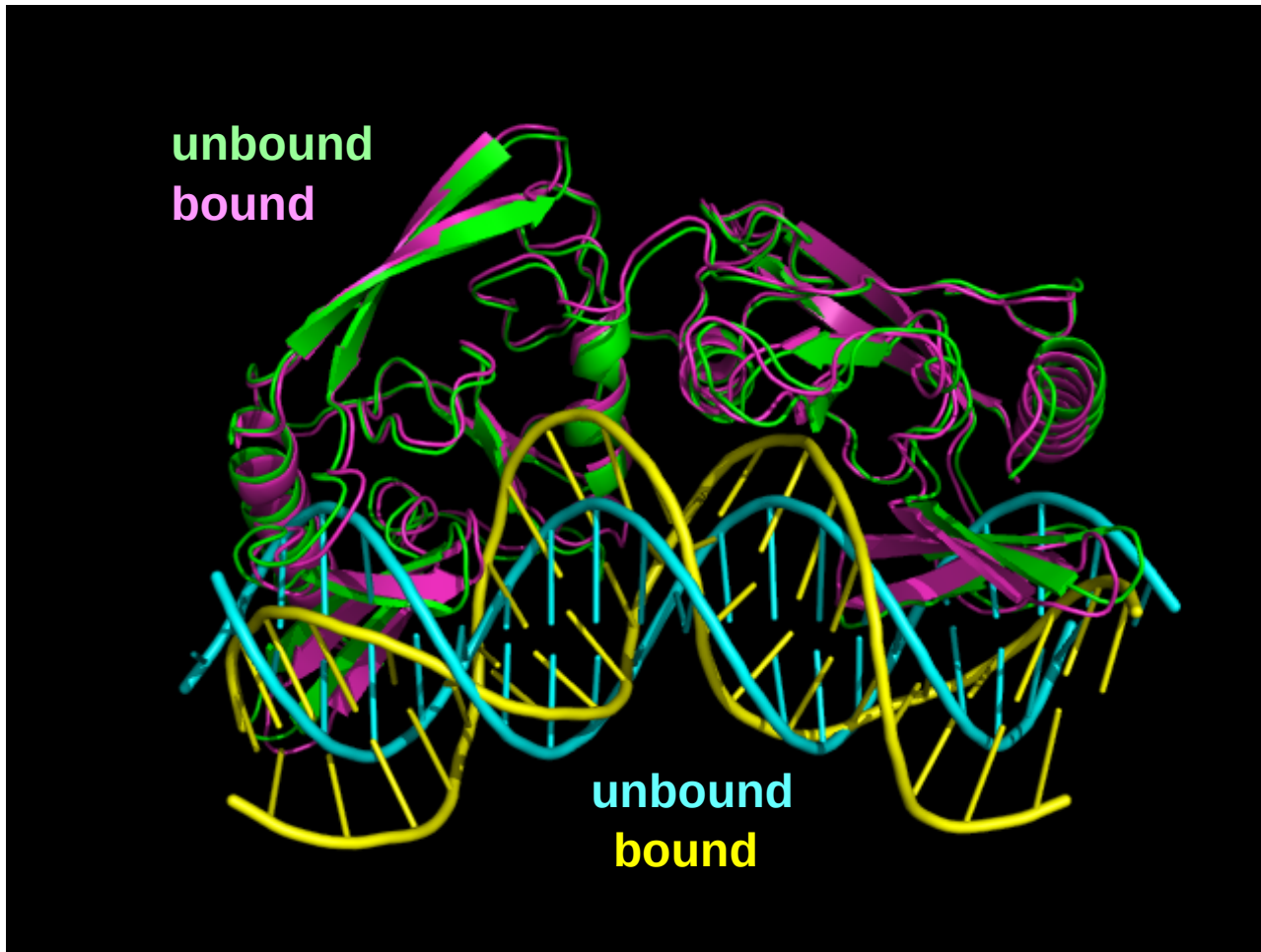
Side chain (protein surface)
Backbone in loop regions
Secondary structures
Domains

RNA

Base flipping
Backbone in loop regions
Helices bend / twist
Unstructured regions (single-strand)



amplitude



Flexibility

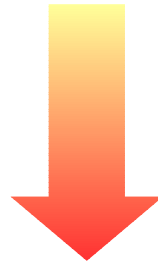
Proteins

Side chain (protein surface)
Backbone in loop regions
Secondary structures

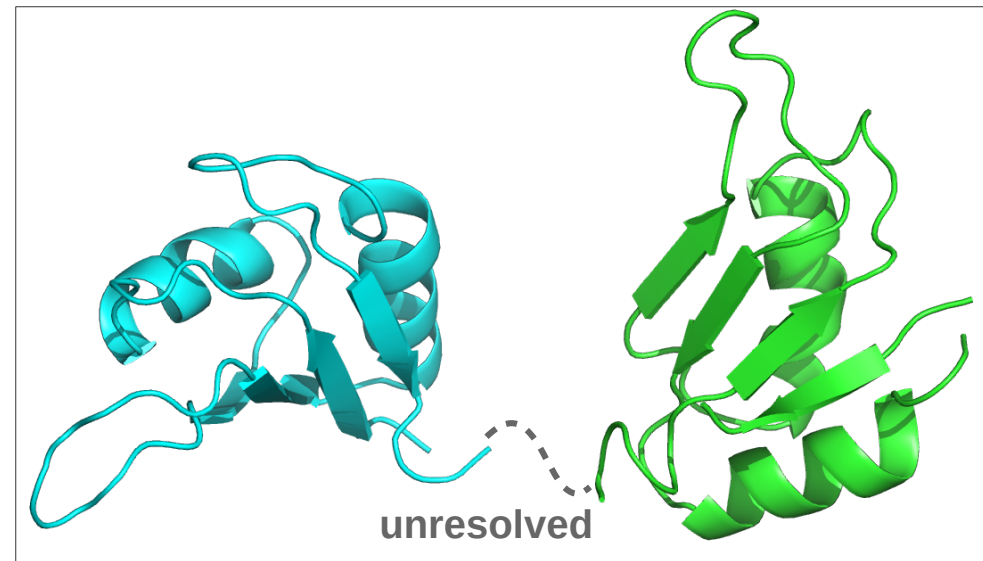
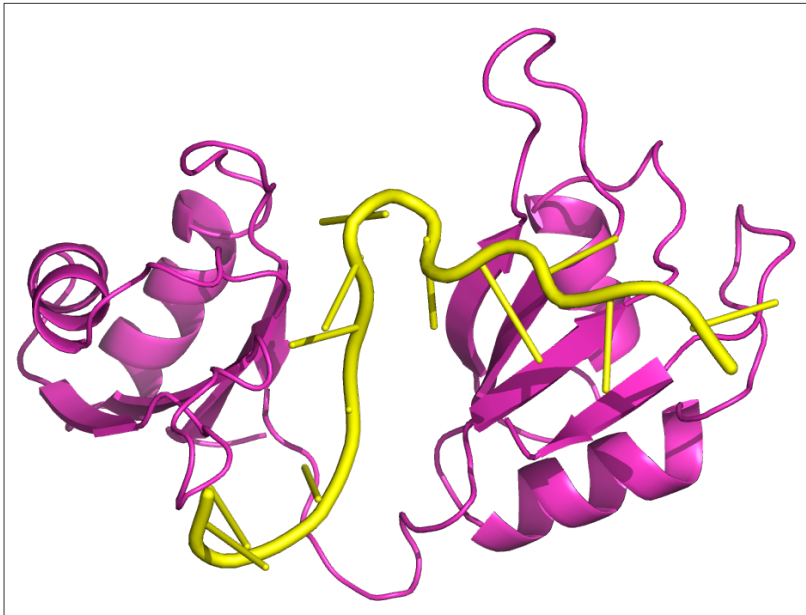
Domains

RNA

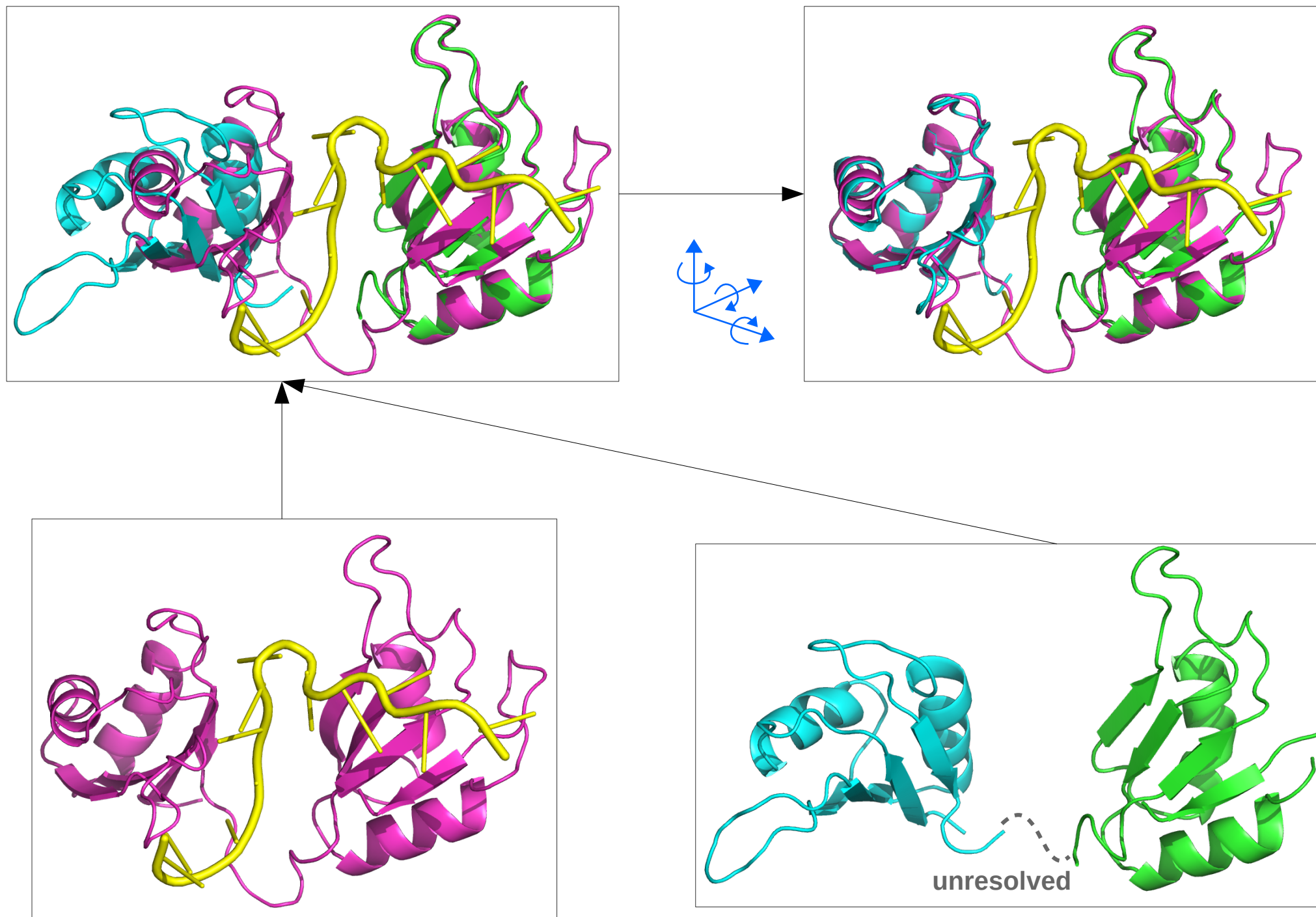
Base flipping
Backbone in loop regions
Helices bend / twist
Unstructured regions (single-stranded)



amplitude



Flexibility



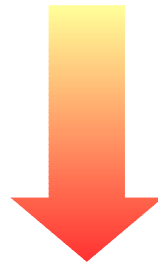
Flexibility

Proteins

Side chain (protein surface)
Backbone in loop regions
Secondary structures
Domains

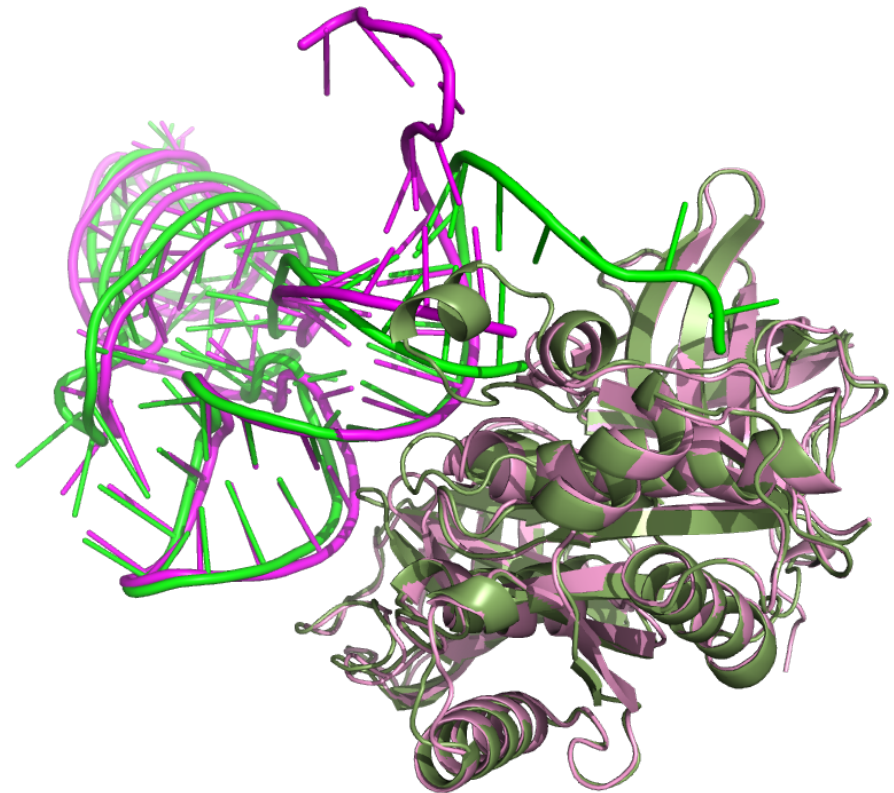
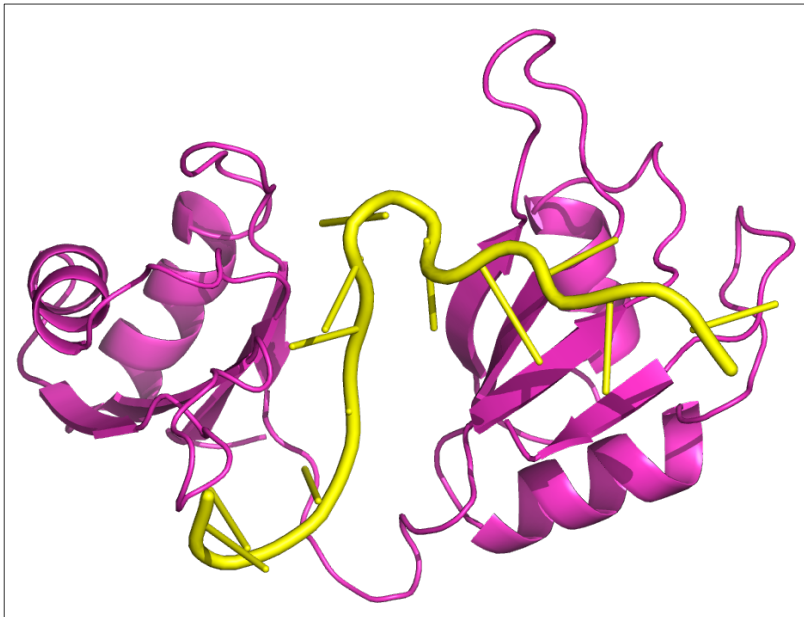
RNA

Base flipping
Backbone in loop regions
Helices bend / twist
Unstructured regions (single-stranded)



amplitude

Undefined in solution

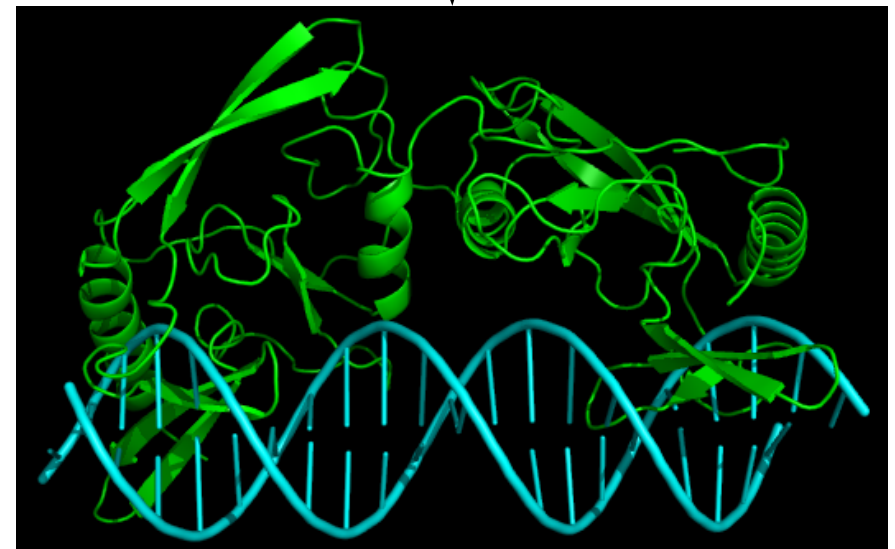
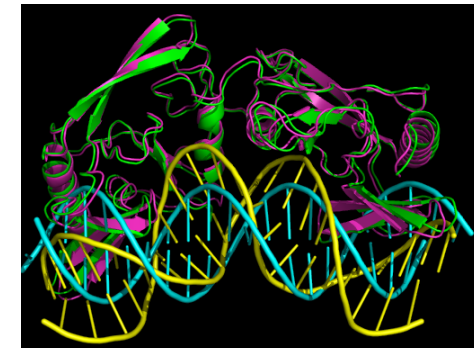
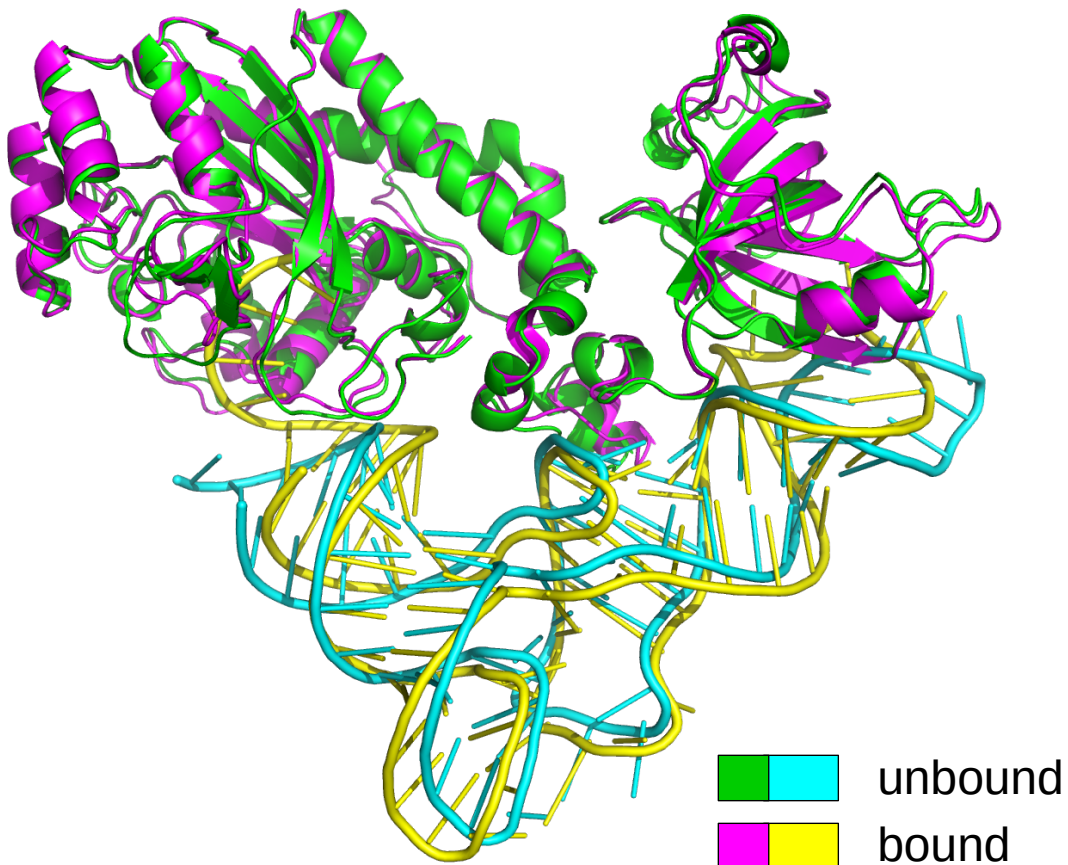


Flexibility

The perfect superpositions of unbound structures on the bound complex gives the upper limit of model quality that can be obtained by rigid body modeling.

If the perfect fit brings atomic clashes, it will not be found by docking.

Best fit : 3.2 Å



Experimental data on protein – RNA interfaces

The docking problem: Sampling
Evaluation of docking models
Scoring

Flexibility: Flexible docking
Fragment-based docking

Data-driven docking: Contact/interface -driven
Fitting in 3D shape

Other docking paradigms

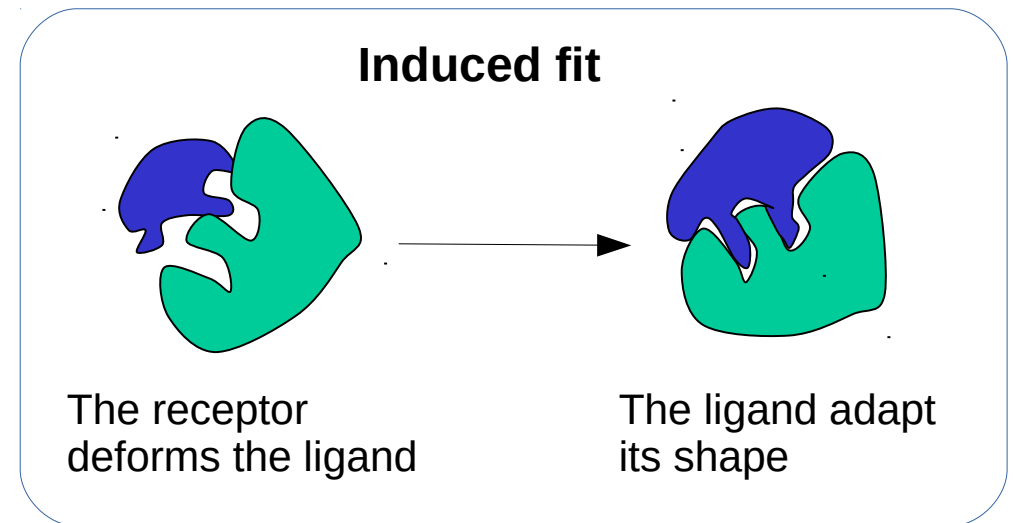
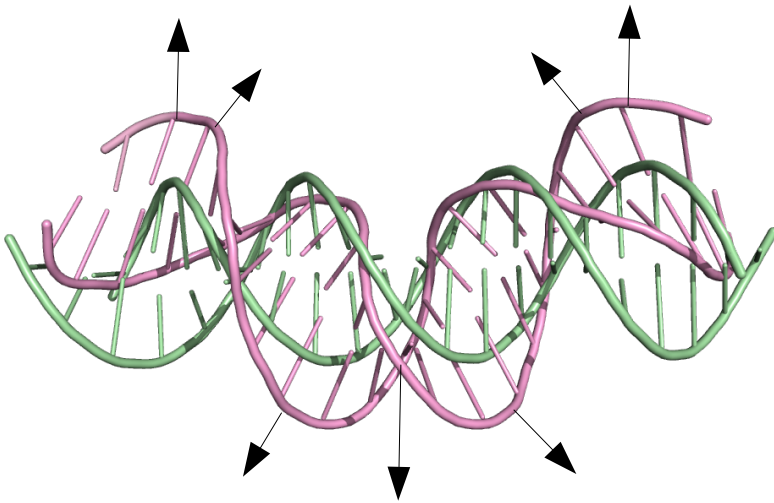
Modeling pipeline

Harmonic modes

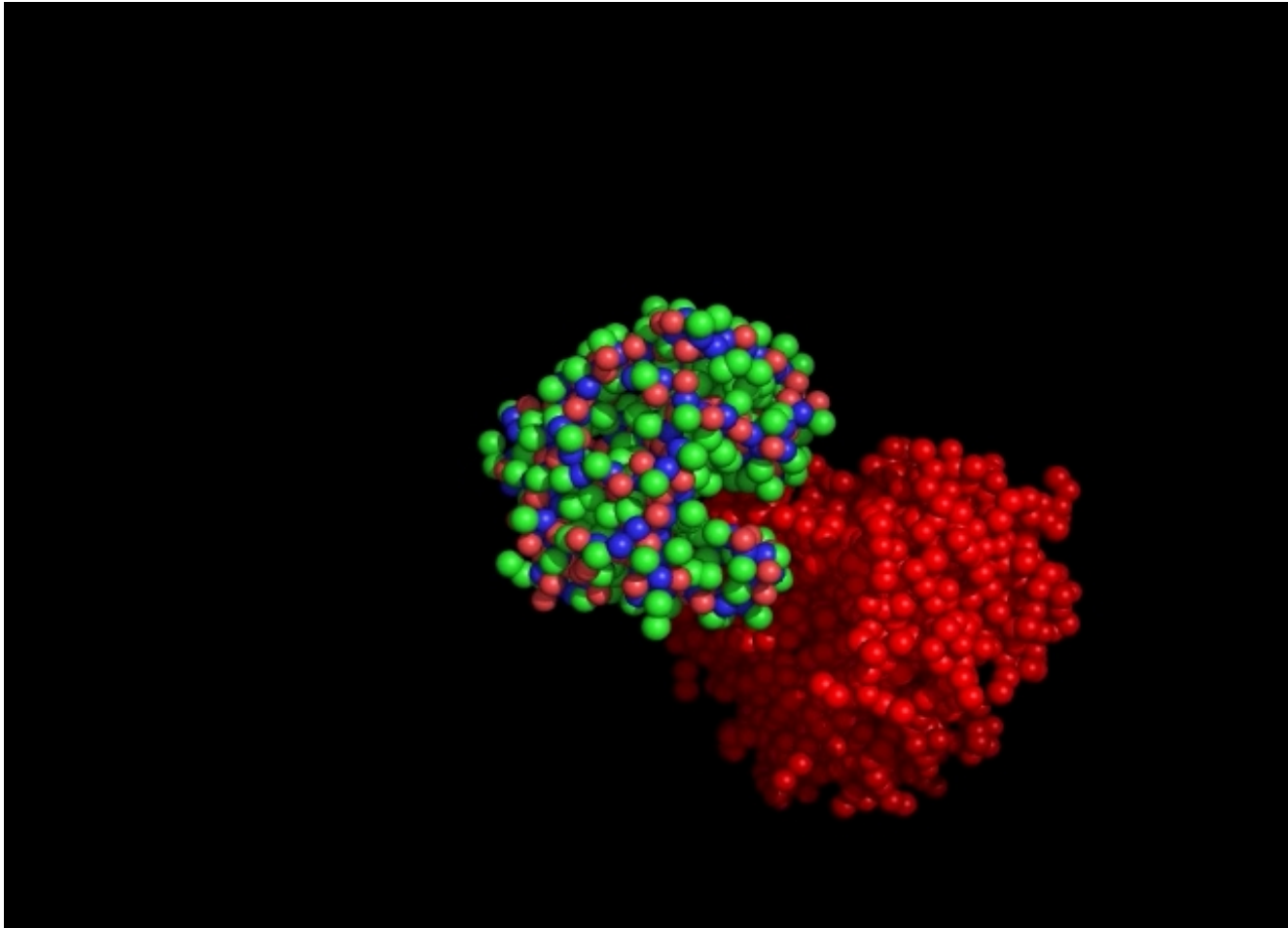
Energetically favorable directions of deformation, orthogonal to each-other.

- Computed based on the structure of each molecule and its inter-atomic forces.
- Added as additional DOF along the minimization.

Hopefully one or few modes will bring the unbound RNA closer to the bound form (but it is not guaranteed !).



ATTRACT: harmonic modes

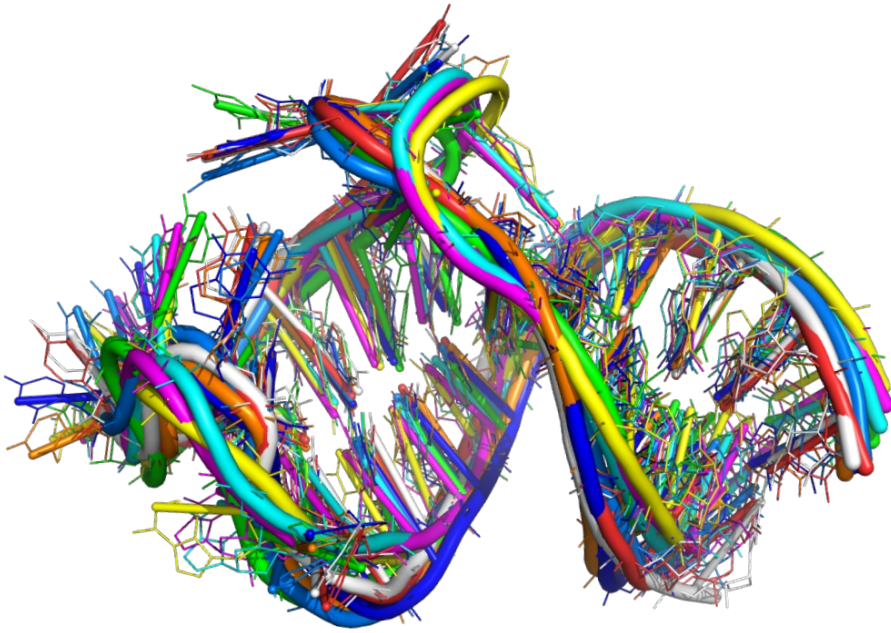


Conformational ensemble

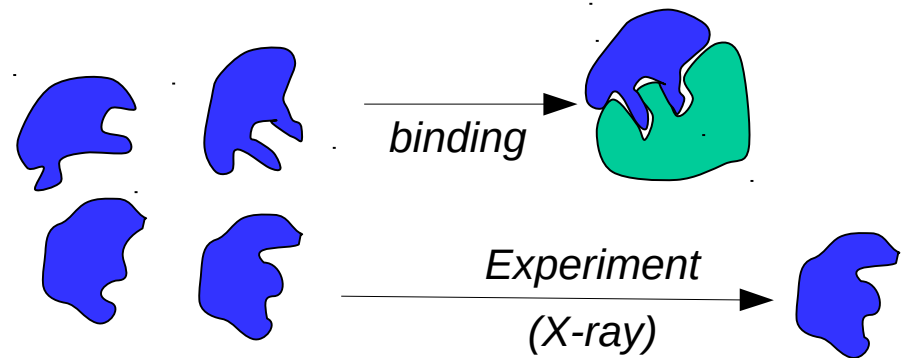
Ensemble of unbound conformations docked in parallel.

Can be obtained from Molecular Dynamics, NMR ...

Increase the probability to be close enough to the bound form (but it is not guaranteed !),



Conformational selection



Refinement

After docking, use MD with all-atom force-field to refine models

=> Only small adjustments. Needs to have already a quite correct solution

!! time-consuming

Experimental data on protein – RNA interfaces

The docking problem: Sampling
Evaluation of docking models
Scoring

Flexibility: Flexible docking

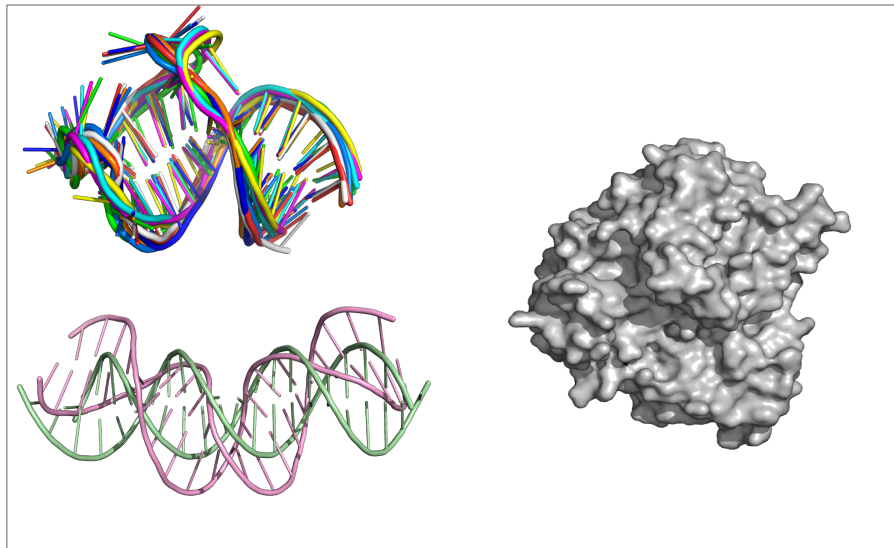
Fragment-based docking

Data-driven docking: Contact/interface -driven
Fitting in 3D shape

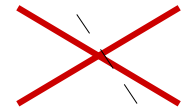
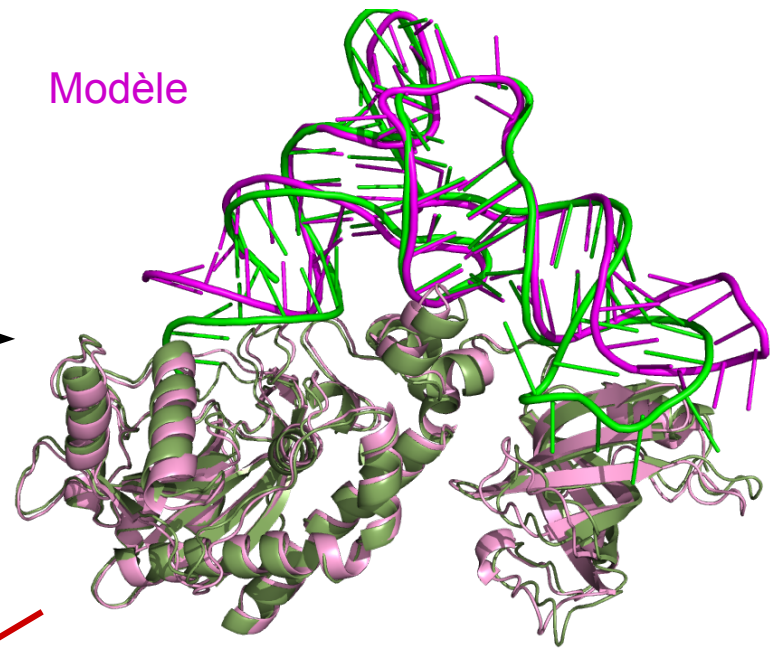
Other docking paradigms

Modeling pipeline

Fragment – based docking



Flexible
docking

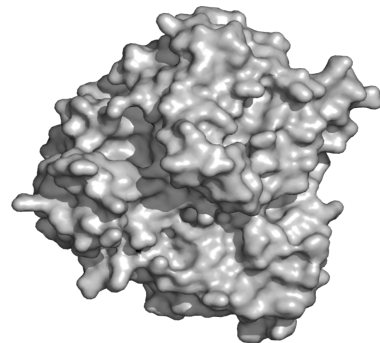


Known RNA sequence

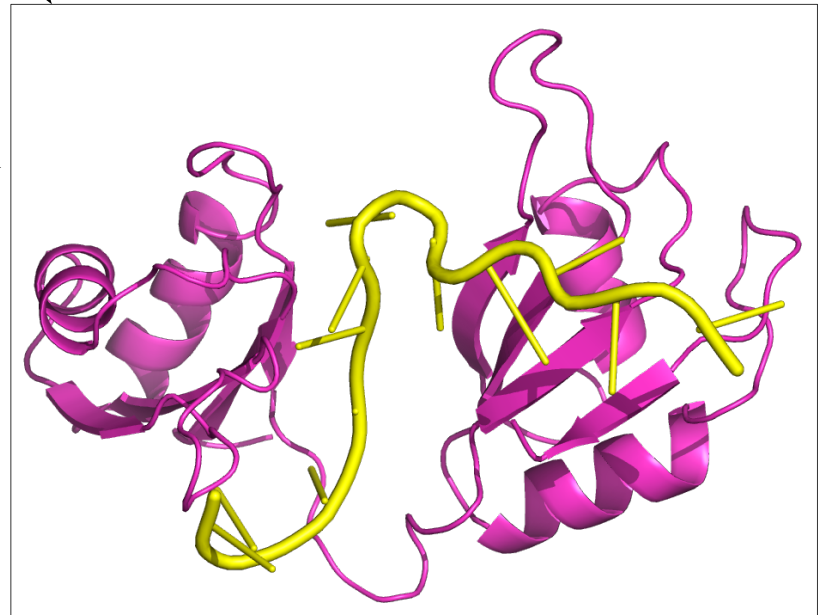
ssRNA: Disordered

?

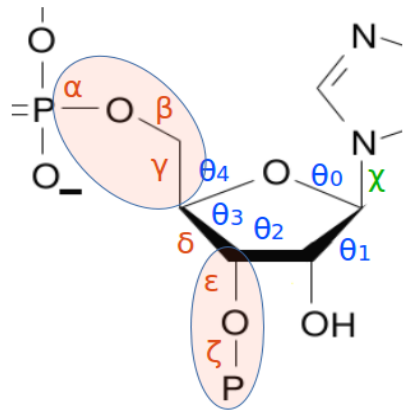
$|\{\text{conformations}\}| \sim \infty$



~~Flexible
docking~~



Fragment – based docking



- 12 DOF per nucleotide => combinatorial explosion
- Protein-induced conformations

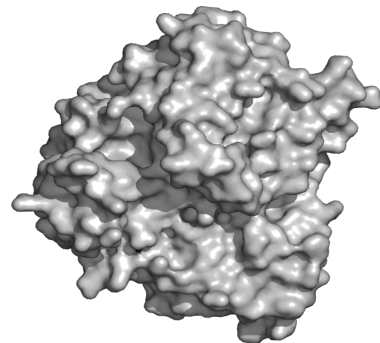


Approximation : discrete **local** conformations
=> structural fragment library

Known RNA sequence

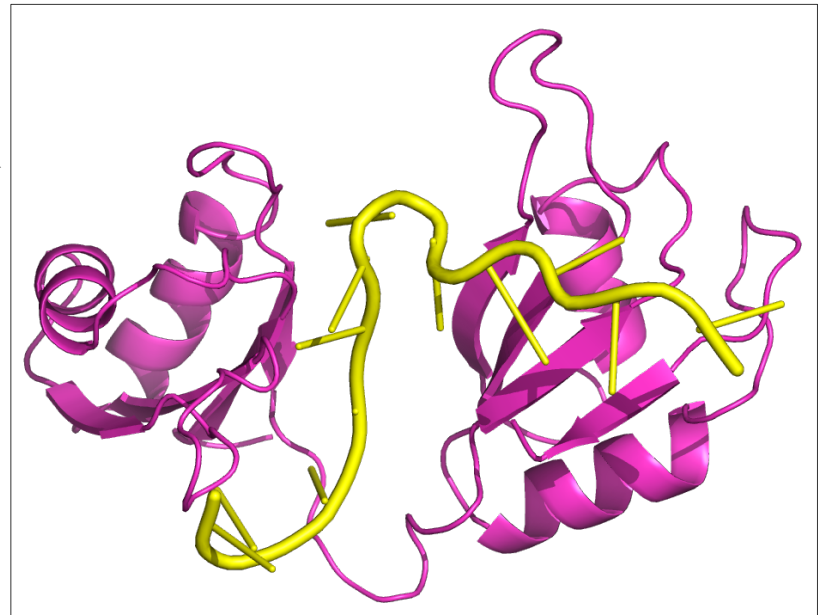
ssRNA: Disordered

?

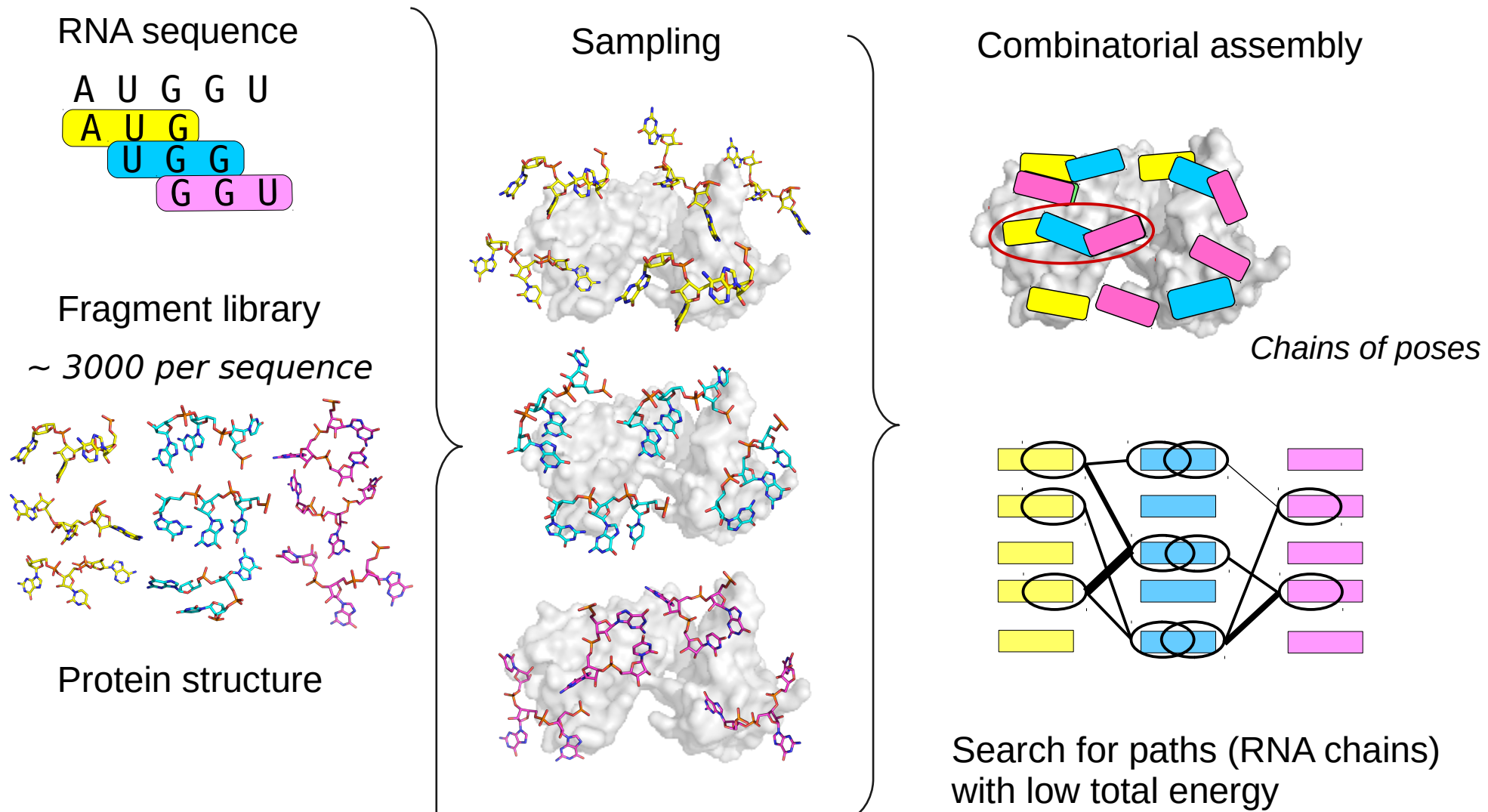


$|\{\text{conformations}\}| \sim \infty$

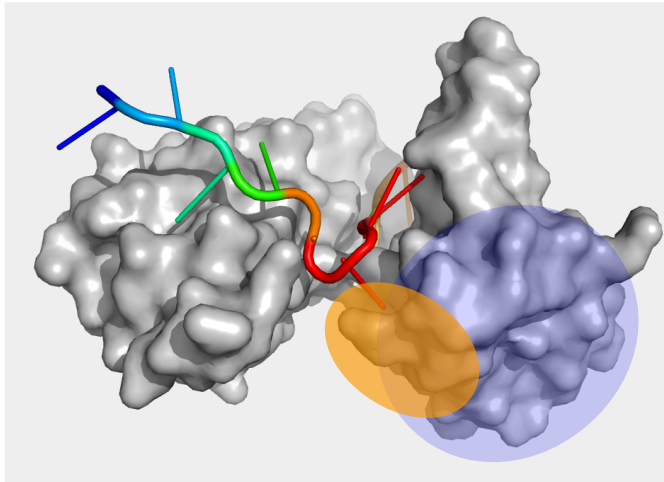
~~Flexible
docking~~



Fragment – based docking



Fragment – based docking



$\sim 10^{5-6}$ poses / fragment

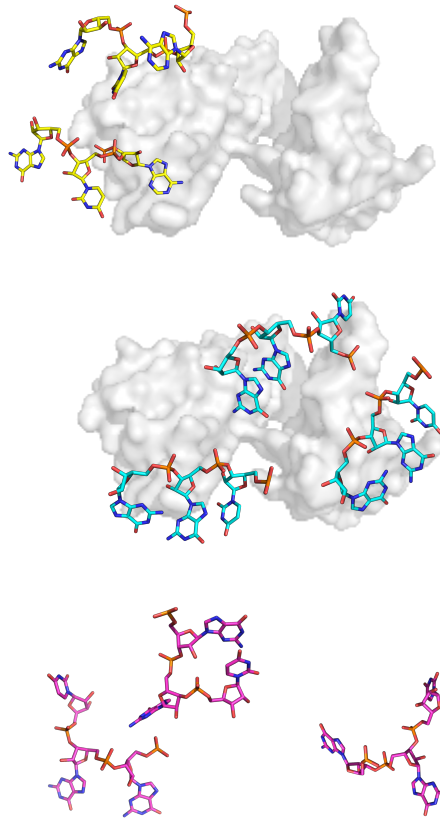
$\sim 5-8$ fragmt / RNA

$\Rightarrow 10^{25-48}$ chains

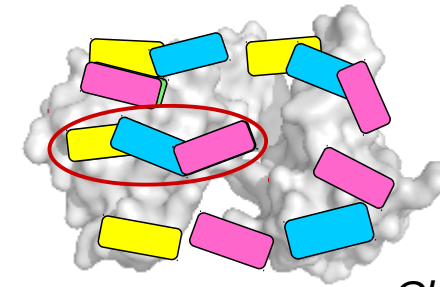
Main difficulties :

- All frag must be correctly sampled
- Scoring problem \Rightarrow lot of false positive

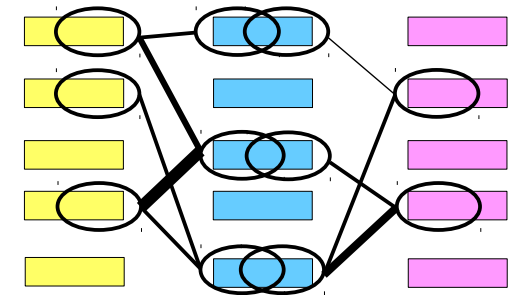
Sampling



Combinatorial assembly



Chains of poses

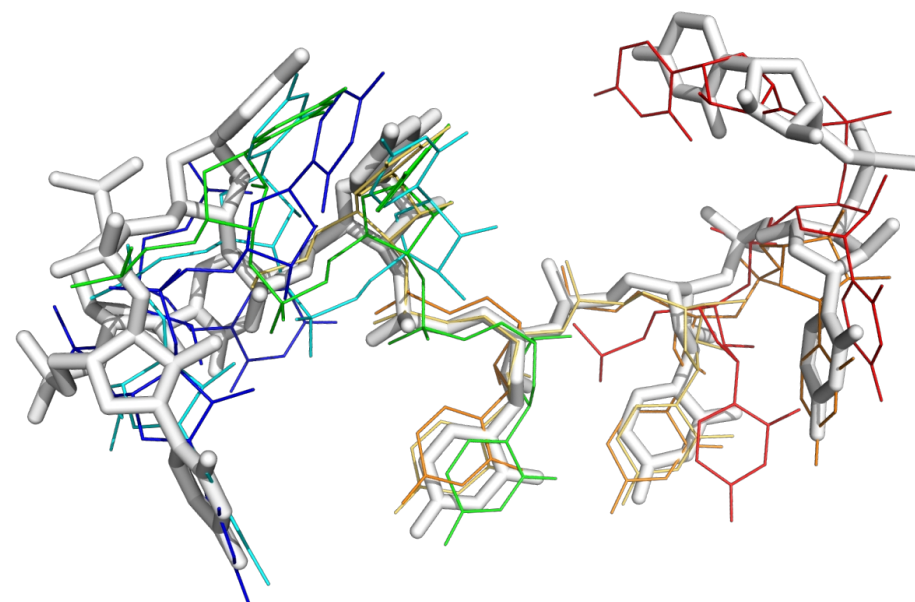
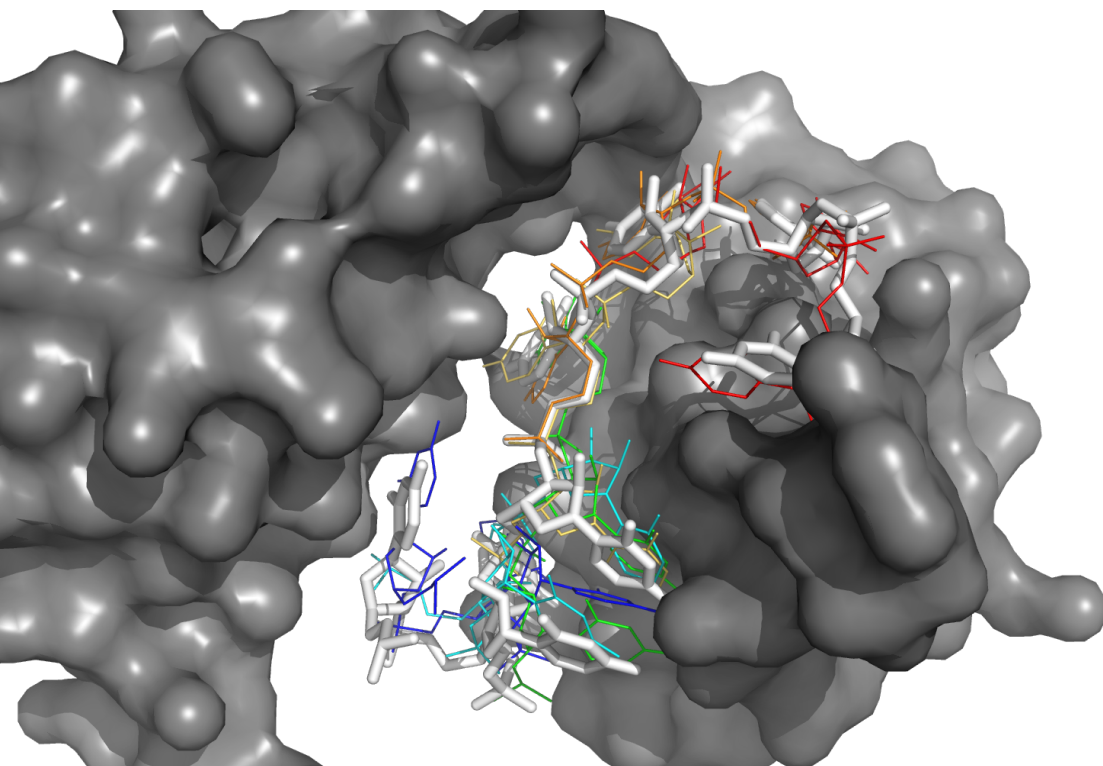


Search for paths (RNA chains)
with low total energy



test case
w. known structure

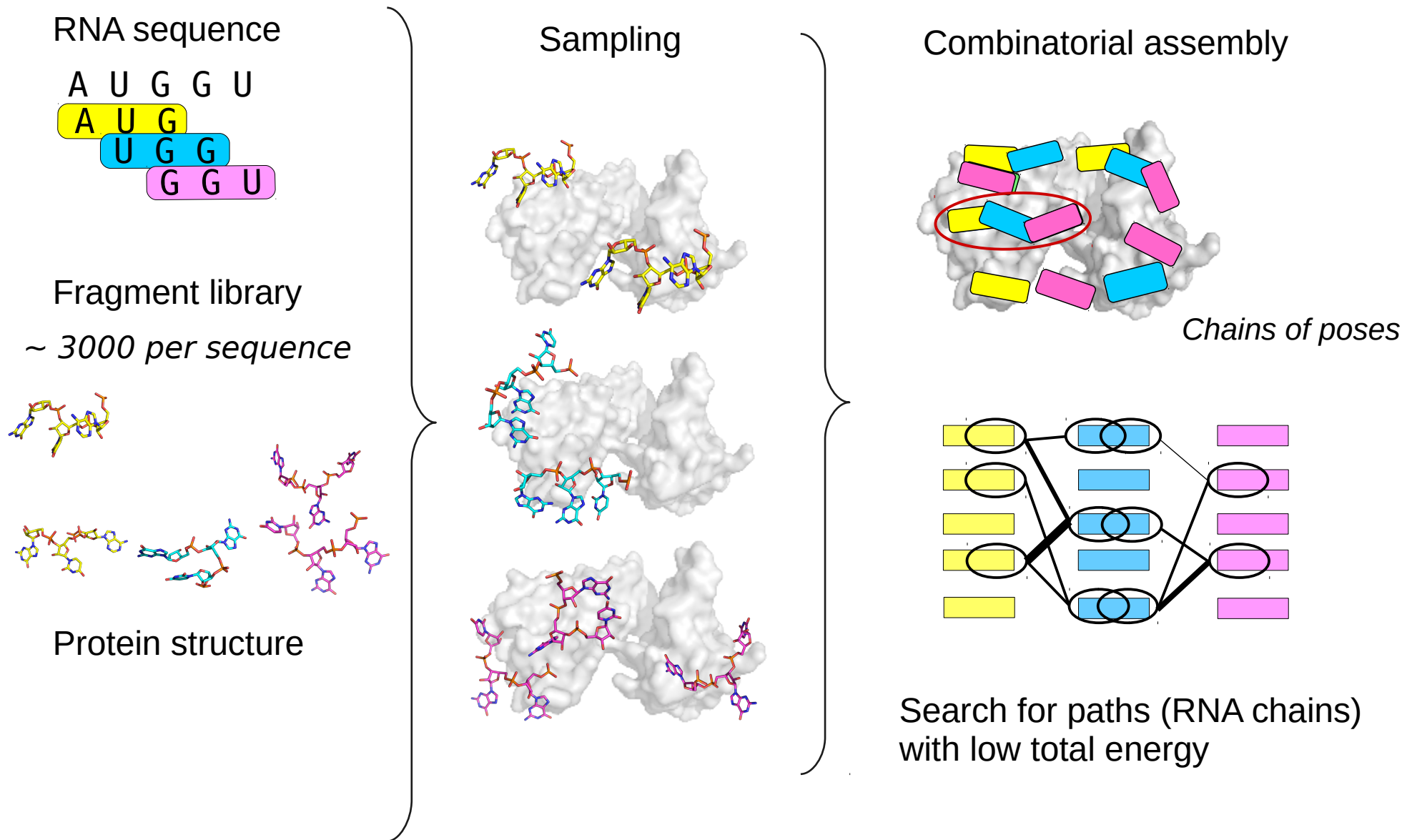
Xray **1b7f** : Sex-lethal + poly-U

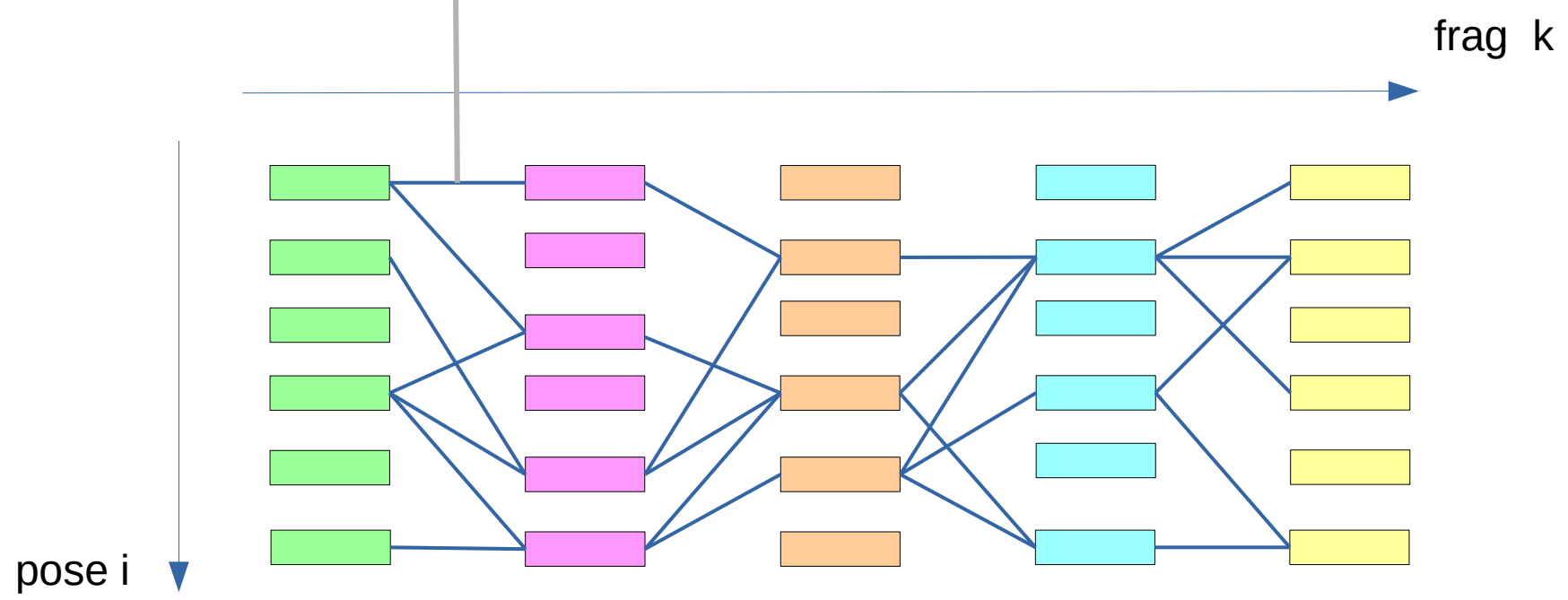
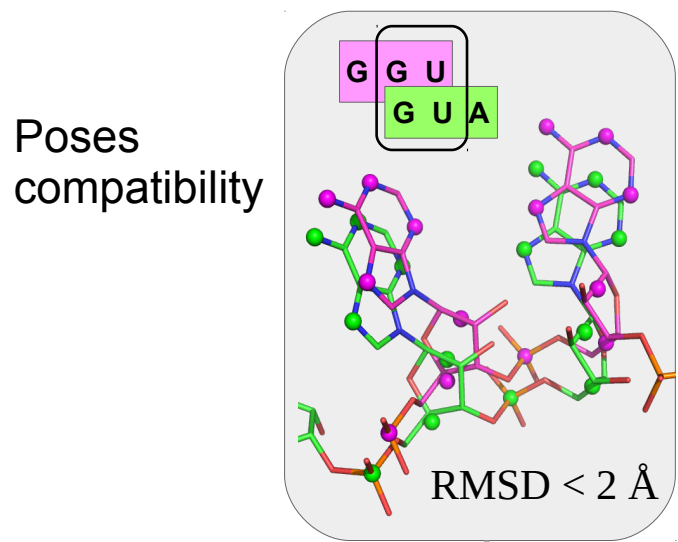


	Best RMSD
frag1	2.3 Å
frag2	1.7 Å
frag3	1.3 Å
frag4	0.9 Å
frag5	1.0 Å
frag6	1.8 Å

Total: $5 \cdot 10^5$ poses

Fragment – based docking



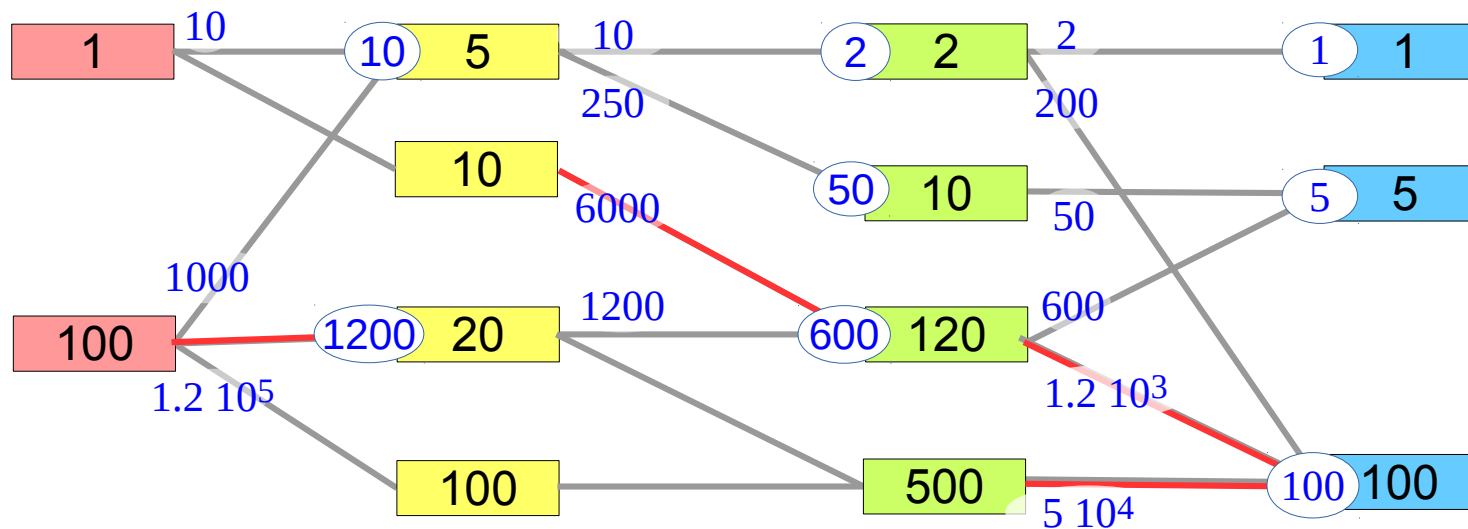


Chain sampling with mean-rank constrain

Filter connections by **dead-end elimination**

- Starting backward, consider the best-case meanrank at each pose
- For next pose, eliminate connections that cannot lead to correct meanrank
- Eliminate non-connected poses if needed
- Retain at each new pose the best-case meanrank

Ex : ($\prod \text{rank}_i$) threshold = 5000



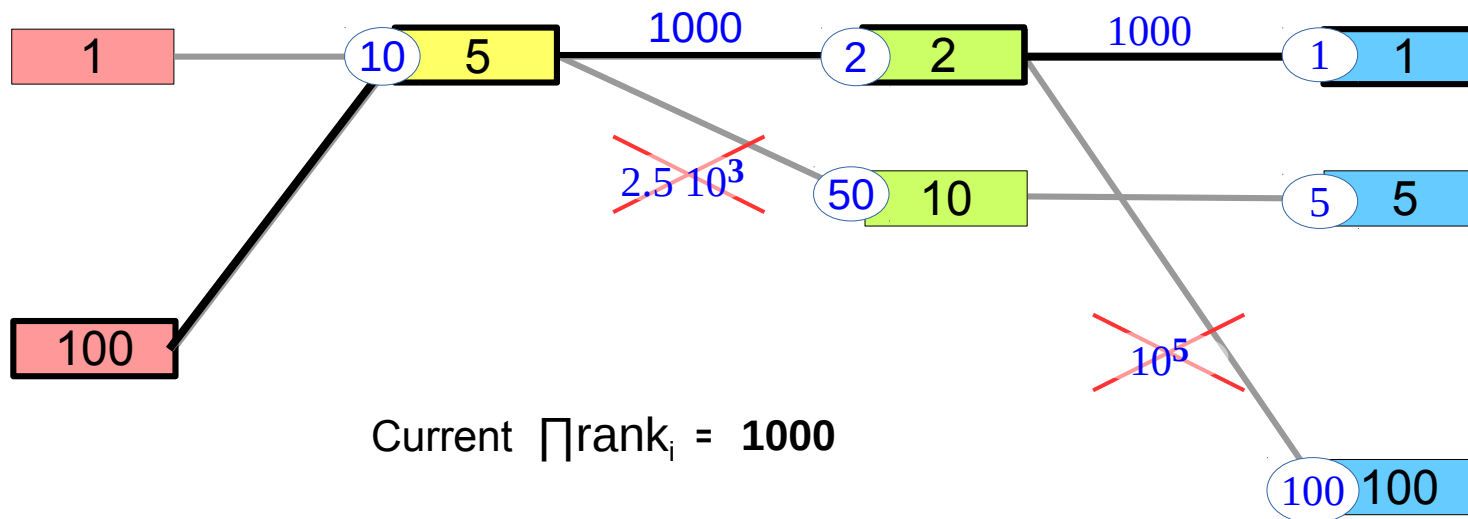
Chain sampling with mean-rank constrain

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- Eliminate non-connected poses if needed
- Retain at each new pose the best-case meanrank

Walk the filtered tree w. dead-end elimination

Ex : ($\prod \text{rank}_i$) threshold = 5000



Experimental data on protein – RNA interfaces

The docking problem: Sampling
Evaluation of docking models
Scoring

Flexibility: Flexible docking
Fragment-based docking

Data-driven docking: Contact/interface -driven
Fitting in 3D shape

Other docking paradigms

Modeling pipeline

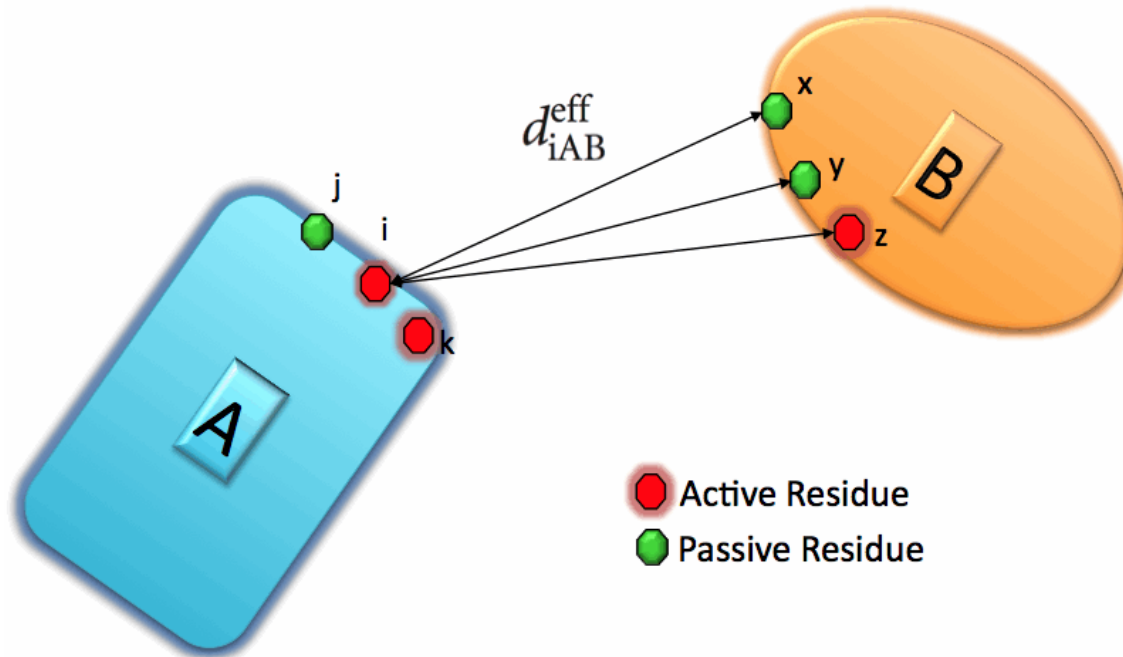
Data-driven docking

Contact / interface data

Experimental: NMR
Mutagenesis
FRET
...

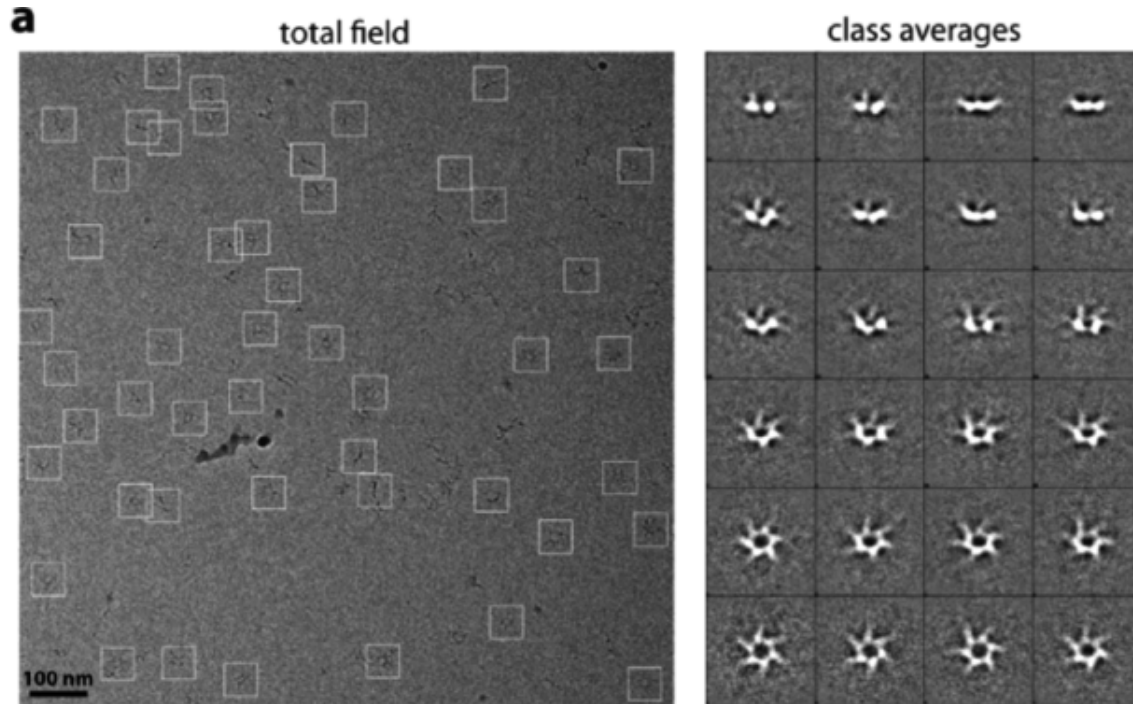
Biological: Existing mutation
Concerved contact in homologs
...

Ambiguous interaction restraints



Experimental structures

Cryo-Electron Microscopy

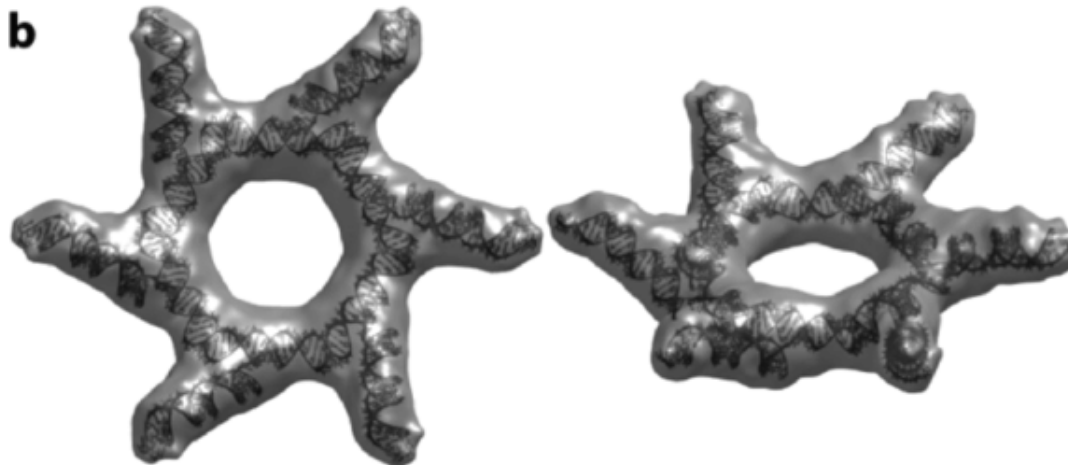


For very large systems
e.g. Ribosomes

Multiple conformations



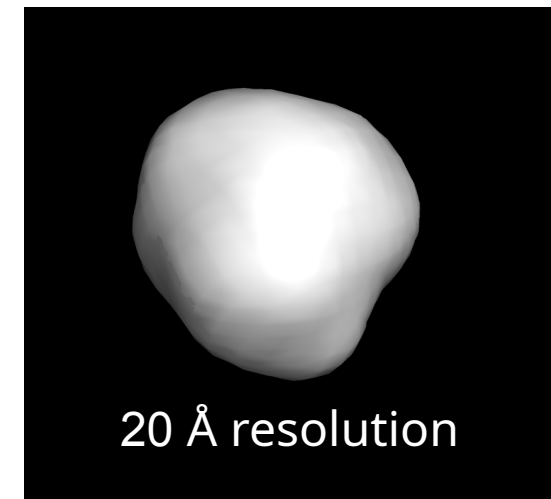
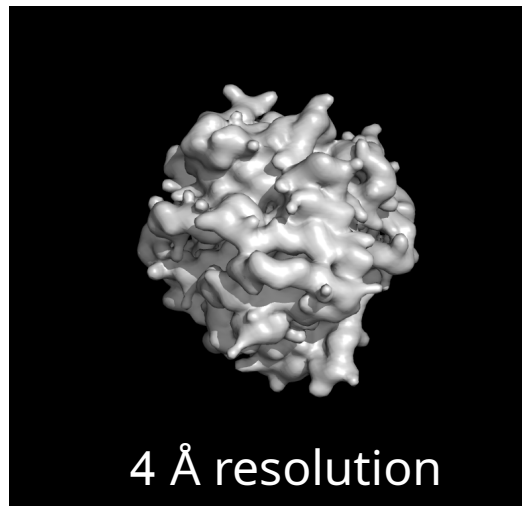
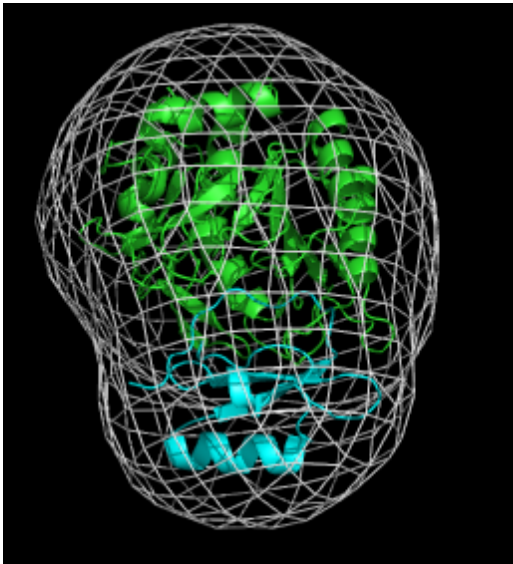
Not for small systems
Resolution often $> 4 \text{ \AA}$



Shape data => Fitting in 3D map

SAXS, Cryo-EM ...

Cryo-electron microscopy generates electron density maps.
The resolution varies a lot!

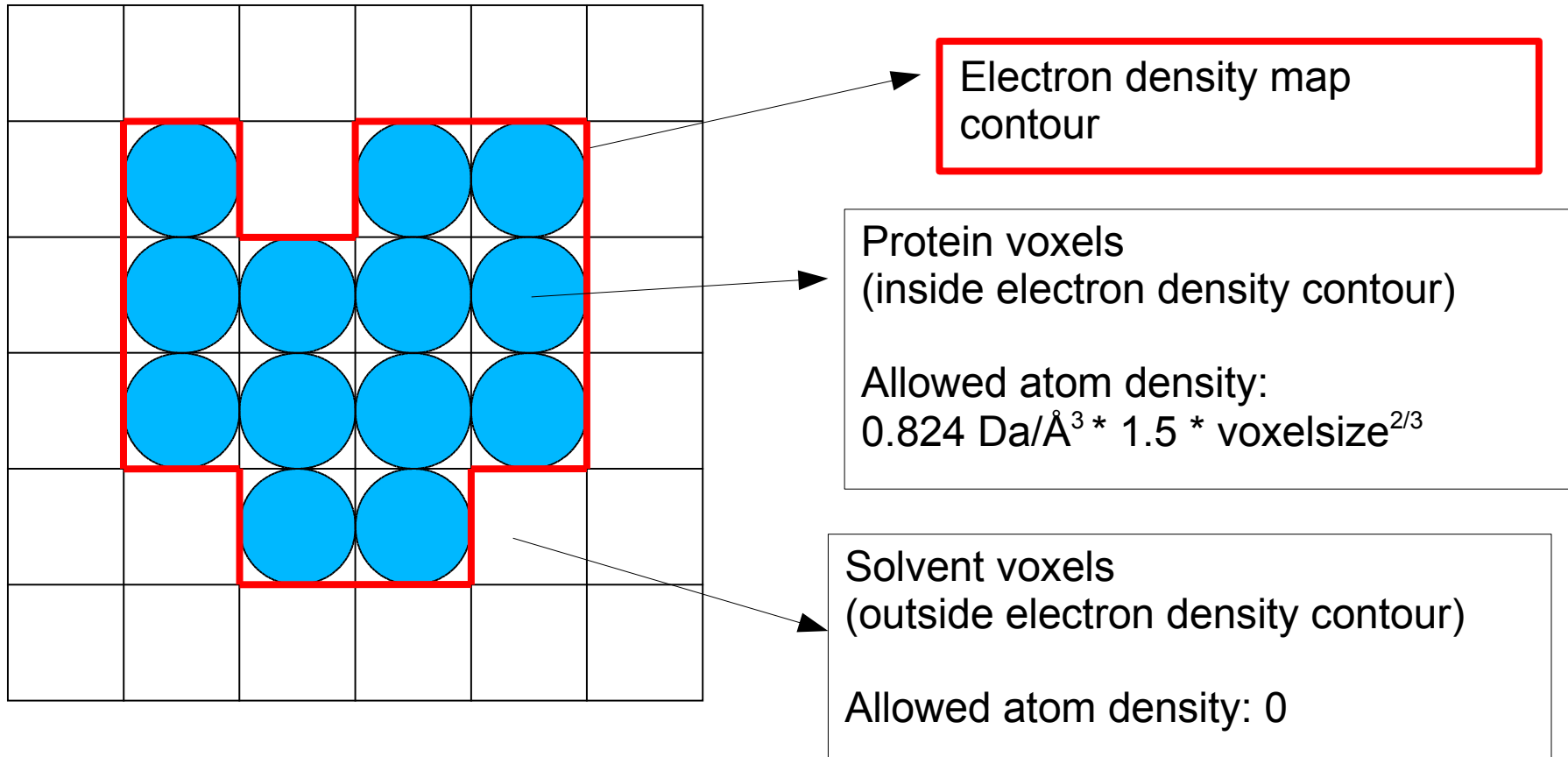


=> Fitting in 3D map

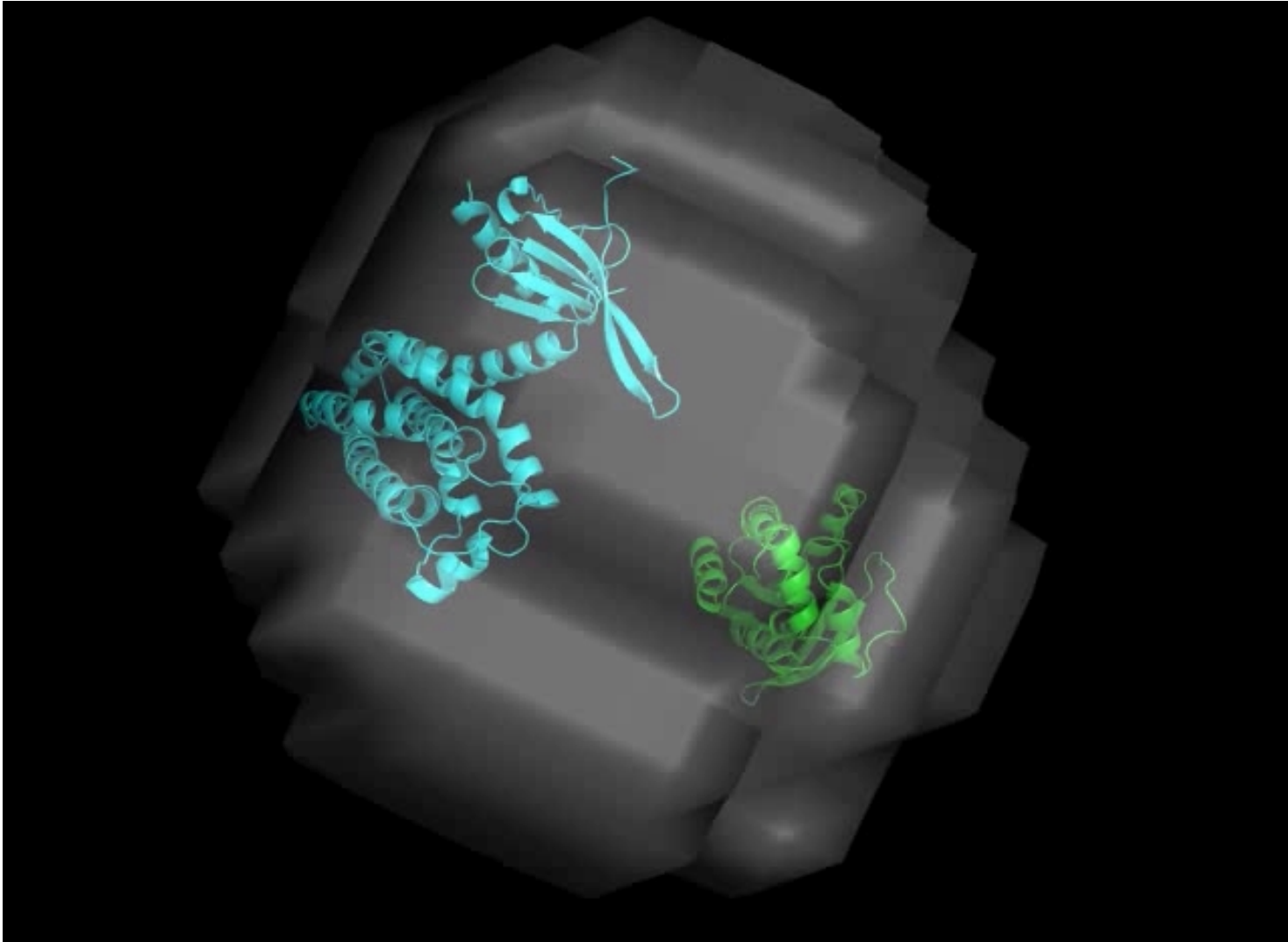
Shape data => Fitting in 3D map

SAXS, Cryo-EM ...

Sampling: Voxel-based atom density mask



ATTRACT-EM



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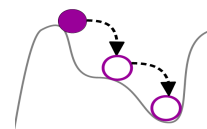
Data-driven docking: Contact/interface -driven
Fitting in 3D shape

Other docking paradigms

Modeling pipeline

Docking paradigms

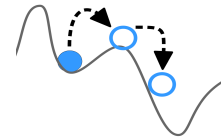
Minimisation-driven sampling



Gradient-descent

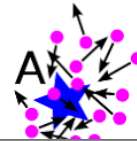
HADDOCK

ATTRACT

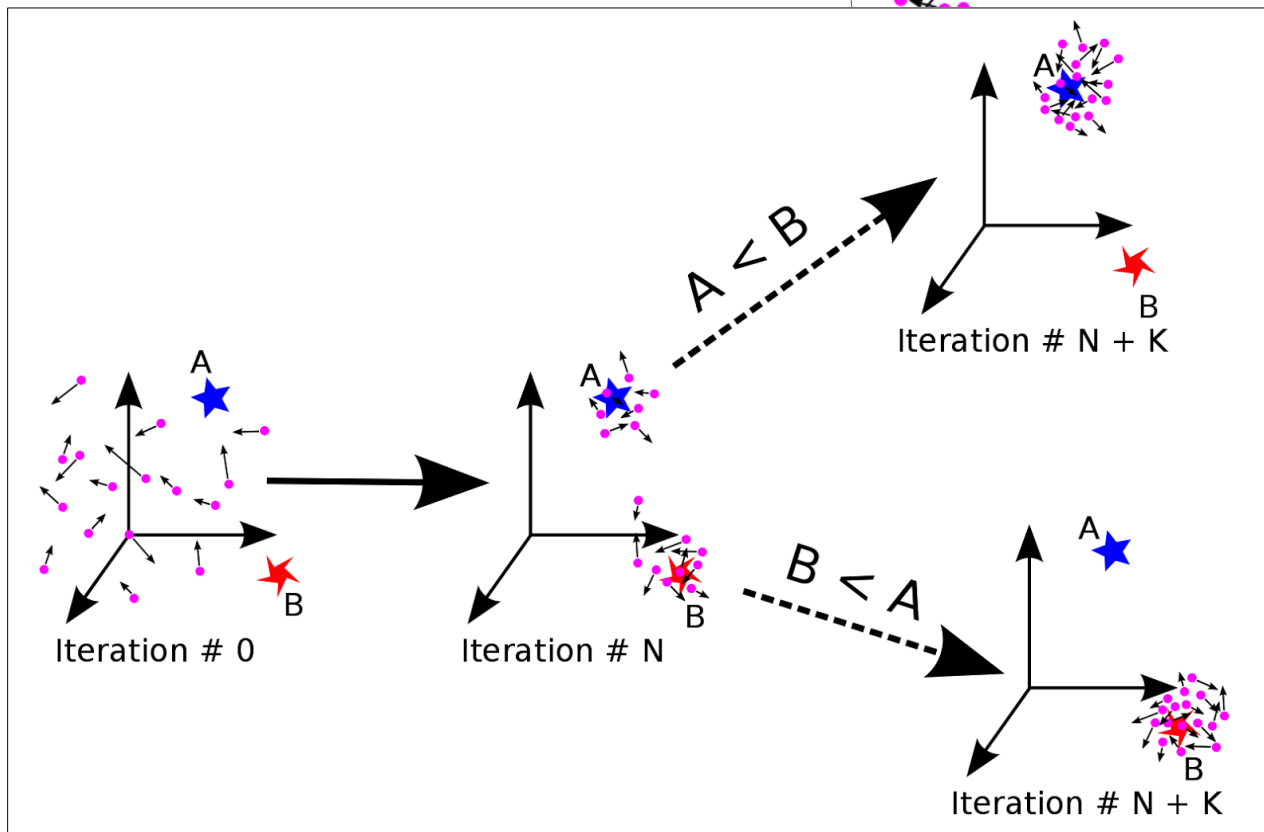


Monte Carlo

RosettaDock



Swarm optimization



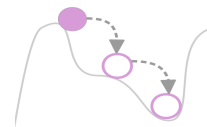
Docking paradigms

Systematic sampling

FFT

ClusPro, ZDOCK, FroDock,
GRAMM, FTDock, DOT ...

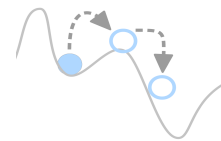
Minimisation-driven sampling



Gradient-descent

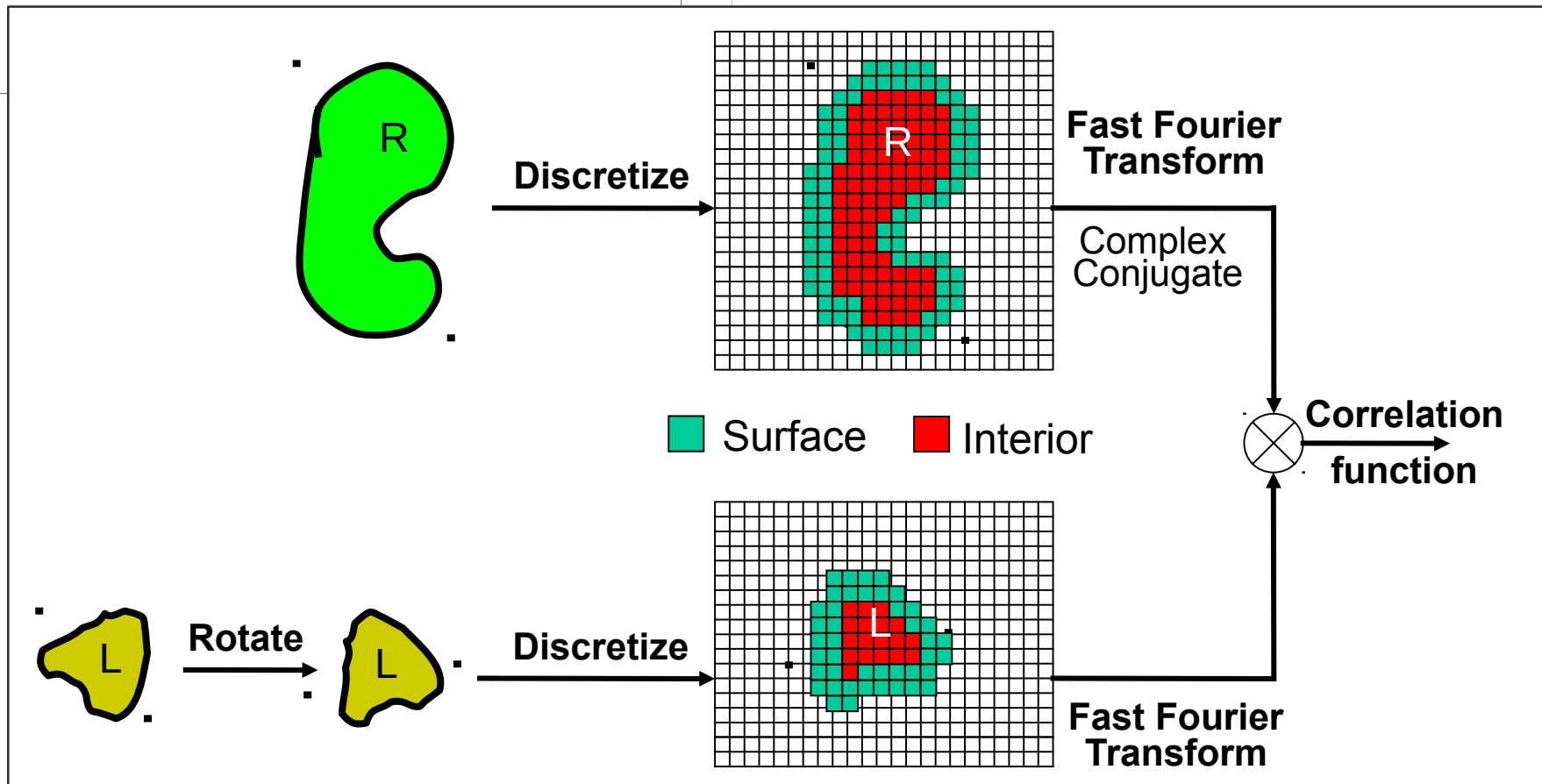
HADDOCK

ATTRACT



Monte Carlo

RosettaDock



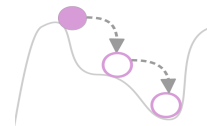
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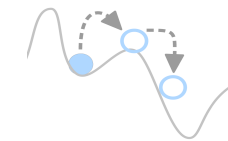
Minimisation-driven sampling



Gradient-descent

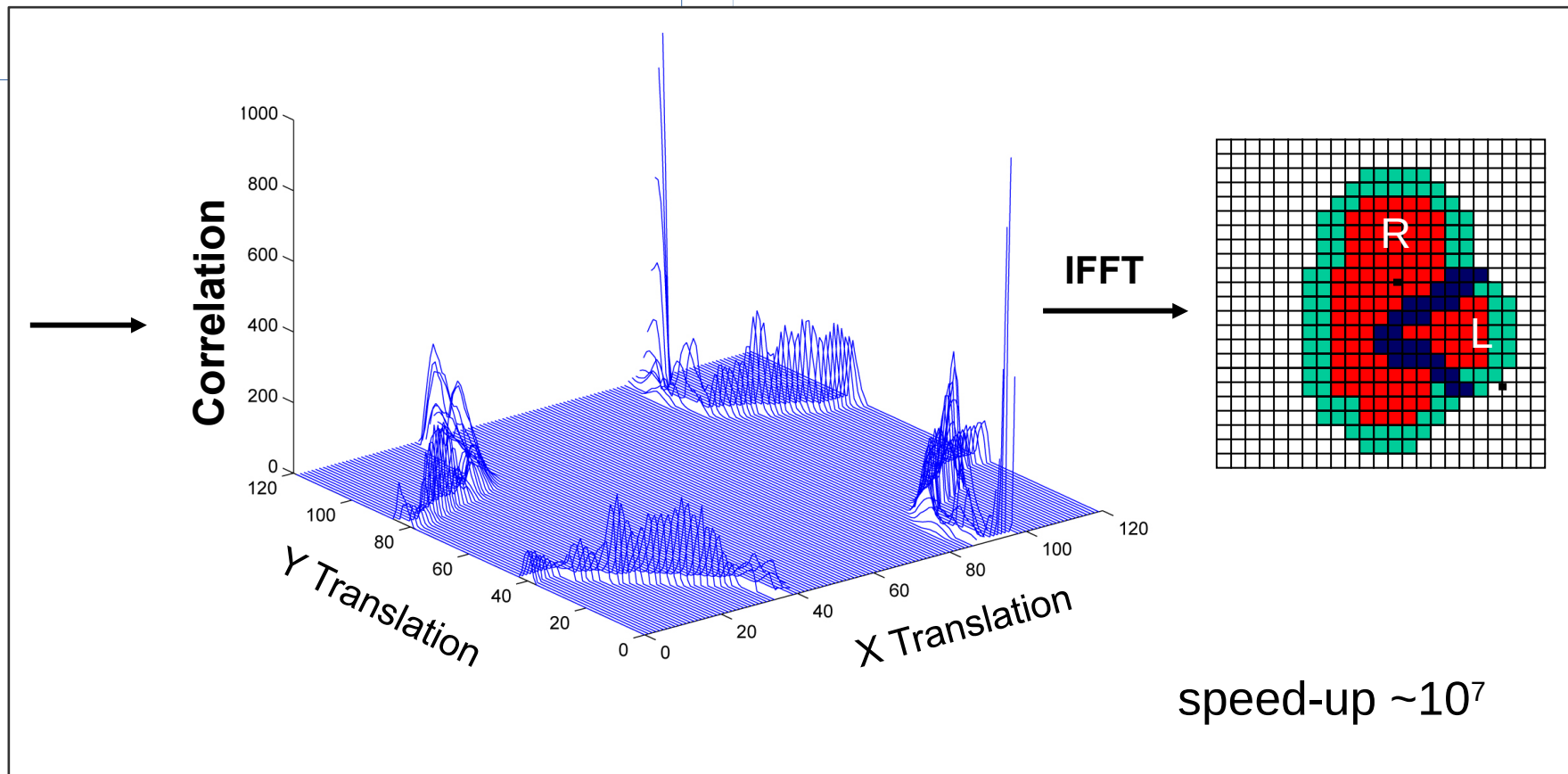
HADDOCK

ATTRACT



Monte Carlo

RosettaDock



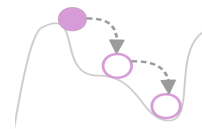
Docking paradigms

Systematic sampling

FFT ClusPro, ZDOCK, FroDock, GRAMM, FTDock, DOT ...

Spherical FFT HEX

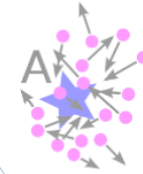
Minimisation-driven sampling



Gradient-descent HADDOCK

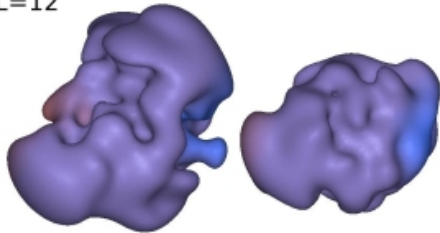


Monte Carlo ATTRACT
RosettaDock

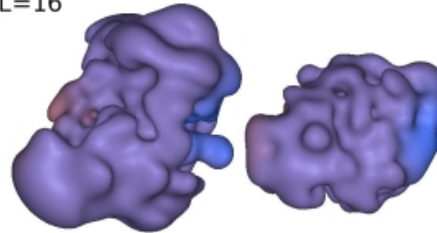


Swarm optimization SwarmDock

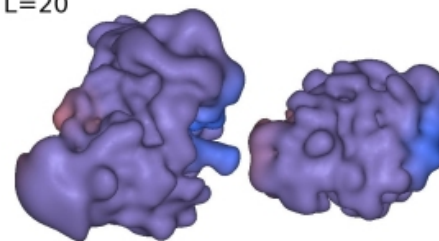
L=12



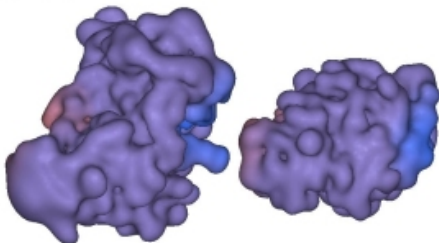
L=16



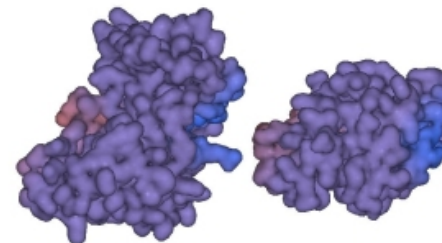
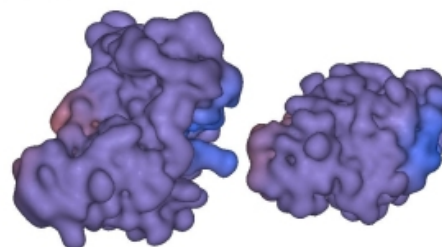
L=20



L=24



L=28

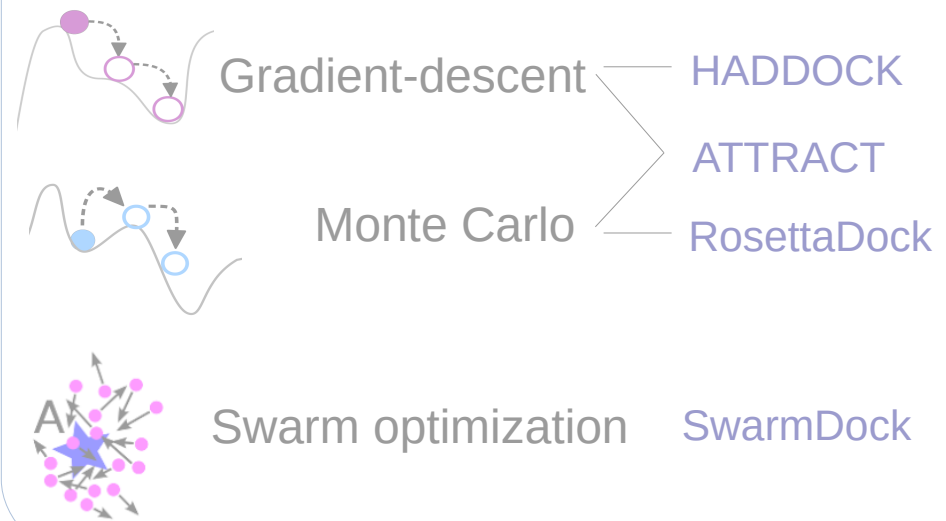


Docking paradigms

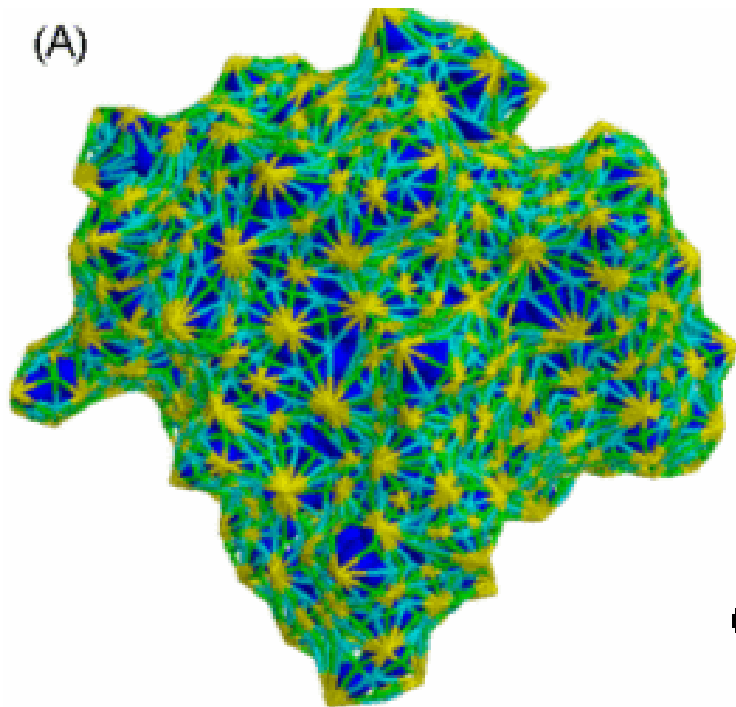
Systematic sampling

FFT	ClusPro, ZDOCK, FroDock, GRAMM, FTDock, DOT ...
Spherical FFT	HEX
Geometric hashing	PatchDock

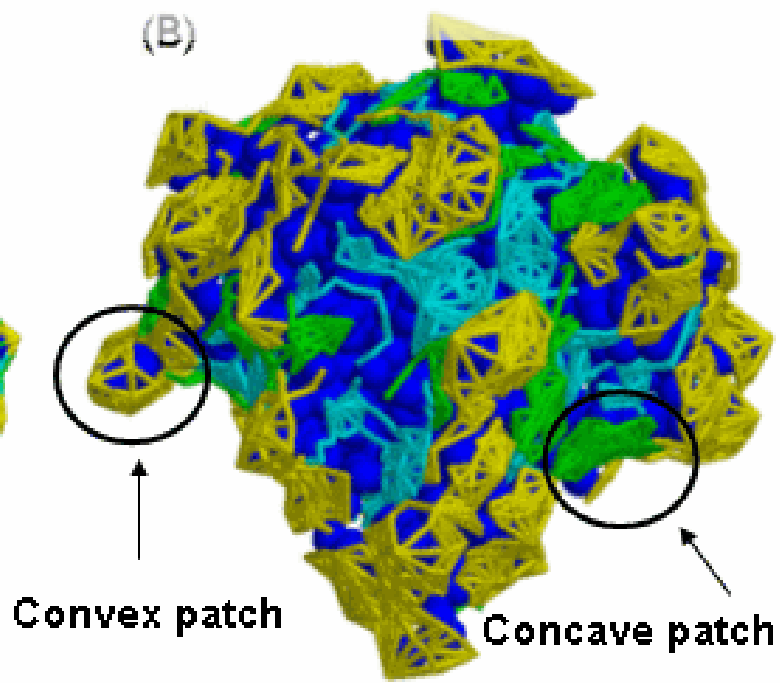
Minimisation-driven sampling



(A)



(B)



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