

Integrative modeling with IMP **SEA complex**

A case study with ambiguity and low structural coverage

Ilan E. Chemmama, Ignacia Echeverria, Ph.D., Seung Joong Kim, Ph.D.

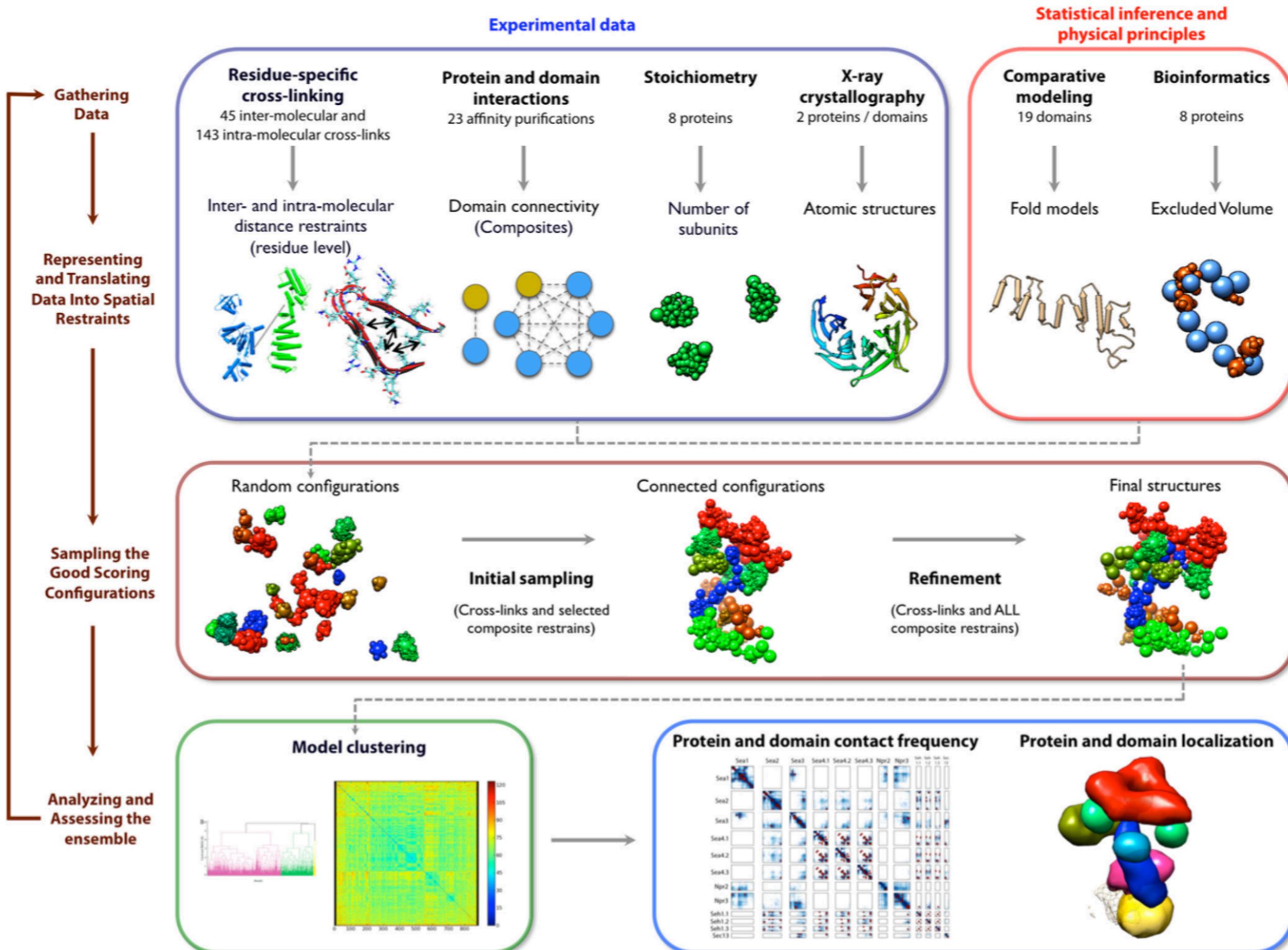
Andrey Šali Laboratory

Workshop on Computational Biophysics

University of California, San Francisco

16 December 2016

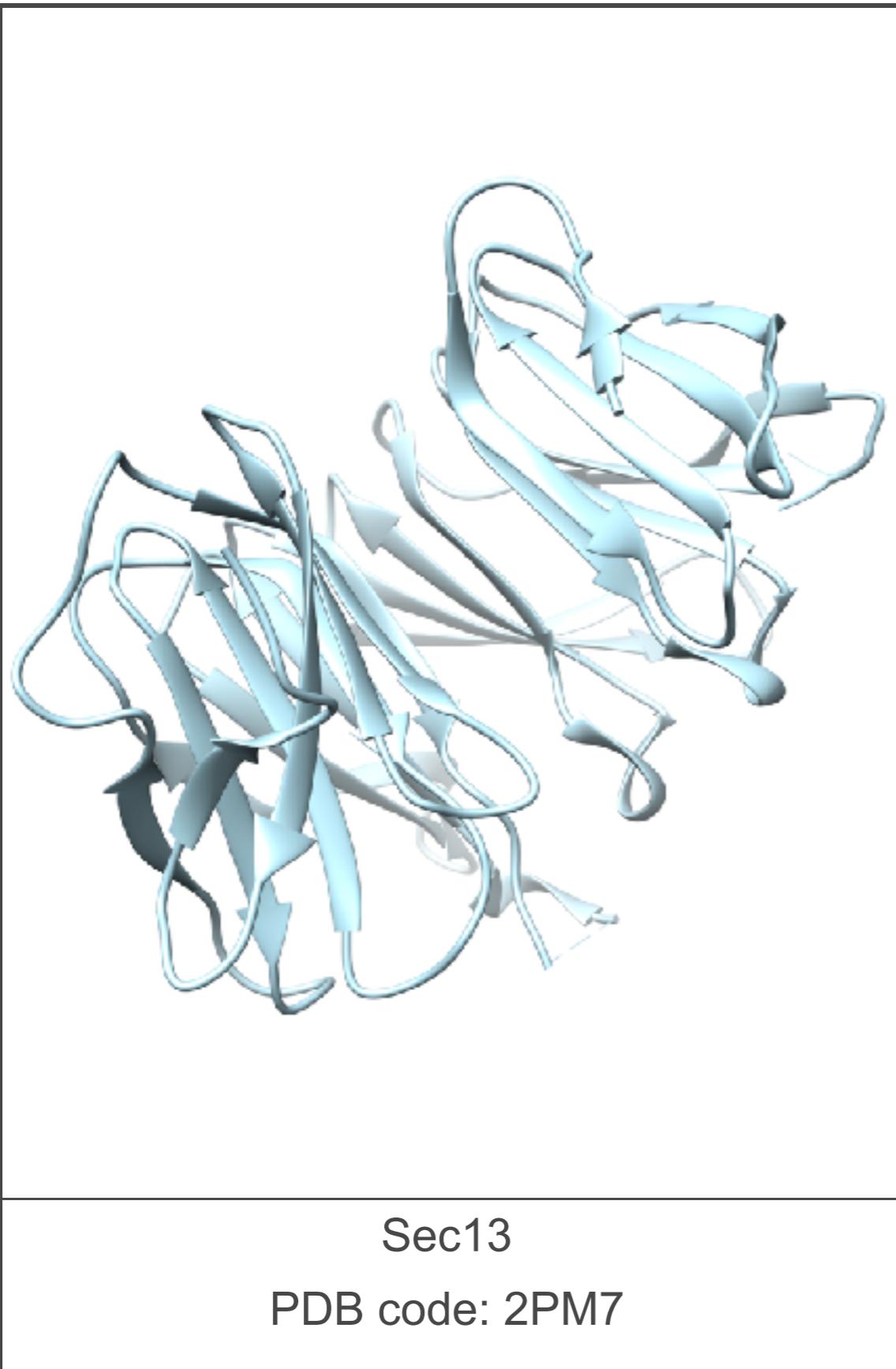
Integrative modeling of the SEA complex



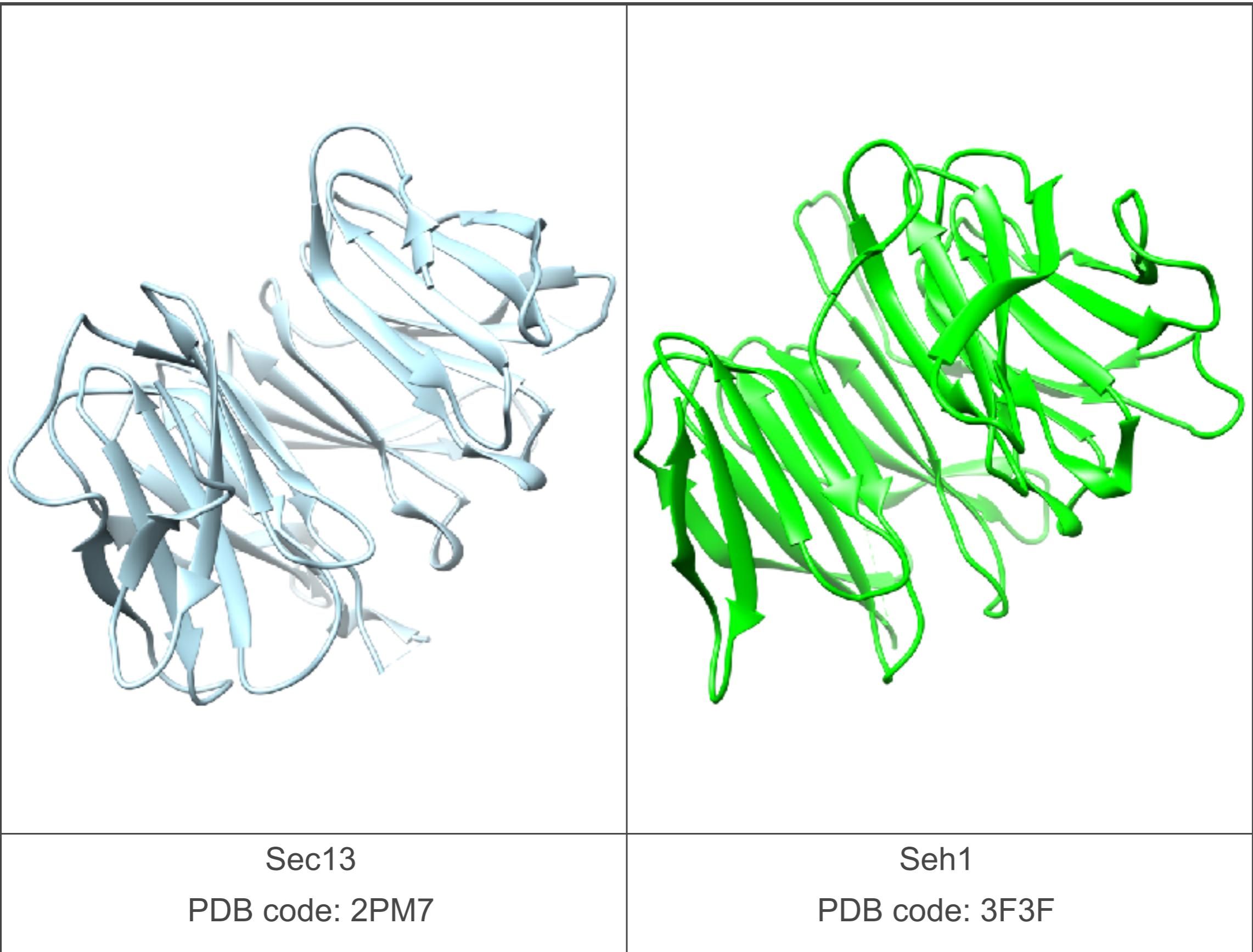
**Gathering Data,
Representing the system
and
Translating them into spatial
restraints**

Prior structural knowledge: crystal structures

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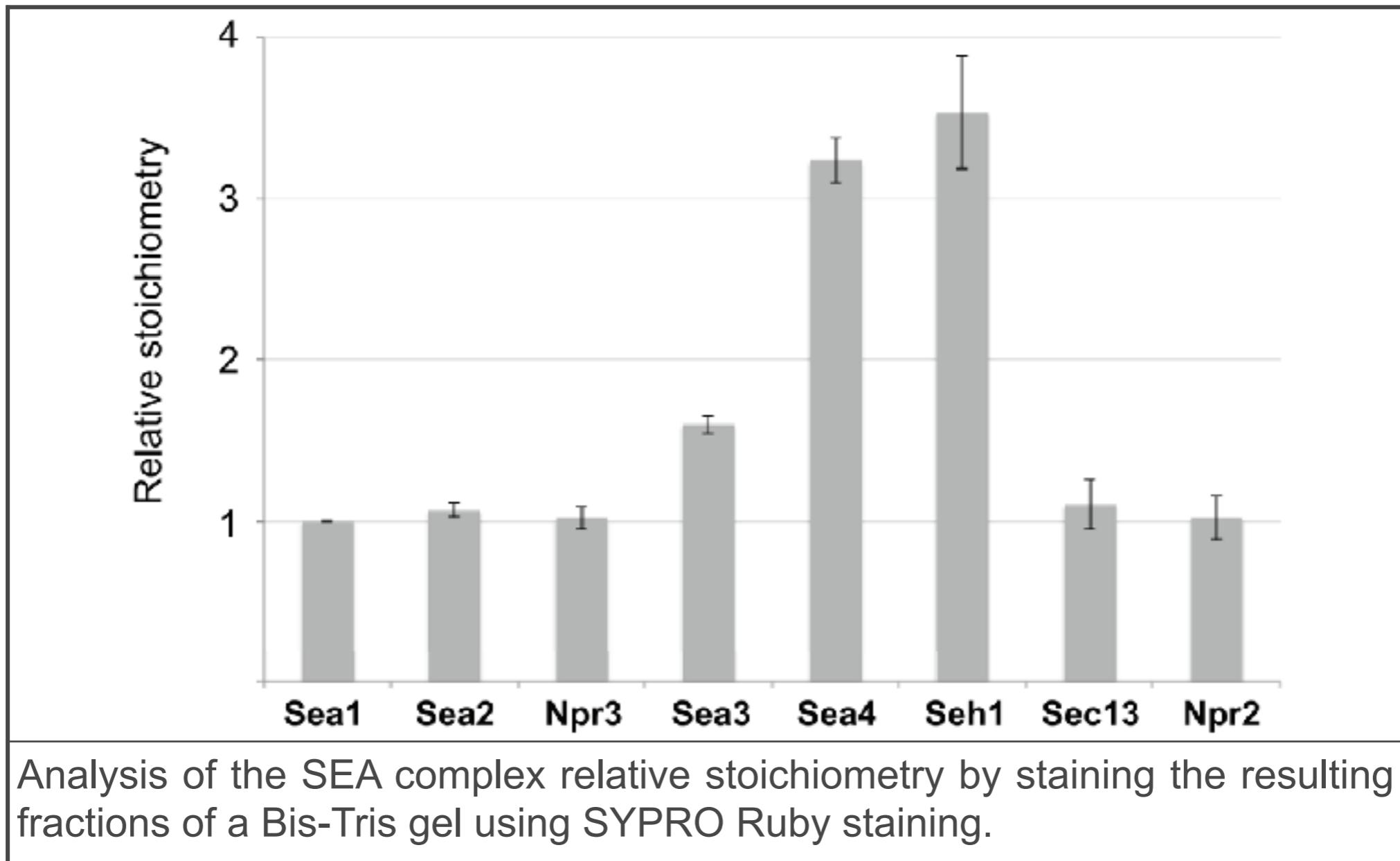


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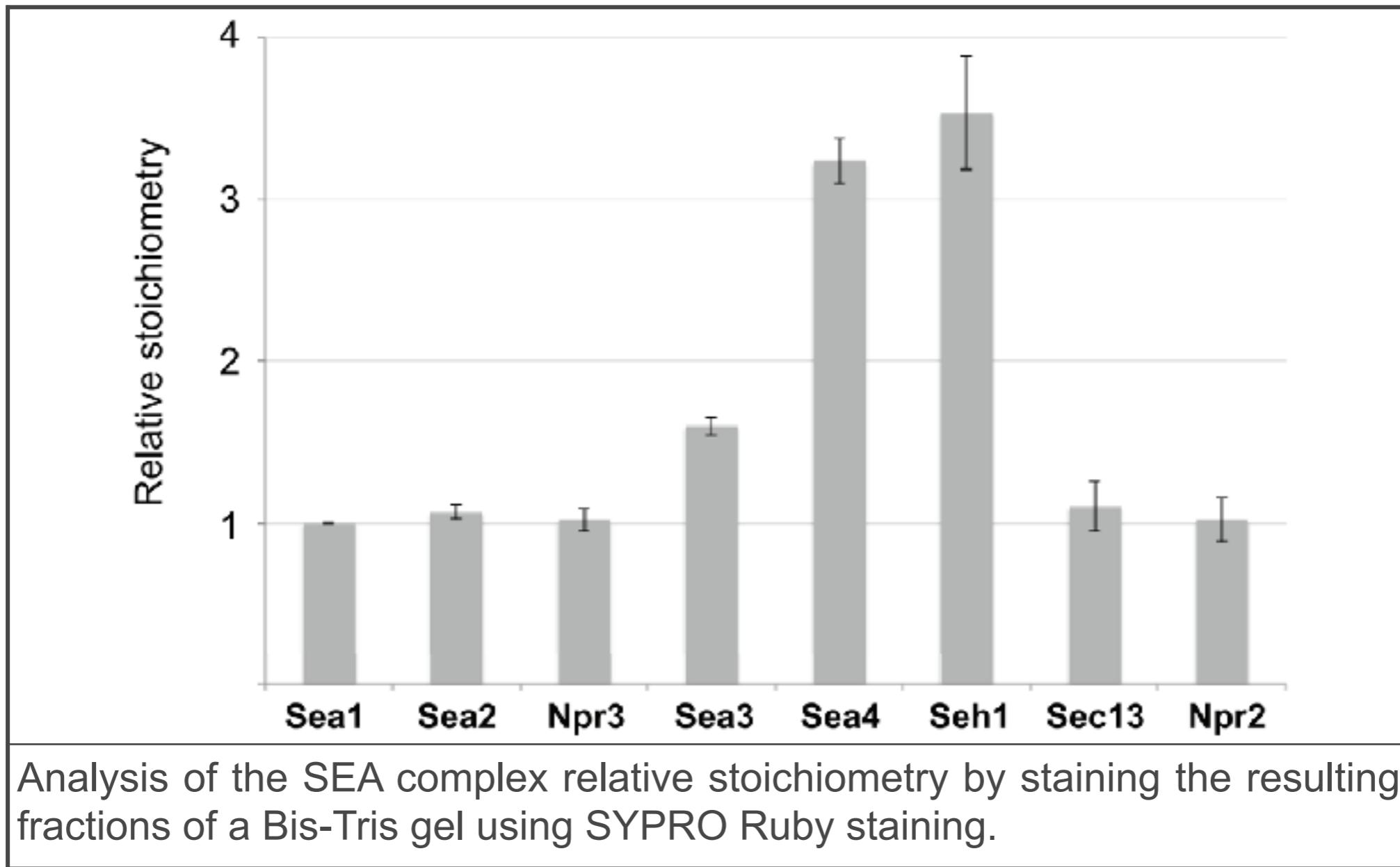


Stoichiometry

Stoichiometry



Stoichiometry



Npr2:Npr3:SEA1:SEA2:SEA3:SEA4:Sec13:Seh1 = 1:1:1:1:1:3:1:3

Representation in IMP 2.6.2:Topology

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```
m = IMP.Model()  
  
topology =  
IMP.pmi.topology.TopologyReader(topology_file)  
domains = topology.component_list
```

Representation for the topology of the biomolecules is build using prior structural knowledge such as stoichiometry, crystal structures (Sec13, Seh1), comparative models built from homologs, secondary structure predictions,....

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component_name	domain_name	fasta_fn	fasta_id	pdb_fn	chain	residue_range	pdb_offset	bead_size	em_residues_per_gaussian
Npr2	Npr2_1	SeaComplex.fasta	Npr2		A	1, 8	0	10	0
Npr2	Npr2_2	SeaComplex.fasta	Npr2	Npr2_9-127.pdb	A	9,127	0	10	0
Npr2	Npr2_3	SeaComplex.fasta	Npr2		A	128,256	0	65	0
Npr2	Npr2_4	SeaComplex.fasta	Npr2	Npr2_257-327.pdb	A	257,327	0	10	0
Npr2	Npr2_501	SeaComplex.fasta	Npr2		A	328,355	0	28	0
Npr2	Npr2_502	SeaComplex.fasta	Npr2		A	356,430	0	75	0
Npr2	Npr2_503	SeaComplex.fasta	Npr2		A	431,493	0	63	0
Npr2	Npr2_504	SeaComplex.fasta	Npr2		A	494,562	0	69	0
Npr2	Npr2_6	SeaComplex.fasta	Npr2	Npr2_563-610.pdb	A	563,610	0	10	0
Npr2	Npr2_7	SeaComplex.fasta	Npr2		A	611,615	0	10	0
Seh1.1	Seh1.1_2	SeaComplex.fasta	Seh1	3F3F.pdb	A	1,248	0	10	0
Seh1.1	Seh1.1_3	SeaComplex.fasta	Seh1		A	249,287	0	39	0
Seh1.1	Seh1.1_4	SeaComplex.fasta	Seh1	3F3F.pdb	A	288,346	0	10	0
Seh1.1	Seh1.1_5	SeaComplex.fasta	Seh1		A	347,349	0	10	0
Seh1.2	Seh1.2_2	SeaComplex.fasta	Seh1	3F3F.pdb	A	1,248	0	10	0
Seh1.2	Seh1.2_3	SeaComplex.fasta	Seh1		A	249,287	0	39	0
Seh1.2	Seh1.2_4	SeaComplex.fasta	Seh1	3F3F.pdb	A	288,346	0	10	0
Seh1.2	Seh1.2_5	SeaComplex.fasta	Seh1		A	347,349	0	10	0
Seh1.3	Seh1.3_2	SeaComplex.fasta	Seh1	3F3F.pdb	A	1,248	0	39	0
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Seh1.3	Seh1.3_4	SeaComplex.fasta	Seh1	3F3F.pdb	A	288,346	0	10	0
Seh1.3	Seh1.3_5	SeaComplex.fasta	Seh1		A	347,349	0	10	0
Sec13	Sec13_1	SeaComplex.fasta	Sec13		D	1,1	0	10	0
Sec13	Sec13_2	SeaComplex.fasta	Sec13	2PM7.pdb	D	2,158	0	10	0
Sec13	Sec13_3	SeaComplex.fasta	Sec13		D	159,165	0	10	0
Sec13	Sec13_4	SeaComplex.fasta	Sec13	2PM7.pdb	D	166,296	0	10	0
Sec13	Sec13_5	SeaComplex.fasta	Sec13		D	297,297	0	10	0

Representation and Movers

Representation and Movers

```
rigid_bodies=[  
    ["Npr2_2"], ["Npr2_4"], ["Npr2_6"],  
    ["Npr3_1"],  
    ["Npr3_3", "Npr3_5"],  
    ["Npr3_7"], ["Npr3_9"], ["Npr3_11"],  
    ["Sec13_2", "Sec13_4"],  
    ["SEA1_2"],  
    ["SEA1_4", "SEA1_6", "SEA1_8"],  
    ["SEA1_10"],  
    ["SEA2_2", "SEA2_4", "SEA2_6", "SEA2_8"],  
    ["SEA2_10"],  
    ["SEA3_2", "SEA3_4", "SEA3_6", "SEA3_8"],  
    ["SEA3_10"], ["SEA3_12"],  
    ["Seh1.1_2", "Seh1.1_4"],  
    ["SEA4.1_2", "SEA4.1_4", "SEA4.1_6", "SEA4.1_8", "SEA4.1_10"],  
    ["SEA4.1_12", "SEA4.1_14"],  
    ["SEA4.1_16", "SEA4.1_18"]]  
]
```

Define the movers by defining the list of rigid body. A rigid body is a set of particle beads that will move together during the sampling stages.

Thus, a set of particle beads within a rigid body will contribute only 6 degrees of freedom to the sampling all together
=> huge speed up in computing time

Representation and Movers

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    ["Sec13_2", "Sec13_4"],  
    ["SEA1_2"],  
    ["SEA1_4", "SEA1_6", "SEA1_8"],  
    ["SEA1_10"],  
    ["SEA2_2", "SEA2_4", "SEA2_6", "SEA2_8"],  
    ["SEA2_10"],  
    ["SEA3_2", "SEA3_4", "SEA3_6", "SEA3_8"],  
    ["SEA3_10"], ["SEA3_12"],  
    ["Seh1.1_2", "Seh1.1_4"],  
    ["SEA4.1_2", "SEA4.1_4", "SEA4.1_6", "SEA4.1_8", "SEA4.1_10"],  
    ["SEA4.1_12", "SEA4.1_14"],  
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]
```

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```
rb_max_trans = 2.00  
rb_max_rot = 0.04  
bead_max_trans = 3.00  
representation.set_rigid_bodies_max_rot(rb_max_rot)  
representation.set_floppy_bodies_max_trans(bead_max_trans)  
representation.set_rigid_bodies_max_trans(rb_max_trans)
```

Define how big the moves are during samples.

Building the system and define aesthetics

Building the system and define aesthetics

```
bm = IMP.pmi.macros.BuildModel(  
    m,  
    component_topologies=domains,  
    list_of_rigid_bodies=rigid_bodies)  
representation = bm.get_representation()
```

Here, we build our representation, needed to start restraining part of our system

Building the system and define aesthetics

```
bm = IMP.pmi.macros.BuildModel(  
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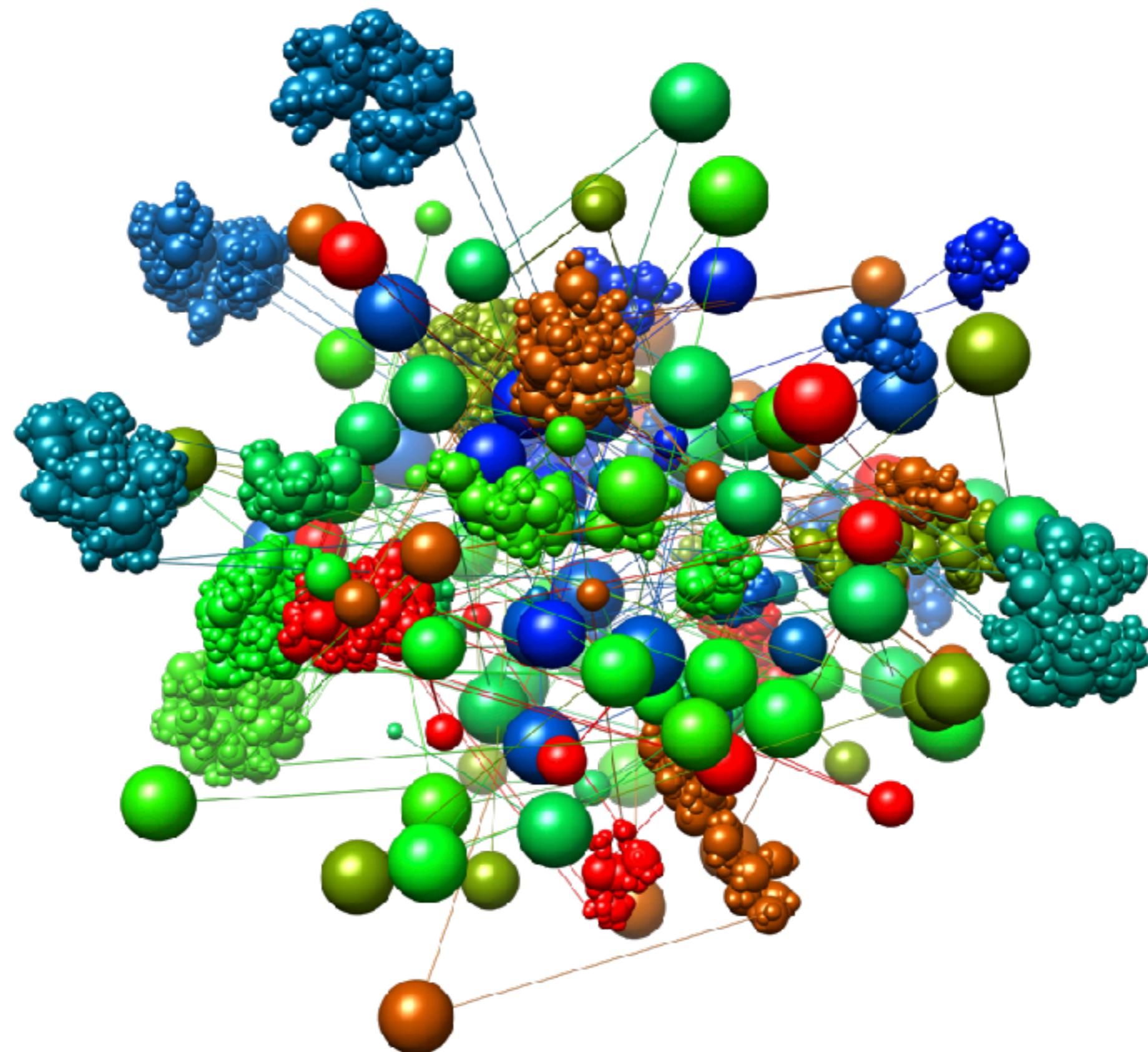
```
for nc,component in enumerate(domains):  
    name = component.name  
    sel =  
    IMP.atom.Selection(representation.prot,molecule=name)  
    ps = sel.get_selected_particles()  
    clr = IMP.display.get_rgb_color(float(nc)/len(domains))  
    for p in ps:  
        if not IMP.display.Colored.get_is_setup(p):  
            IMP.display.Colored.setup_particle(p,clr)  
        else:  
            IMP.display.Colored(p).set_color(clr)
```

Here, we build our representation, needed to start restraining part of our system

A little of aesthetic for the output:
We select each one of the proteins and assign a color (not essential set-up to have good-looking output).

Example of a not-optimized conformation

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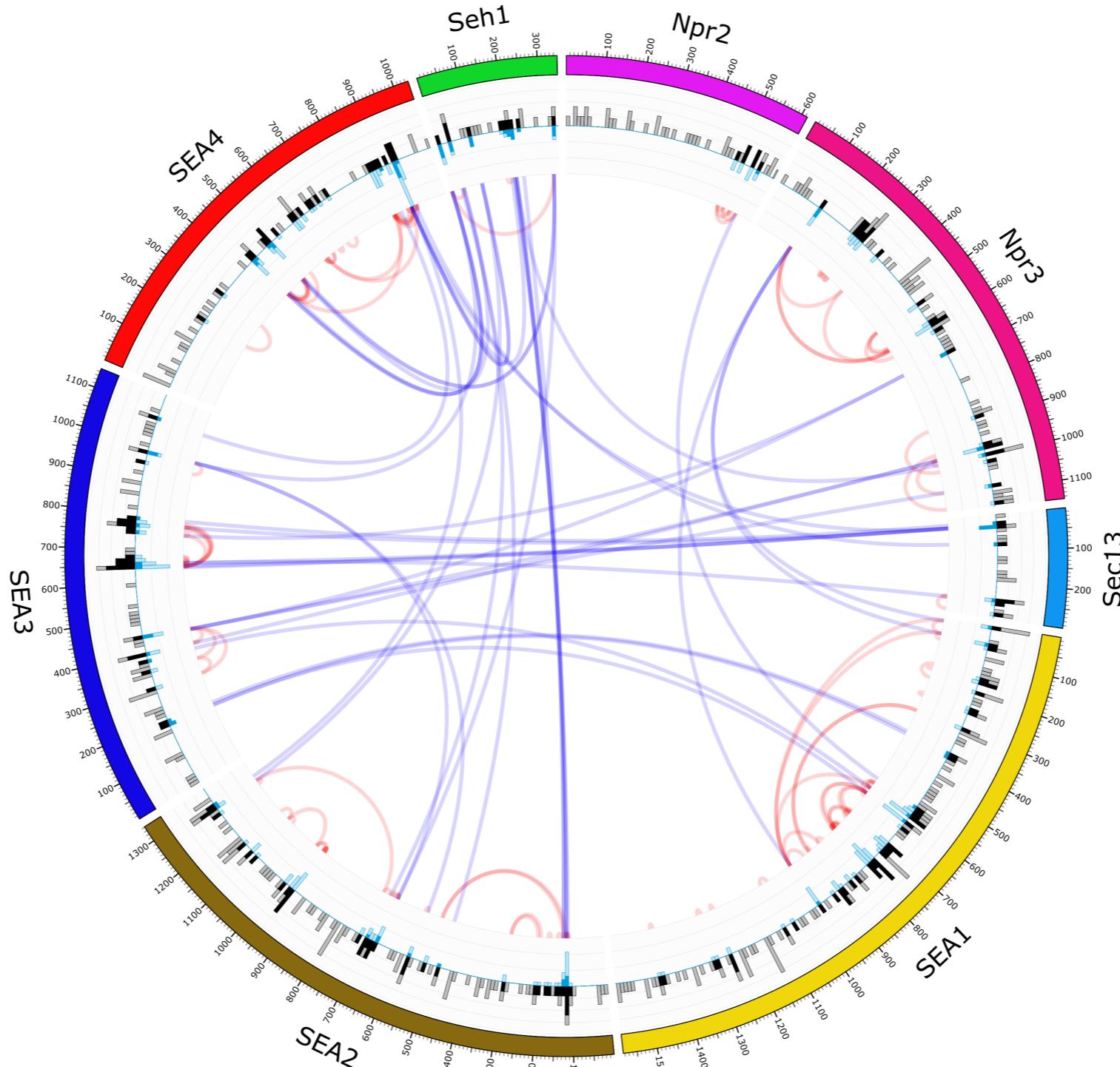


Initial shuffled configuration of the SEA complex.

Example of a physics-based restraint

<pre>ev = IMP.pmi.restraints.stereochemistry.ExcludedVolumeSphere(representation, resolution=10) ev.add_to_model() outputobjects.append(ev)</pre>	Excluded volume on beads with size closest to 10. This speeds up the excluded volume computation rather than to compute it for individual residue bead
---	--

Residue-specific cross-links



Residue-specific cross-links used for the integrative modeling of the SEA complex. 45 inter-xls (**blue**) and 143 intra-xls (**red**) were obtained (188 total).
The **black histogram** represents the density of cross-links in the region.

Encoding the XL restraint

Encoding the XL restraint

prot1	res1	prot2	res2	id	
Npr3	684	SEA3	743	1	<pre>kw = IMP.pmi.io.crosslink.CrossLinkDataBaseKeywordsConverter() kw.set_unique_id_key("id") kw.set_protein1_key("prot1") kw.set_protein2_key("prot2") kw.set_residue1_key("res1") kw.set_residue2_key("res2") kw.set_id_score_key(None)</pre>
SEA1	392	SEA3	189	2	
SEA1	51	Npr3	132	3	
SEA1	562	Npr2	562	4	<pre>xldb = IMP.pmi.io.crosslink.CrossLinkDataBase(kw) xldb.create_set_from_file(datadirectory+'xlinks_SEA.csv')</pre>

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```
xl1 = IMP.pmi.restraints.crosslinking.CrossLinkingMassSpectrometryRestraint(
```

```
representation=representation,
CrossLinkDataBase=xldb,
length=21,
label="SeaComplex",
resolution=1.0,
slope=0.02)
```

This calls the IMP.pmi function for using XLs as spatial restraints using a Bayesian formalism of XL treatment.

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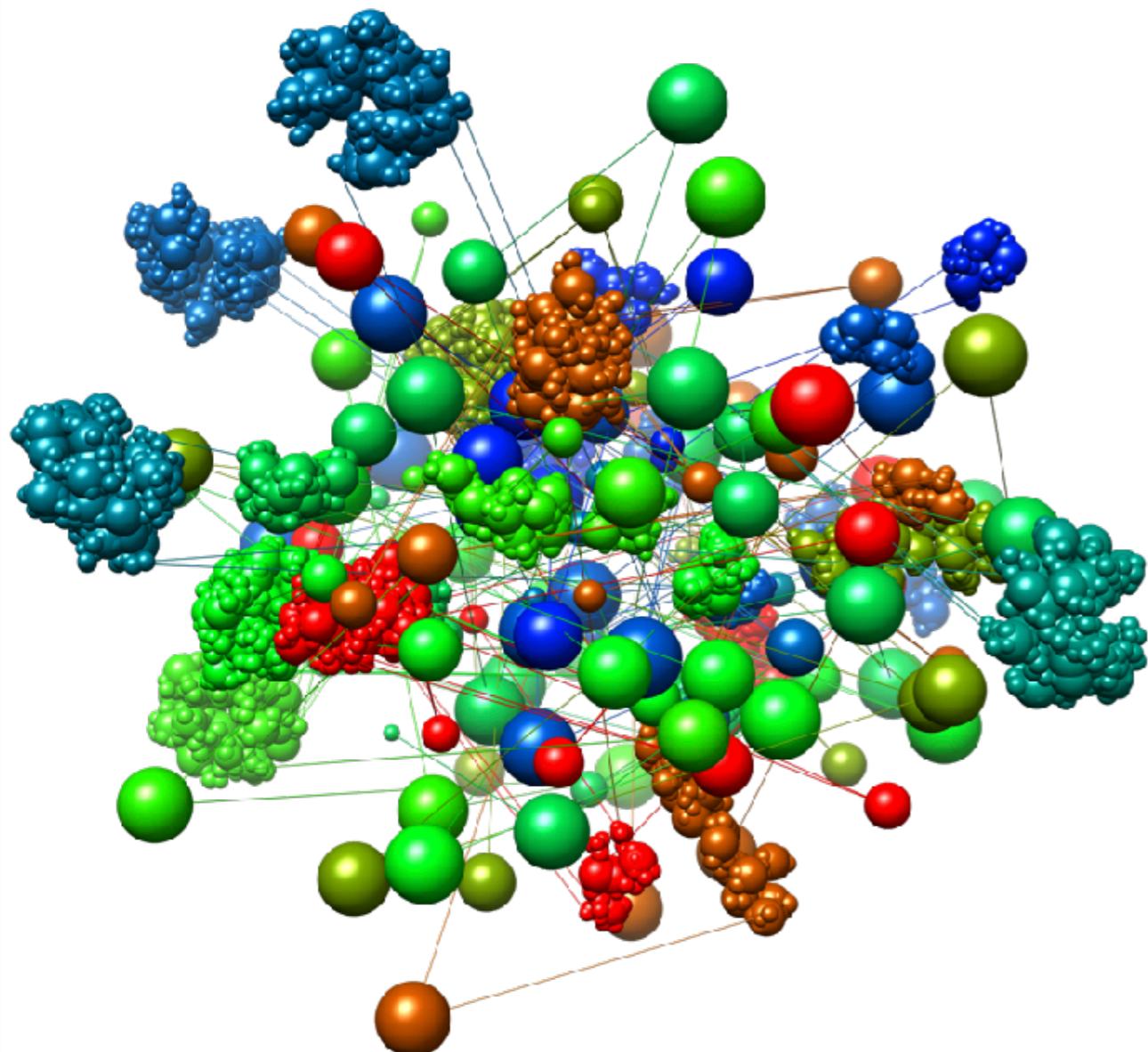
```
xl1.rs.set_weight(25.0)
xl1.add_to_model()
```

```
sampleobjects.append(xl1)
outputobjects.append(xl1)
```

Set the weight of the restraint relative to other restraint and add it to the model

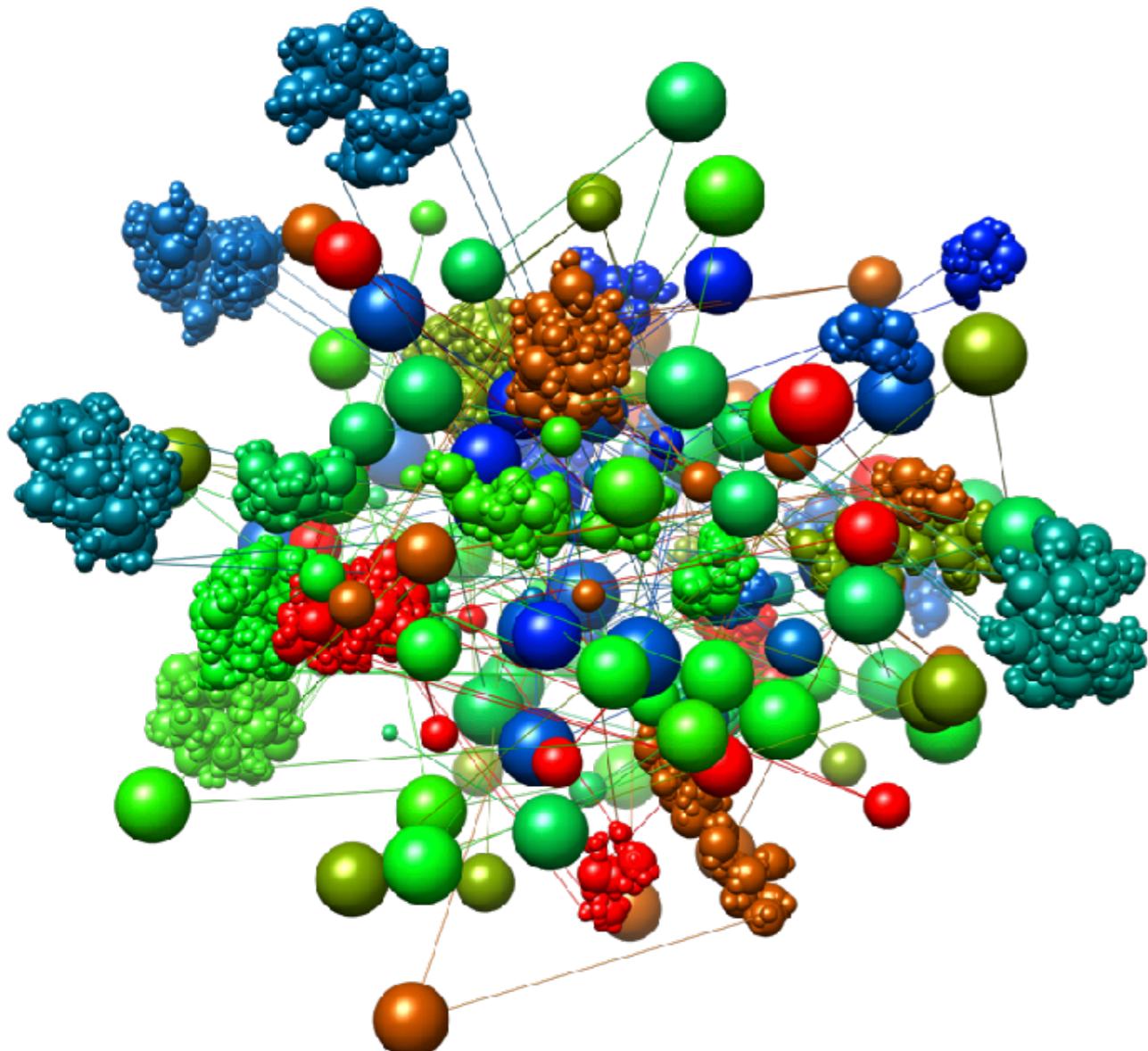
Monte Carlo Gibbs Sampling with Replica Exchange

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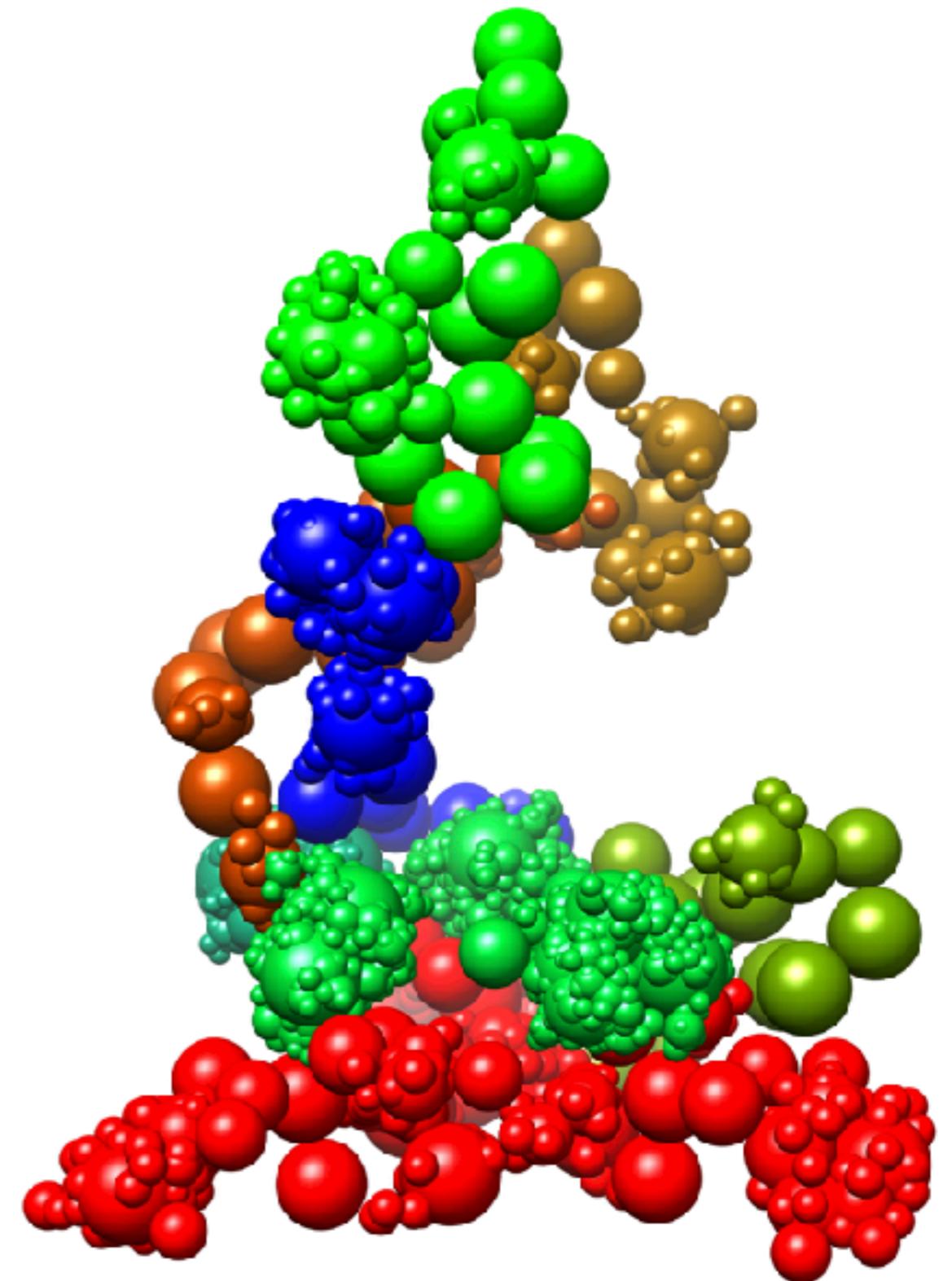


Initial shuffled configuration of the SEA complex.

Monte Carlo Gibbs Sampling with Replica Exchange



Initial shuffled configuration of the SEA complex.



Good-scoring model after sampling

Calling the Sampler!

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```
representation.shuffle_configuration(50)
```

For each of the independent replicate trajectories, we start from a different “shuffled” configuration to ensure uncorrelated, independent trajectories.

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```

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```
mc1=IMP.pmi.macros.ReplicaExchange0(
```

```
    representation,
    monte_carlo_sample_objects=sampleobjects,
    output_objects=outputobjects,
    crosslink_restraints=[x11],
    monte_carlo_temperature=1.0,
    simulated_annealing=True,
    simulated_annealing_minimum_temperature=1.0,
    simulated_annealing_maximum_temperature=2.5,
    simulated_annealing_minimum_temperature_nframes=200,
    simulated_annealing_maximum_temperature_nframes=20,
    replica_exchange_minimum_temperature=1.0,
    replica_exchange_maximum_temperature=2.5,
    number_of_best_scoring_models=1,
    monte_carlo_steps=num_mc_steps,
    number_of_frames=5000,
    global_output_directory="output")
```

Define the sampling parameters for Monte Carlo Gibbs Sampling with Replica Exchange. If run with mpi, automatically setup replica exchange by evenly spaced independent replicas between replica_exchnage_minimum_temperature and replica_exchnage_maximum_temperature for number_of_frames in the directory “./output”

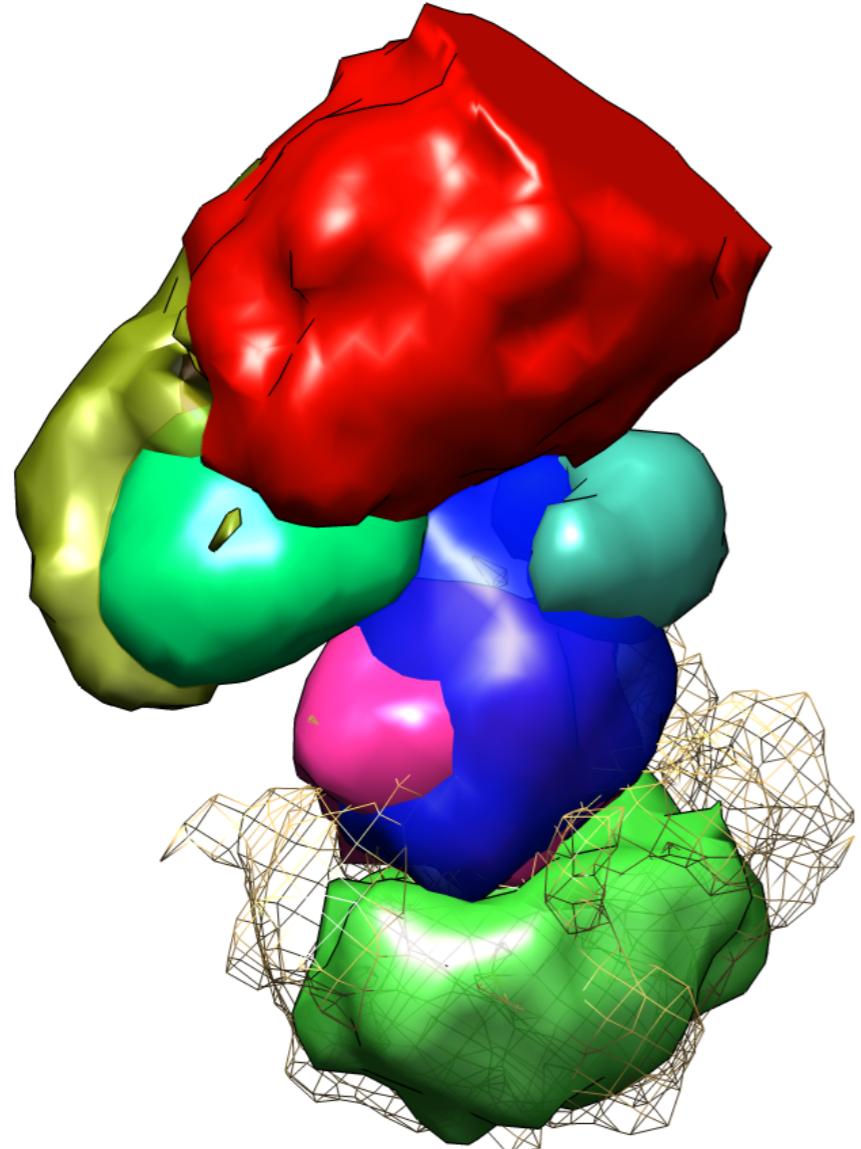
Calling the Sampler!

representation.shuffle_configuration(50)	For each of the independent replicate trajectories, we start from a different “shuffled” configuration to ensure uncorrelated, independent trajectories.
mc1=IMP.pmi.macros.ReplicaExchange0(representation, monte_carlo_sample_objects=sampleobjects, output_objects=outputobjects, crosslink_restraints=[xl1], monte_carlo_temperature=1.0, simulated_annealing=True, simulated_annealing_minimum_temperature=1.0, simulated_annealing_maximum_temperature=2.5, simulated_annealing_minimum_temperature_nframes=200, simulated_annealing_maximum_temperature_nframes=20, replica_exchange_minimum_temperature=1.0, replica_exchange_maximum_temperature=2.5, number_of_best_scoring_models=1, monte_carlo_steps=num_mc_steps, number_of_frames=5000, global_output_directory="output")	Define the sampling parameters for Monte Carlo Gibbs Sampling with Replica Exchange. If run with mpi, automatically setup replica exchange by evenly spaced independent replicas between replica_exchnage_minimum_temperature and replica_exchnage_maximum_temperature for number_of_frames in the directory “./output”
mc1.execute_macro()	Start the sampling by executing the sampler.

Before running a few frames... Spoilers

Adding symmetry

Adding symmetry

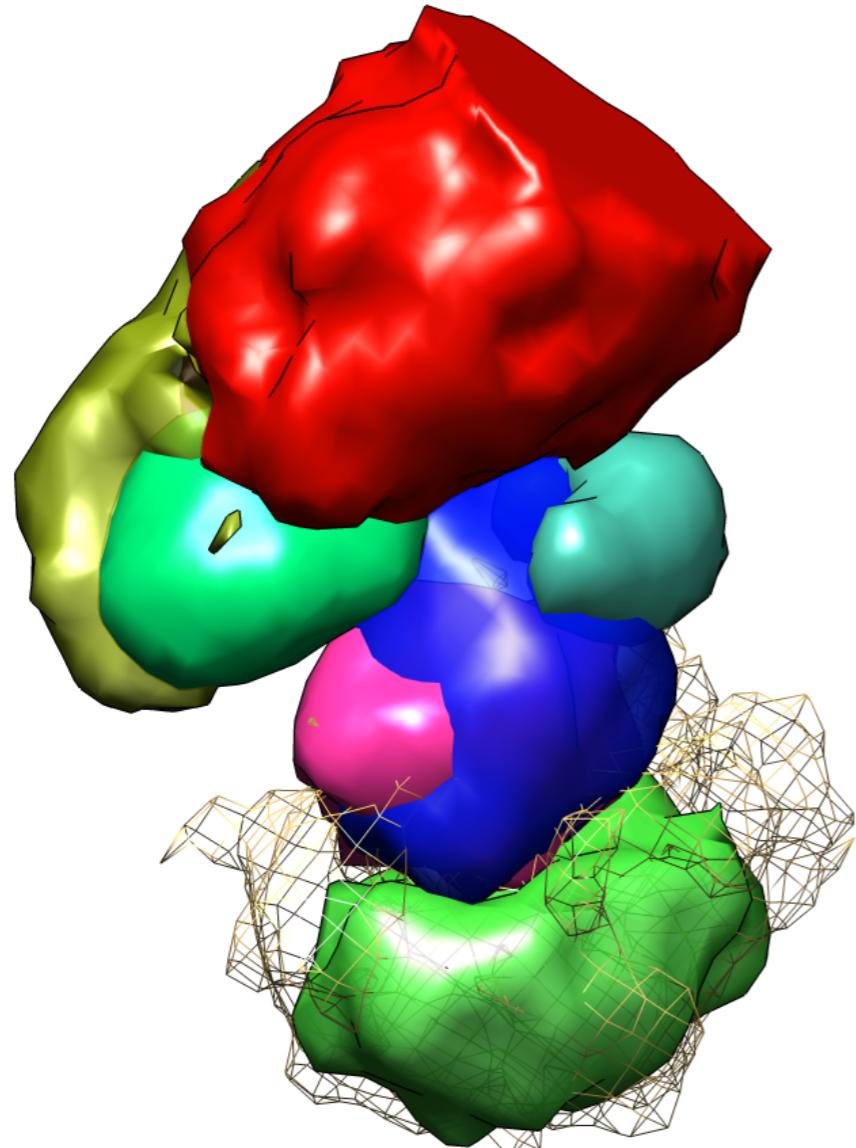


Resulting localization probability density for 1:3 stoichiometry but without symmetry.

Emergence of 3-fold symmetry (C_3) for SEA4 and Seh1.

Constraint the system to have symmetry.

Adding symmetry



Resulting localization probability density for 1:3 stoichiometry but without symmetry.

Emergence of 3-fold symmetry (C3) for SEA4 and Seh1.

Constraint the system to have symmetry.

```
create_rotational_symmetry2(representation, "SEA4.1", ["SEA4.2", "SEA4.3"])
create_rotational_symmetry2(representation, "Seh1.1", ["Seh1.2", "Seh1.3"])
```

Set the weight of the restraint relative to other restraint and add it to the model

Encoding a symmetry constraint in IMP

```
def create_rotational_symmetry2(rps,
                                 maincopy,
                                 copies,
                                 rotational_axis=IMP.algebra.Vector3D(0, 0, 1.0),
                                 nSymmetry=None,
                                 skip_gaussian_in_clones=False):

    from math import pi
    rps.representation_is_modified = True
    ncopies = len(copies) + 1
    main_hiers = IMP.atom.get_leaves(rps.hier_dict[maincopy])

    for k in range(len(copies)):
        if (nSymmetry is None):
            rotation_angle = 2.0 * pi / float(ncopies) * float(k + 1)
        else:
            if (k % 2 == 0):
                rotation_angle = 2.0 * pi / float(nSymmetry) * float((k + 2) / 2)
            else:
                rotation_angle = -2.0 * pi / float(nSymmetry) * float((k + 1) / 2)
        rotation3D = IMP.algebra.get_rotation_about_axis(rotational_axis, rotation_angle)
        sm = IMP.core.TransformationSymmetry(rotation3D)
        clone_hiers = IMP.atom.get_leaves(rps.hier_dict[copies[k]])

        lc = IMP.container.ListSingletonContainer(rps.m)
        for n, p in enumerate(main_hiers):
            if (skip_gaussian_in_clones):
                if (IMP.core.Gaussian.get_is_setup(p)) \
                   and not (IMP.atom.Fragment.get_is_setup(p) or \
                             IMP.atom.Residue.get_is_setup(p)):
                    continue
            pc = clone_hiers[n]
            IMP.core.Reference.setup_particle(pc.get_particle(), p.get_particle())
            lc.add(pc.get_particle().get_index())

        c = IMP.container.SingletonsConstraint(sm, None, lc)
        rps.m.add_score_state(c)
        print("Completed setting " + str(maincopy) + \
              " as a reference for " + str(copies[k]) + \
              " by rotating it in " + str(rotation_angle / 2.0 / pi * 360) + \
              " degree around the " + str(rotational_axis) + " axis.")

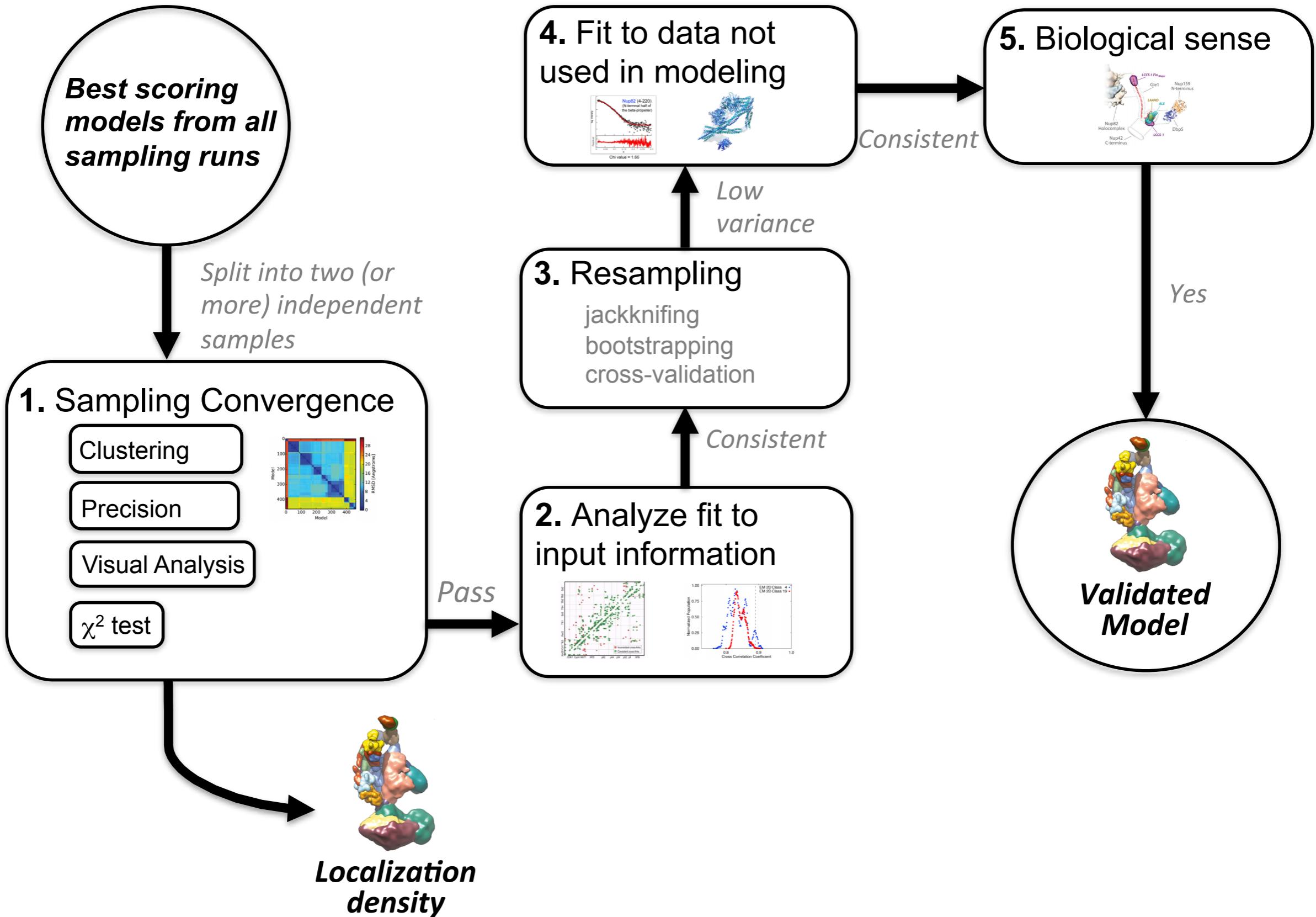
    rps.m.update()
```

Create a symmetry constraint, given the main copy and its clone. The clones will have congruent configuration, transformed by **rotation_angle** and **rotational_axis**.

Let it run, let it run, let it run...

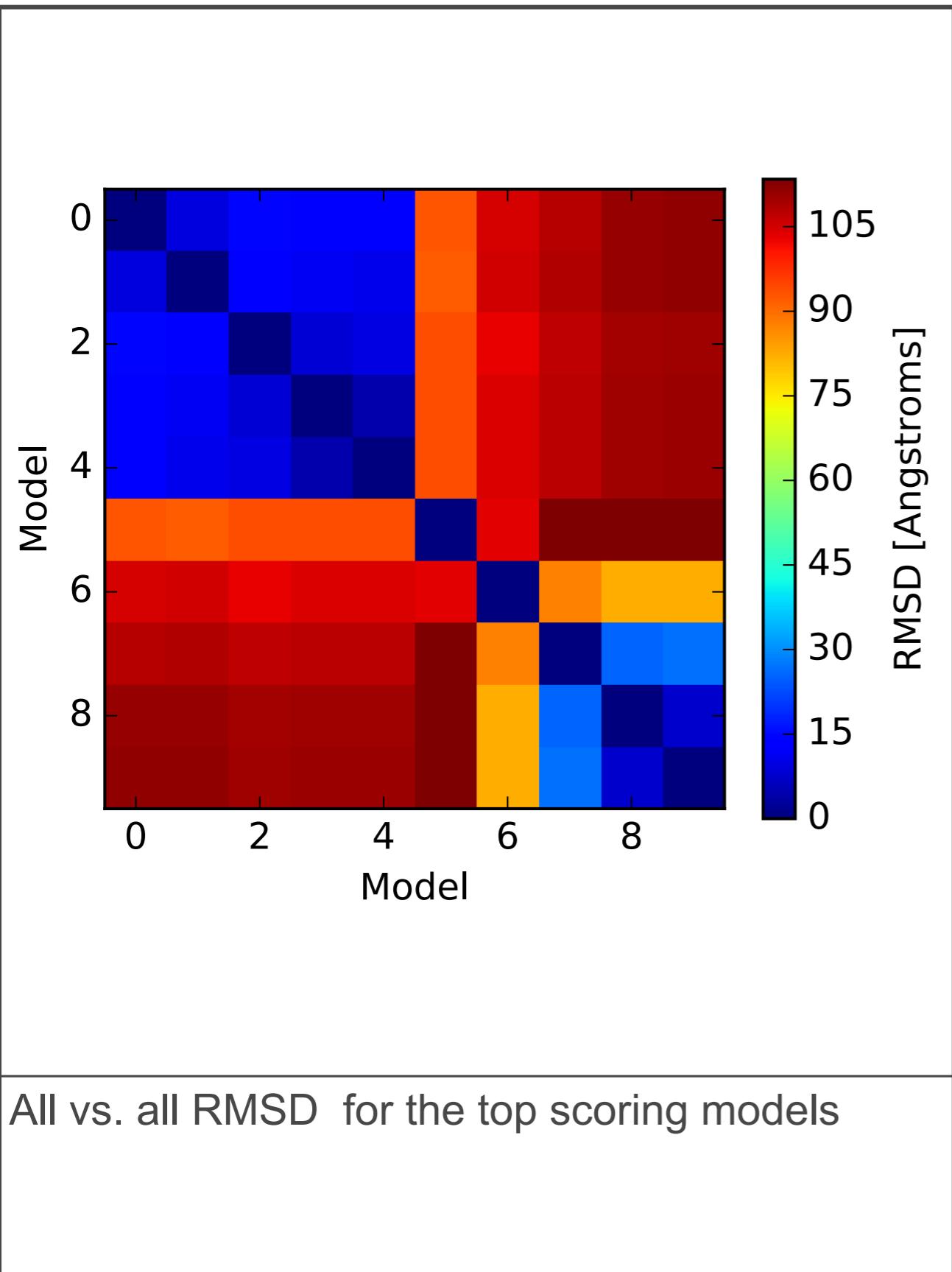
Two examples trajectories are provided as examples.

Analyzing the results

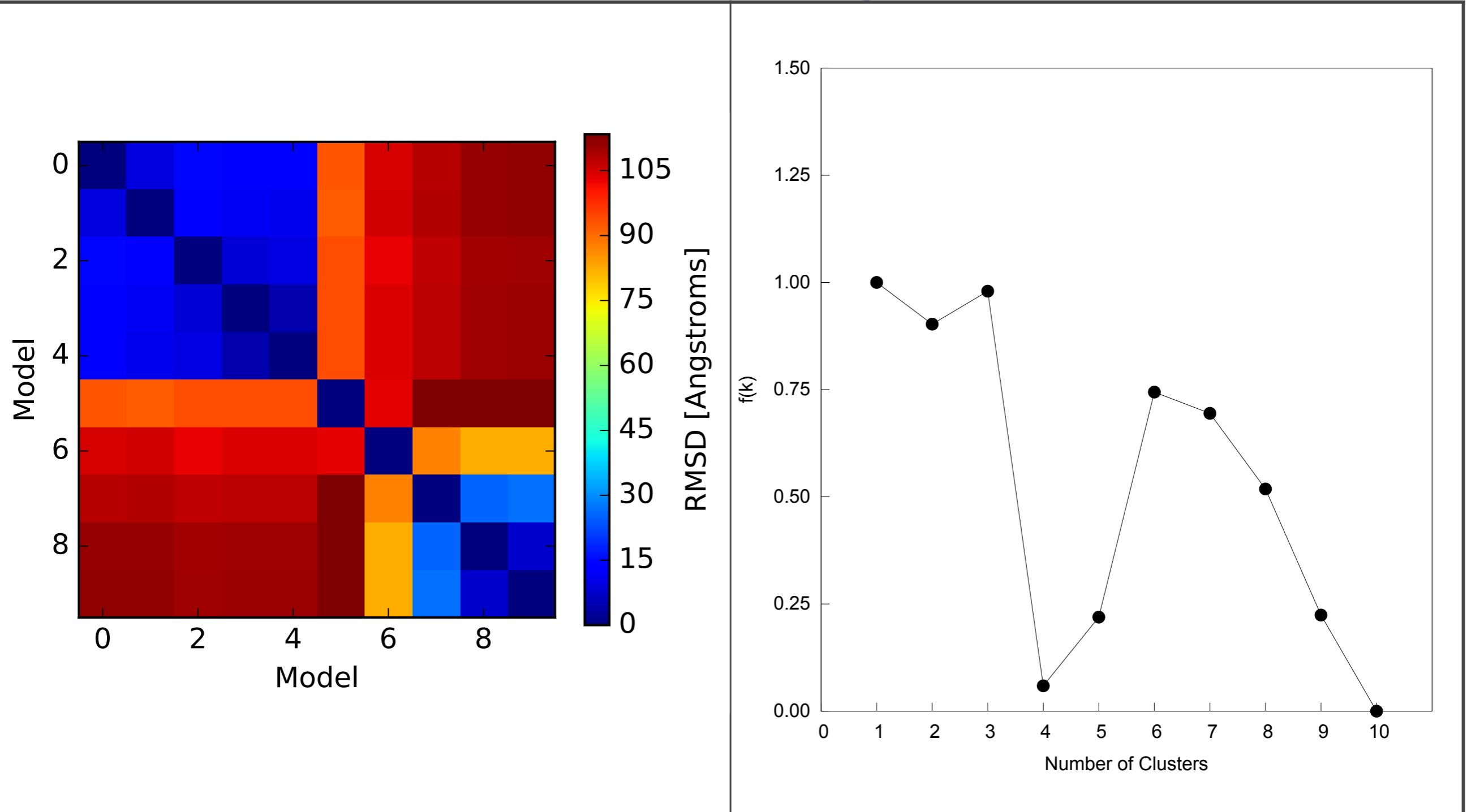


Clustering

Clustering



Clustering



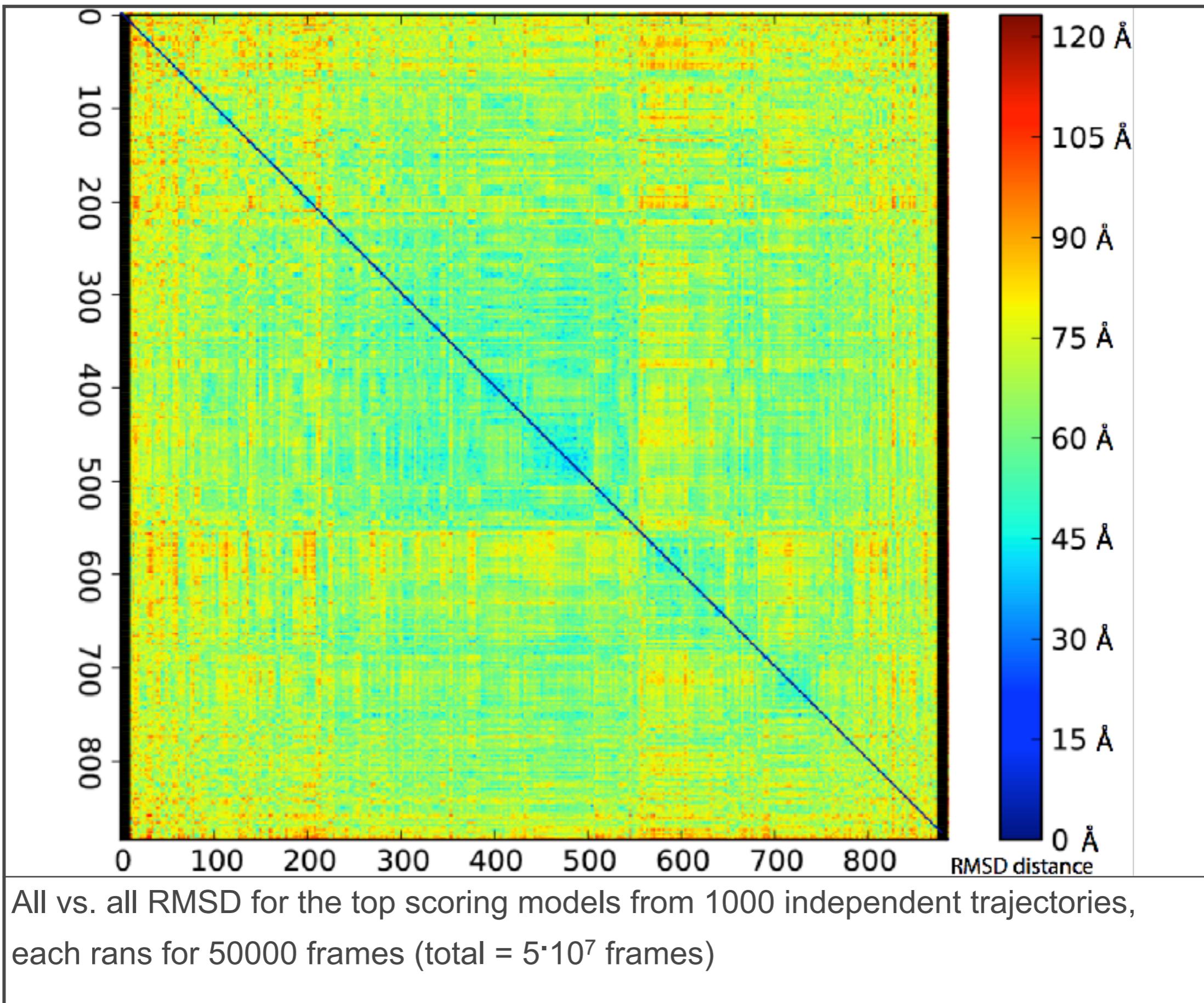
All vs. all RMSD for the top scoring models

Determination of the optimal number of clusters using a distortion metric. Here, the extremum is four.

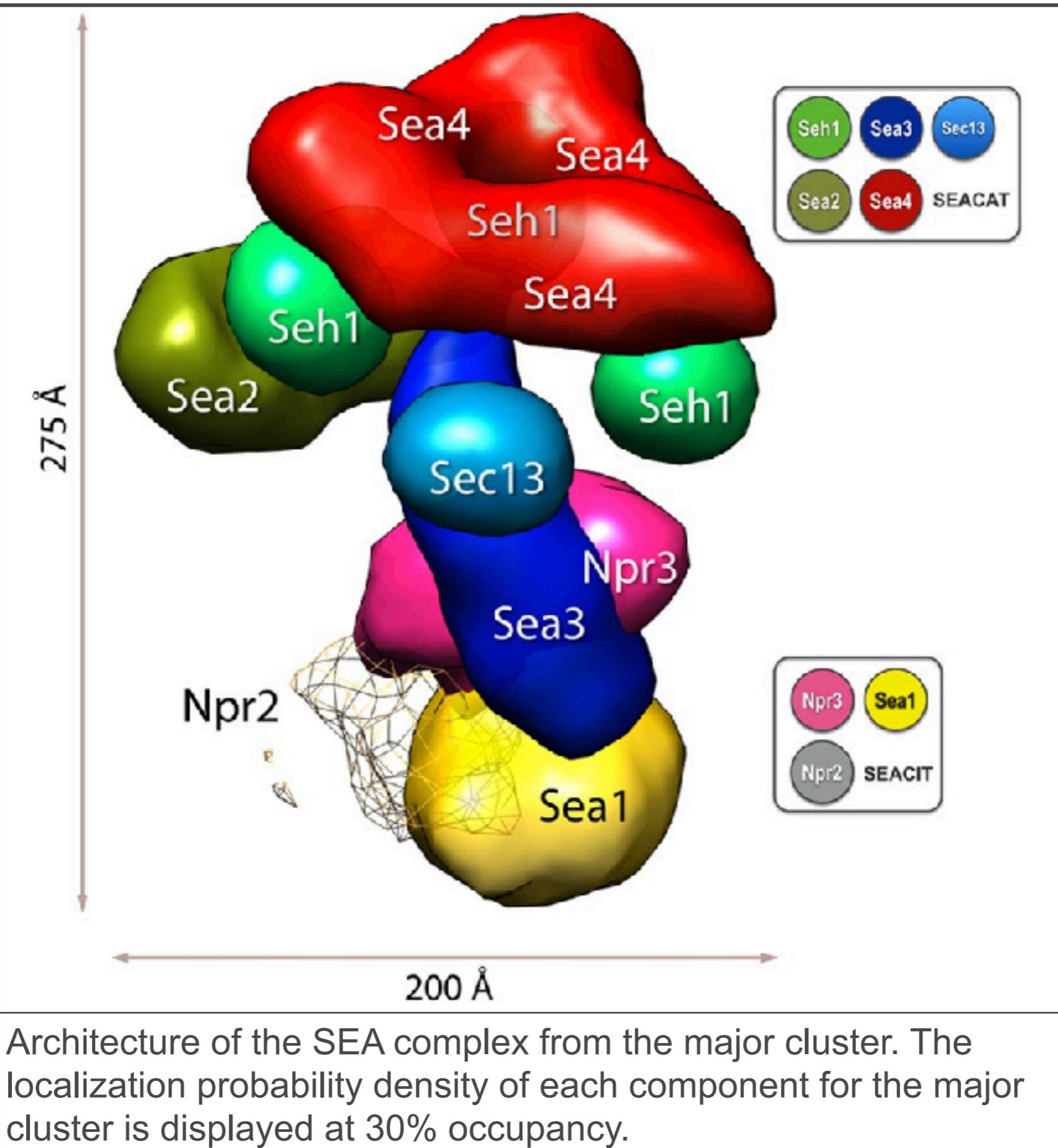
χ^2 -contingency test for homogeneity of population

Sample 0	Sample 1	
0	100	For such a small number of frames, obviously, the test fails. None of the cluster populations displays homogeneity, i.e. the two runs did not contribute equally to each clusters
60	0	
20	0	Sampling has not converged
20	0	

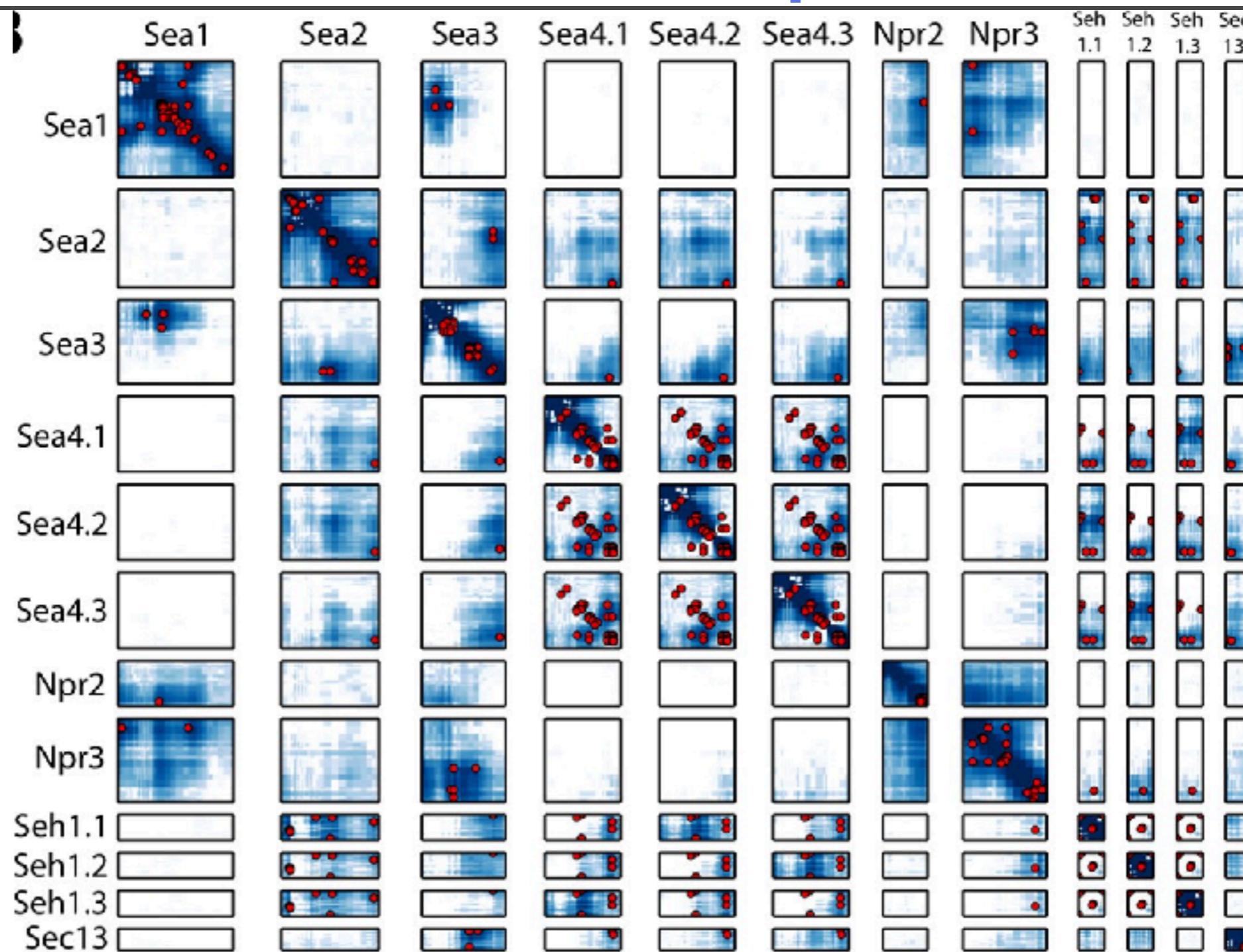
Clustering for serious sampling



Probability Localization Density



Satisfaction of input data



Contact map for the SEA complex resulting from the major cluster. The proximities of any two residues in the molecular architecture were measured by their relative contact frequency. A contact between a pair of residues was defined as an instance when their corresponding bead surfaces were less than 30 Å from each other. Cross-links are plotted as red dots, and the residue contact frequency is indicated by a color ranging from white (0) to dark blue (1). Each box contains the contact frequency between the corresponding pair of SEA complex proteins