

From PIPSA to MultiPIPSA

Stefan Richter

HITS gGmbH

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What is PIPSA

- Protein Interaction Property Similarity Analysis
- Method developed by Blomberg N, Gabdoulline RR, Nilges M, and Wade 1999.
- Compares interaction fields (electrostatic potential) on superimposed structures and calculates heatmap or cluster tree
- Initially available as commandline tool, since 2007 also as webserver: <http://pipsa.h-its.org>
- Now presented here with extensions as a python module and as you see later by Lukas used in a jupyter notebook.

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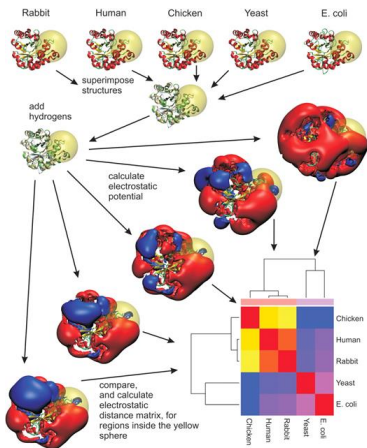
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PIPSA example



Stefan Richter, Anne Wenzel, Matthias Stein, Razif R. Gabdoulline, Rebecca C. Wade; Nucleic Acids Research, 2008

Prerequisites

- Python 2 or 3 is possible, ideally use anaconda (see next slide).
- Download multipipsa distribution, unpack
- Installing multipipsa

```
1 conda activate mypipsa
2 python setup.py install
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Use of Anaconda

Useful commands

```
1  # Setting addition sources for packages in conda
2  conda config --add channels bioconda
3  # Create an environment that uses python 3.4 as default
4  conda create -n myenv python=3.4
5  # Install the scipy package with a give version number
6  conda install -n myenv scipy=0.15.0
7  # make all packages/interpreters installed in myenv available
8  conda activate myenv
9  # Create a new environment based on an existing environment
10 conda create --name myclone --clone myenv
11 # Write out the information about a specific environment to a file
12 conda list --explicit > spec-file.txt
13 # Create an environemnt based on a file
14 conda create --name myenv --file spec-file.txt
15 # Install into the current environement all packages from a file
16 conda install --name myenv --file spec-file.txt
17 # List available environments from a conda installation
18 conda info --envs
19 # Get rid of the conda environment
20 conda deactivate
```

Example from multipipsa/testMultipipsa.py

Make sure, your structures are

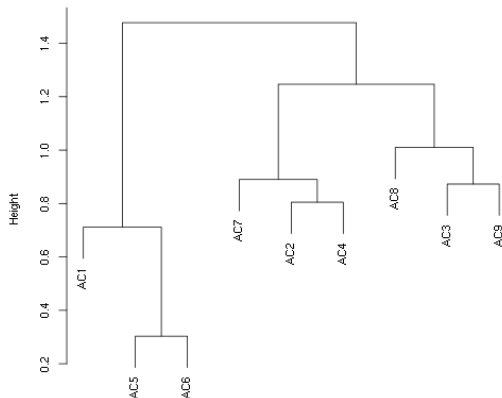
Running APBS

```
1 ingrp = ["AC1", "AC5", "AC6"]
2 outgrp = ["AC2", "AC3", "AC4", "AC7", "AC8", "AC9"]
3 apbs = ApbsRun(
4     dataDir=os.getcwd()+"/example/pdbs",
5     temp=298.15, ios=0.100, structures=ingrp+outgrp)
6 apbs.runPdb2Pqr()
7 apbs.runApbs()
```

Standard PIPSA

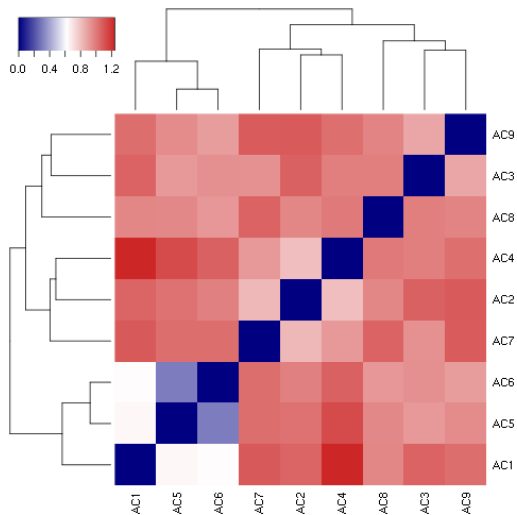
```
1 pr.runPipsa(ingrp+outgrp)
2 #####
3 cl0 = ClusterPipsa(structures=ingrp+outgrp,pipsaRoot="../../pipsa",
4                   dataDir=os.getcwd()+"/example/pdfs",
5                   distanceType=DistanceType.PIPSA,
6                   graphicsFileRoot="clust0")
7 pr.runClusterPipsa(ingrp+outgrp,
8                   points=[],cluster=cl0)
9 #####
10 cl1 = ClusterPipsa(structures=ingrp+outgrp,pipsaRoot="../../pipsa",
11                  dataDir=os.getcwd()+"/example/pdfs",
12                  distanceType=DistanceType.PIPSA,
13                  graphicsFileRoot="clust1")
14 pr.runClusterPipsa(ingrp+outgrp,
15                  points=pr.getStructure().getCAAtoms(),cluster=cl1)
16 #####
17 cl2 = ClusterPipsa(structures=ingrp+outgrp,pipsaRoot="../../pipsa",
18                  dataDir=os.getcwd()+"/example/pdfs",
19                  distanceType=DistanceType.PIPSA,
20                  graphicsFileRoot="clust2")
21 pr.runClusterPipsa(ingrp+outgrp,
22                  points=pr.getStructure().getCAAtoms()[10],cluster=cl2)
```

Standard PIPSA – Results



```
structure(c(0.641872261435248, 0.627694193059009, 1.06301458127346,  
1.06677082824757, 1.24177292610203, 1.09361784915939, 0.965401470891774,
```

Standard PIPSA – Results



Example from

`multipipsa/testMultipipsa.py`

Calculate Binding Scores

Reproduces the Fig. 5 of the binding scores of the reference in the footnote below.

```
1 pr = PipsaRun(pipsaRoot="../../pipsa",  
2     dataDir=os.getcwd()+"/example/pdbs",  
3     pointsTemplate="AC5")  
4 pr.runBindingScorePipsa(ingrp, outgrp)  
5 pr.savePDBResult(filename="SimHodgkin.pdb",  
6     scoreType=ScoreType.MS,  
7     similarityType=SimilarityType.HODGKIN)
```

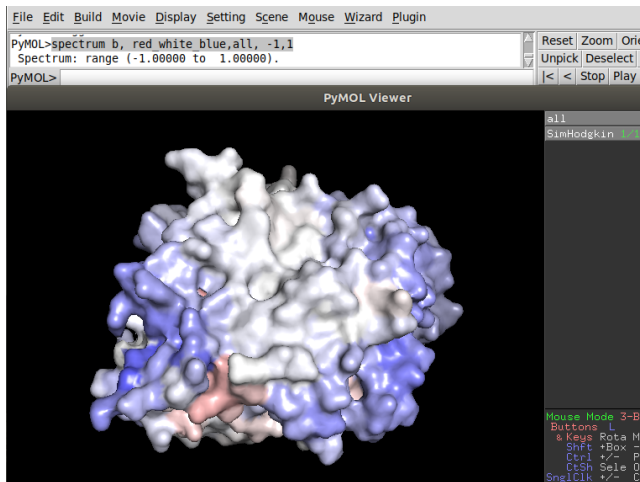
Tong, Wade and Bruce, Comparative electrostatic analysis of adenylyl cyclase for isoform dependent regulation properties., Proteins 2016; 84:1844-1858; doi: 10.1002/prot.25167.

Score is saved in B-Factor column

Excerpt from SimHodgkin.pdb file of previous example

1	ATOM	2364	CZ	PHE	B	300	38.312	-9.485	36.510	0.00	0.20	C
2	ATOM	2365	N	GLN	B	301	44.946	-8.363	34.106	0.00	0.38	N
3	ATOM	2366	CA	GLN	B	301	46.308	-8.562	33.712	0.00	0.38	C
4	ATOM	2367	C	GLN	B	301	46.943	-9.327	34.831	0.00	0.38	C
5	ATOM	2368	O	GLN	B	301	46.523	-9.218	35.981	0.00	0.38	O
6	ATOM	2369	CB	GLN	B	301	47.086	-7.243	33.560	0.00	0.38	C
7	ATOM	2370	CG	GLN	B	301	46.532	-6.335	32.460	0.00	0.38	C
8	ATOM	2371	CD	GLN	B	301	47.373	-5.066	32.412	0.00	0.38	C
9	ATOM	2372	NE2	GLN	B	301	47.920	-4.745	31.209	0.00	0.38	N
10	ATOM	2373	OE1	GLN	B	301	47.539	-4.378	33.417	0.00	0.38	O
11	ATOM	2374	N	MET	B	302	47.972	-10.137	34.521	0.00	0.44	N
12	ATOM	2375	CA	MET	B	302	48.602	-10.927	35.540	0.00	0.44	C
13	ATOM	2376	C	MET	B	302	49.886	-10.255	35.926	0.00	0.44	C

Visualization in pymol



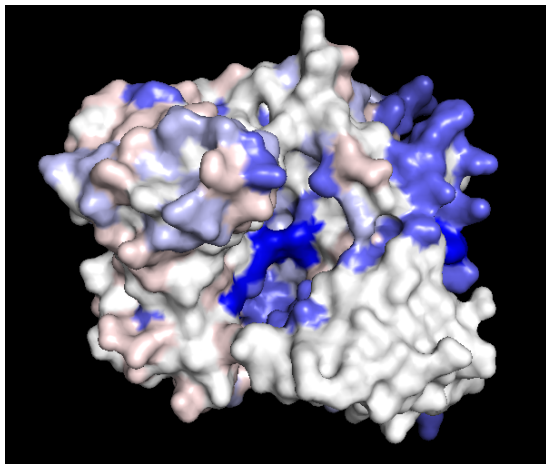
Select surface representation in pymol and enter the command `spectrum b, red_white_blue, all, -1,1`

Standard PIPSA – Compare Result to given cluster

Standard PIPSA

```
1 referenceCluster = np.array([1,1,1,2,2,2,2,2,2])
2 cl = ClusterPipsa(structures=ingrp+outgrp,pipsaRoot="../../../pipsa",
3                   dataDir=os.getcwd()+"/example/pdbs",
4                   referenceCluster=referenceCluster,
5                   distanceType=DistanceType.PIPSA,
6                   graphicsFileRoot="clust")
7 pr.runClusterPipsa(ingrp+outgrp,cluster=cl)
8 pr.savePDBResult(filename="ClusterCorrectedRand.pdb",scoreType=
9                   ScoreType.CORRECTED_RAND,
10                  clusterObject=cl)
```

Standard PIPSA – Areas with very similar clusters as the reference cluster



Standard PIPSA – Areas with high or low information content in the clustering

