Lessons learned from two D3R grand challenges rounds

Prof. Alexandre M.J.J. Bonvin Bijvoet Center for Biomolecular Research Faculty of Science, Utrecht University the Netherlands a.m.j.j.bonvin@uu.nl



Utrecht Bioinformatics Center

Overview

Introduction Pose predictions Lessons from GC2 Applying them to GC3 Binding affinity/ranking Conclusions

HADDOCK: An integrative modeling platform



Incorporates ambiguous and low-resolution data to aid the docking

Capable of docking up to 6 molecules

Symmetries can be leveraged

Powerful algorithms to handle flexibility at the interface

Final flexible refinement in explicit solvent

One of the best performing software in CAPRI



Dominguez, Boelens & Bonvin. JACS 125, 173 (2003).





Succession of energy minimization and molecular dynamics protocols

reminiscent of NMR structure calculations



HADDOCK's user base







HADDOCK protein-ligand protocol



- Receptor binding site:
 - active during it0 to attract the ligand into the pocket.
 - passive for flexible stages to allow exploration by the ligand.
- Parameters:
 - Protein: OPLS (Jorgensen and Tirado-Rives 1988)
 - Ligand: PRODRG (Schüttelkopf and van Aalten 2004)
- No high-T search in semi-flexible refinement.
- Modified scoring: w_vdw_it0 = 1, w_elec_water = 0.1
- Pooled results from two runs in case of buried binding site:
 - Standard with ligand settings (as above)
 - Buried-site settings (k_inter=0.001, w_vdw_it0 = 0)





Unbound receptor docking results

71 cases from the Astex NonNative Set¹

10Å interface				
	Cluster quality water – combined analysis			
	Top 1	Top 2	Тор 3	Top 5
#total	71	71	71	71
#success *	41	49	55	58
#success **	23	26	28	29
#success ***	2	2	2	2
%success *	57,7	69,0	77,5	81,7
%success **	32,4	36,6	39,4	40,8
%success ***	2,8	2,8	2,8	2,8

*: ligand-interface RMSD < 2.0Å

**: ligand-interface RMSD < 1.0Å

***: ligand-interface RMSD < 0.5Å

1. Verdonk et al. J. Chem. Inf. Model. 2008

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GC2: The challenges







Zeynep Kurkcuoglu



end



Panos Koukos

[Faculty of Science Chemistry]





Zeid et al.

Flexible helices at the receptor proximal



GC2: Ligand generation

- Generate 3D conformers from the SMILE strings (using the OpenEye Omega Toolkit)
- Cluster (hierarchically) the conformers and select representative structures
- Ensemble docking in HADDOCK







GC2: Receptor for docking

- BLAST Apo protein sequence against the PDB
- Remove problematic entries, cluster the proteins based on binding pocket backbone-RMSD
- Mutate as necessary, select representatives (4 for stage 1)
- Binding pocket defined as the union of all the residues within 5Å of all the ligands









FXR-34: I-RMSD of 1.94Å

FXR-27: I-RMSD of 1.17Å





Average best RMSD from 5 poses per ligand





What limits our performance?





Improving our docking performance in Stage2

- Tackling the receptor limiting factor:
- Select receptor template based on the similarity of the target ligand to the template ligand
- Tanimoto coefficient

- Tackling the ligand limiting factor:
- Up-sample the major cluster if possible







Stage1 vs Stage 2 performance







Impact on quality and scoring







I-RMSD <= 2.5Å 2.5Å < I-RMSD <= 3.5Å I-RMSD > 3.5Å



Stage 2

Impact on quality and scoring



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I-RMSD <= 2.5Å 2.5Å < I-RMSD <= 3.5Å I-RMSD > 3.5Å

GC2: Main lessons



- Lesson 1: Smart choice of receptor conformation is crucial
- Lesson 2: Need to better select ligand conformations for docking
- Lesson 3: Our docking protocol starts from randomly rotated, separated conformations
 -> need for better starting conformations



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D3R 2017 Grand Challenge 3

 Protein (cathepsin) against 141 (24) small molecules





Selecting the receptor

- Identify templates with high sequence identity (>70%) to the target protein sequence with at least one bound ligand.
- Compare the crystallographic ligand to the Cathepsin set using the Maximum Common Substructure (MCS) (as implemented in ChemmineR)
- Select the receptor with the highest Tanimoto Coefficient (GC2 lesson 1)











Selecting the receptor



Ligand similarity of best template found





Ligand conformation

- Compare similarity of crystallographic template ligands with generated conformers using shape and color tanimoto (as implemented in OpenEye ROCS)
- Select 10 conformers with the highest combined score for ensemble docking (GC2 Lesson 2)









Starting conformations

- Superpose the selected conformers on the crystallographic ligands (OpenEye shape-TK)
- Refine using HADDOCK only short minimization in 2nd stage and final refinement in explicit solvent



(GC2 Lesson 3)

• For stage 1b, receptor, water molecules and other small molecules kept rigid in place (not much impact on performance)





Docking results



Stage 1 heavy-atom RMSD



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Lessons learned!

GC3 – Excellent performance

Receipt ID

Green bar indicates your predictions (requires login)



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• Nothing changed between GC2 and GC3

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• Structure-based prediction using an atomic contact model (see our GC2 paper)



Kurkcuoglu et al. J. Comp. Aid. Mol. Des. 2017



• Ligand-based prediction using target-specific ligand similarity SVR model (see our GC2 paper)



Kurkcuoglu et al. J. Comp. Aid. Mol. Des. 2017





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Conclusions



as catalyzer for learning from failures

- Success factors (for us):
 - Smart selection of receptor and ligand conformations
 - Smart positioning of ligand in binding pocket

Kurkcuoglu et al. J. Comp. Aid. Mol. Des. 2017



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VICI TOP-PUNT

JWC

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WeNMR West-Life EGI-Engage INDIGO-Datacloud BioExcel CoE EOSC-Hub



Thank you for your attention!



HADDOCK online:

- http://haddock.science.uu.nl
- http://bonvinlab.org/software



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