

DrugDesignData Resource (D3R) Grand Challenge 2

Dr. Matthew Baumgartner

Dr. David Evans

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Diagnostic
and
Drug Discovery
Initiative
for
Alzheimer's Disease



invest

Lilly

Dr. Matthew Baumgartner

- ◆ Marie Curie Post Doctoral Fellow at Eli Lilly and Company
- ◆ Ph.D. in Computational Biology at the University of Pittsburgh under Dr. Carlos Camacho

D3R 2016 Grand Challenge Data

- ◆ Industry affinity and structural data donated to D3R organizers by Roche
 - One target: farnesoid x receptor (FXR)
 - IC50 data for 102 compounds in 4 chemical series (potency range of 0.000335 – 62.37 μM for 92 compounds, and 10 having potency $> 100 \mu\text{M}$)
 - 36 co-crystal structures and 1 *apo* (resolution 1.8 - 2.6 Å)

The Challenge

◆ Phase 1

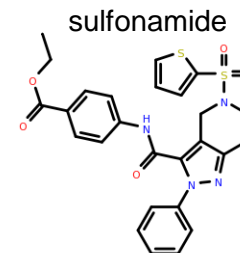
- Predict poses of 36 compounds with blinded co-crystal structures (can submit up to 5 poses/method)
- Predict or rank the potencies of all 102 ligands
- **After submission** deadline (Nov 21, 2016) all of the crystal structures were released

◆ Phase 2

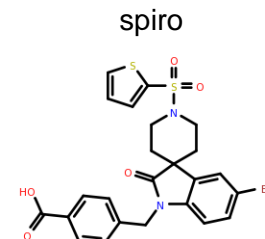
- Re-predict the potencies of all 102 ligands
- **After competition** (Feb 1, 2017) all of the affinity data was released

Test set data

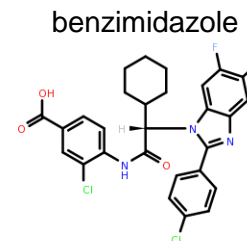
- ◆ Test set compounds:
 - 23 sulfonamides
 - 22 spiros
 - 47 benzimidazoles (PDB crystals)
 - 4 isoxazoles (PDB crystals)
 - 6 others (3 PDB crystals)
- ◆ 26 structures in the PDB in 3 sets:
 - 7 benzimidazoles (references: 3OMM, 3OKH)
 - 9 isoxazoles (reference: 3DCU)
 - 10 others (3 with similarity to the test set “others”)



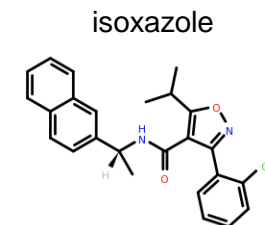
FXR_17



FXR_10



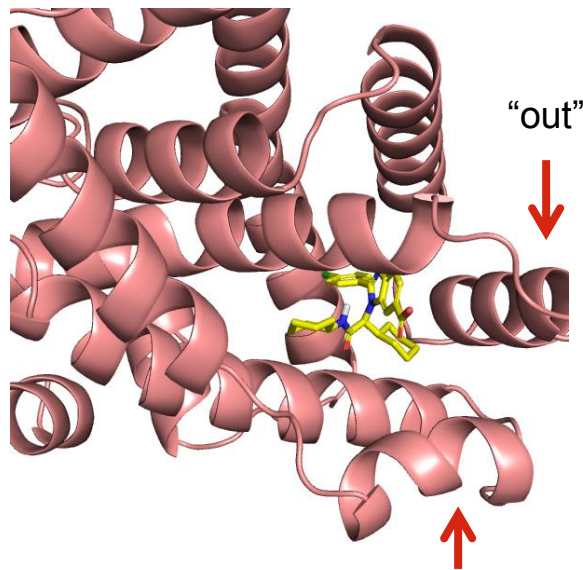
FXR_27



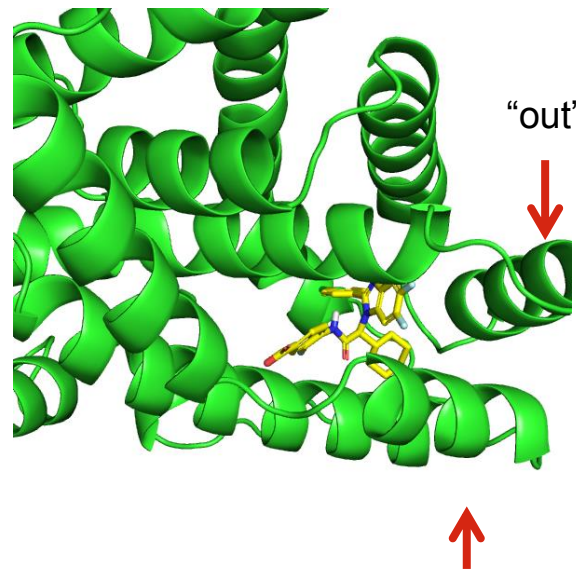
FXR_23

PDB reference structures

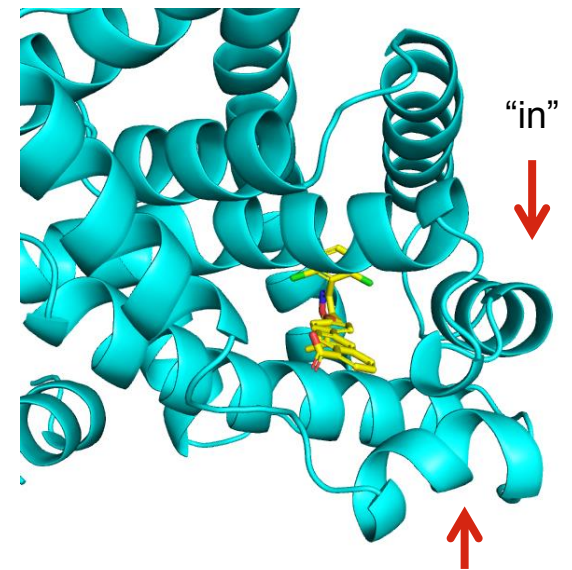
- ◆ Examined the PDB structures and identified three major conformations of the binding site



30KH



30MM



3DCU

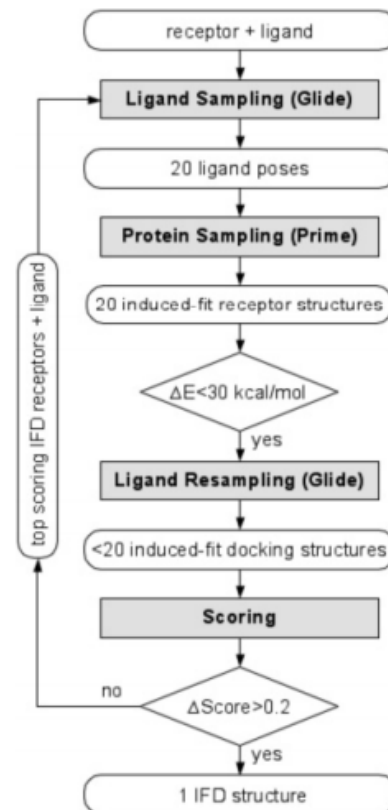
Prediction Approaches

- ◆ Pose prediction
 - **Align-Close** – Generate conformers and align (Cresset Forge) to most similar (by fingerprint) co-crystal ligand and minimize with Vina scoring function
 - **Smina docking** – rigid receptor, flexible ligand (Vina)
 - **Induced fit docking (IFD)** – flexible receptor side chains and ligand
 - **Metadynamics** - pose re-ranking of IFD poses
 - **Manual** – Manually look at all of the data and choose my top poses

| | |
|---------------|-------|
| Align-close | wax1j |
| Smina docking | x7jp3 |
| IFD | piwlh |
| MetaD | qfu33 |
| Manual | psiuj |

Schrodinger's Induced Fit Docking

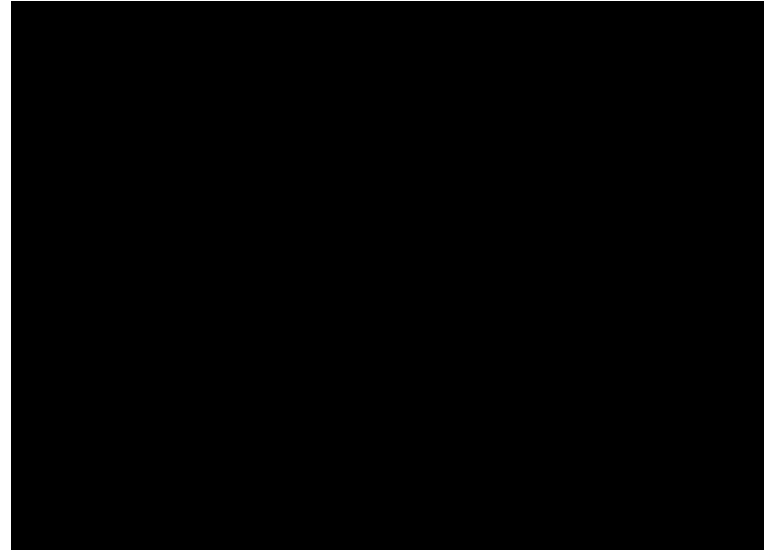
- ◆ Docking which allows protein side chains to move
- ◆ Iterative sampling of the ligand (with Glide) and the protein (with Prime)



Sherman, W.; Day, T.; Jacobson, M. P.; Friesner, R. A.; Farid, R., "Novel Procedure for Modeling Ligand/Receptor Induced Fit Effects," J. Med. Chem., 2006, 49, 534-553

Binding Pose Metadynamics

- ◆ Uses a metadynamics simulation to determine the force needed to displace a ligand from a starting position (IFD pose)
- ◆ The more force required to displace it, the better the pose was
- ◆ Schrodinger uses a complicated measure of the ligand RMSD and a measure of the number of contacts broken as the collective variable (CV)
- ◆ At the end, you get a score of the input pose



Structure Prediction Methods

- ◆ Docking of the test set compounds to the three reference receptors with smina
 - Took the top 5 scoring poses across the three dockings
- ◆ IFD + Metadynamics of test set compounds to 3 reference receptors
 - Top 5 scoring IFD poses were rescored using MetaD
- ◆ Submitted poses from top 5 IFD scored poses and top 5 MetaD scored poses (not necessarily the same poses)

My Results

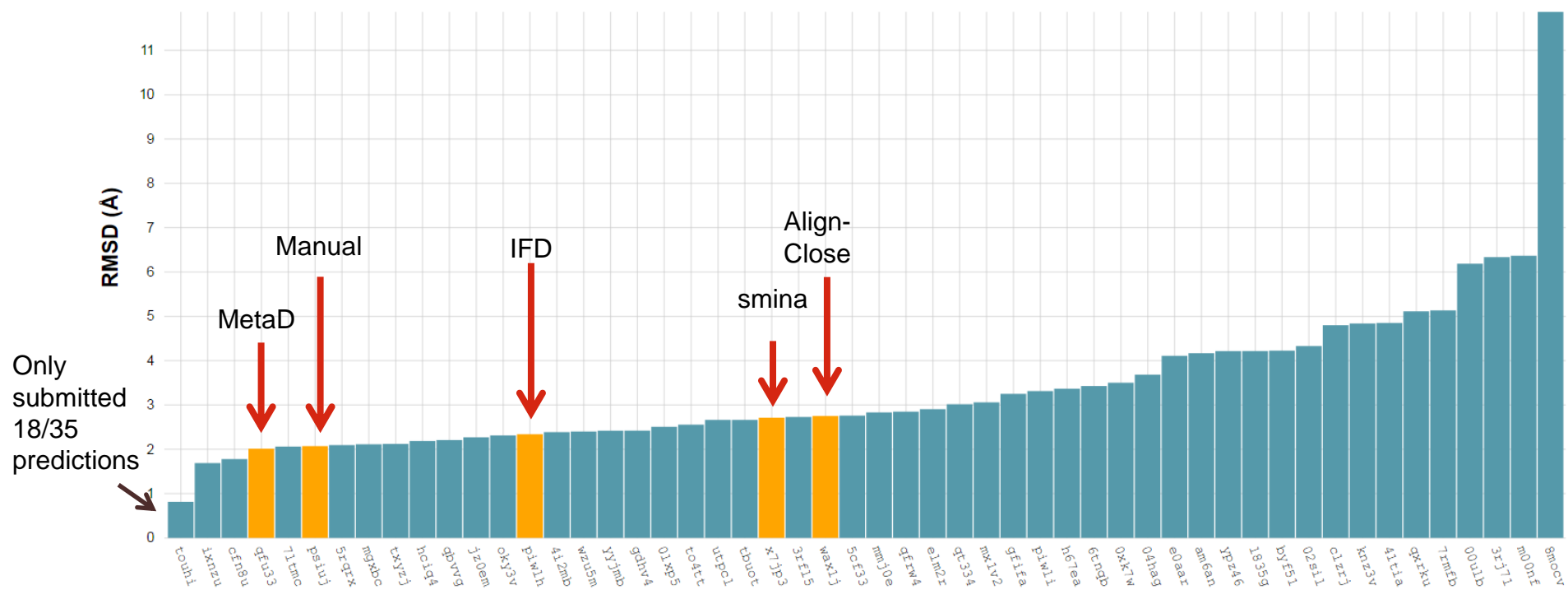
- ◆ Number of structures (out of 35) with a pose under 2A in the top N poses

| | Top 5 | Top 3 | Top 1 |
|--------------|-----------|-----------|-----------|
| Align-close | 21 | 21 | 20 |
| Smina | 19 | 19 | 11 |
| IFD | 19 | 15 | 14 |
| Metadynamics | 23 | 19 | 16 |
| Manual | 23 | 23 | 18 |

- ◆ Metadynamics helped over IFD alone
- ◆ Smina samples just as well as IFD but is much faster
- ◆ Align-Close utilized the most structural info and did the best
- ◆ My manual picking also did pretty well which was nice

- The ligand one of the structures (FXR_33) does not match the test set smiles possibly due to oxidation during the crystallization process so it is excluded from all analysis
- Also note that some of the crystal structures have been identified as having errors (non-planar aromatic rings) so these may change slightly

Overall Results – Best Pose



| Method | Mean RMSD of lowest RMSD pose |
|-------------|-------------------------------|
| MetaD | 2.01 |
| Manual | 2.07 |
| IFD | 2.34 |
| Smina | 2.71 |
| Align-Close | 2.75 |

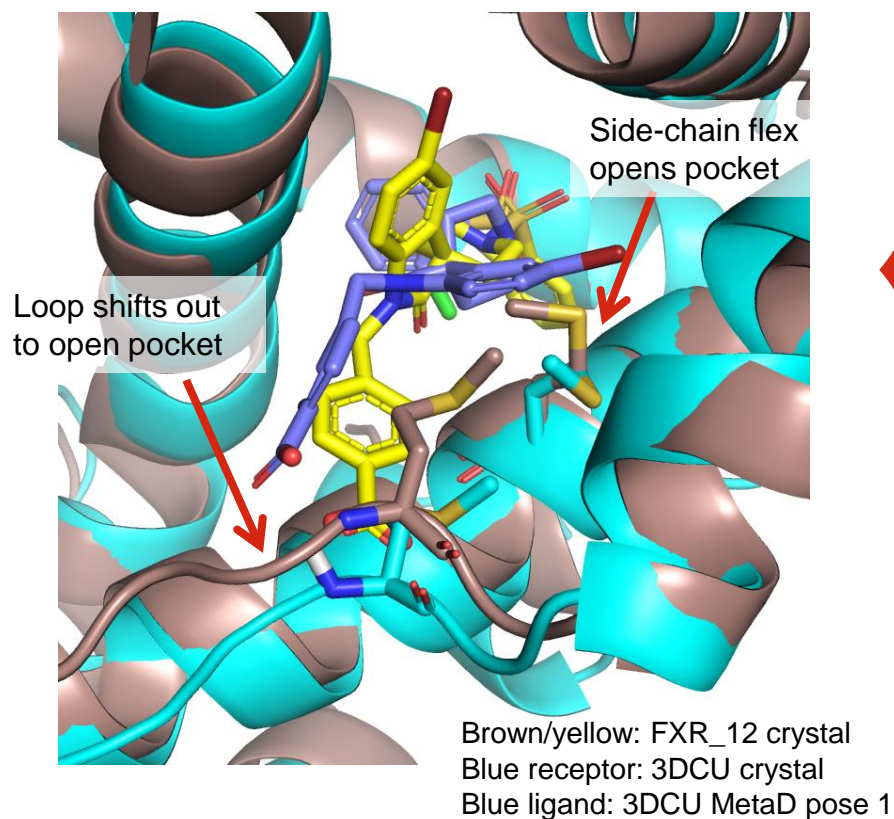
Broken Down by Scaffold

- ◆ All of my methods were successful only on the most common scaffold (benzimidazole)
- ◆ They all did poorly on the other compound scaffolds

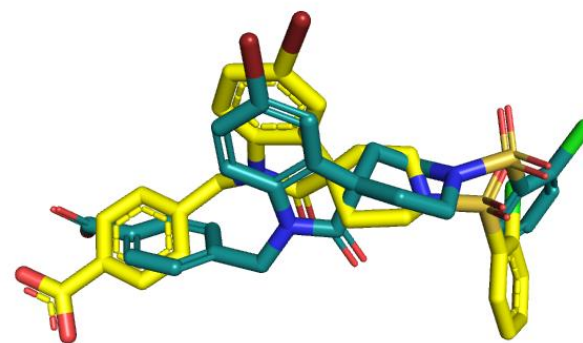
Number of compounds with pose under 2 Å RMSD in the top 5 poses

| | Align Close | Smina | IFD | MetaD | Manual | Total |
|---------------|-------------|----------|----------|----------|----------|-------|
| benzimidazole | 20 | 18 | 17 | 20 | 21 | 22 |
| isoxazole | 0 | 1 | 0 | 1 | 1 | 2 |
| spiro | 0 | 0 | 0 | 0 | 0 | 3 |
| sulfonamides | 1 | 0 | 0 | 1 | 0 | 4 |
| others | 1 | 0 | 1 | 1 | 1 | 4 |

No good poses for any of the spiros



- ◆ Significant differences in the crystals used to predict the pose and the actual pose limited predictions
- ◆ The FXR_12 structure is fairly dissimilar to the other structures (avg pocket RMSD $3.6 \text{ \AA} \pm 1.5$)



Yellow: FXR_12 crystal
Green: Lowest RMSD pose
from all methods (IFD, pose 4)

Is using an ensemble score better?

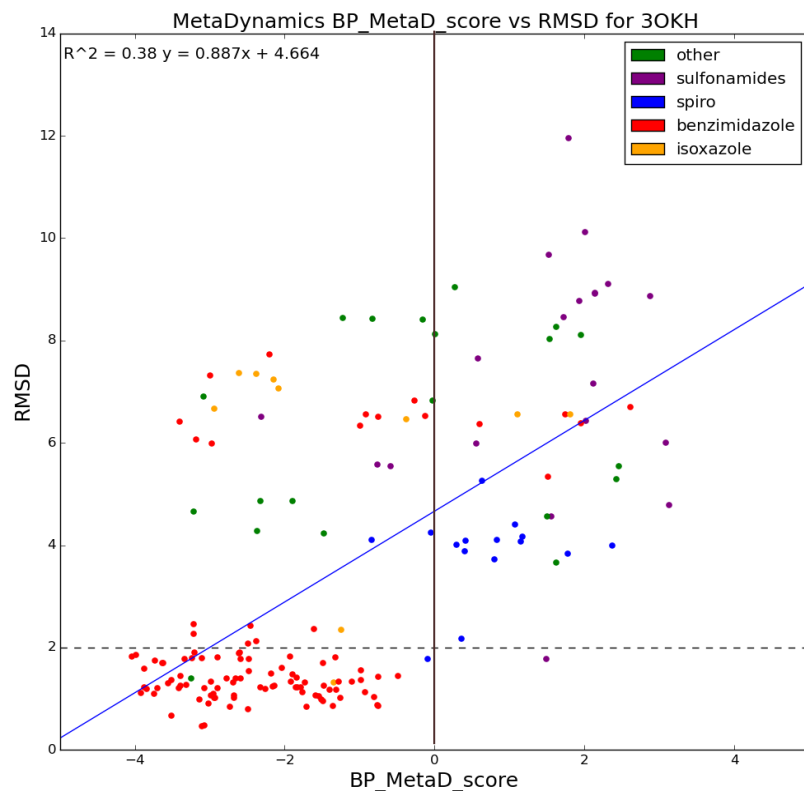
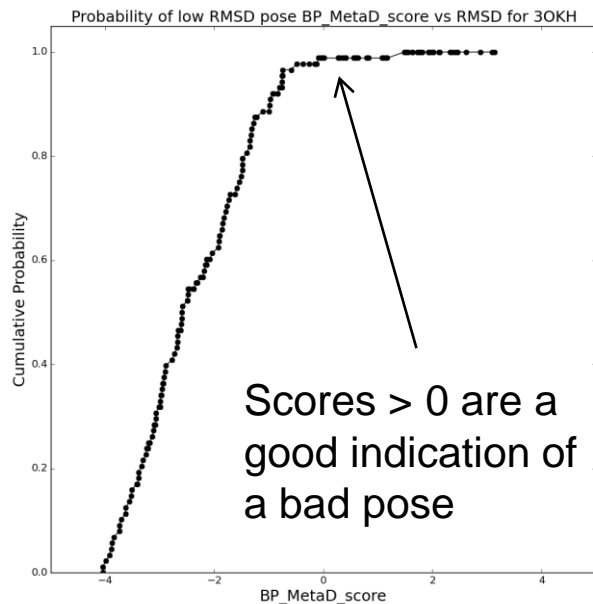
- ◆ For the metadynamics, I took the top 5 scoring poses from the IFD runs to the three structures
- ◆ Was this a good idea?
- ◆ Not really, 3DCU performed terribly
- ◆ Just using 3OKH would have done better than the ensemble method (23/35)

Minimum RMSD of the top 5 poses from Metadynamics

| Compound ID | testset_scaffold | 3OKH | 3OMM | 3DCU | |
|------------------------|------------------|-----------|-----------|----------|--|
| FXR_13 | benzimidazole | 1.21 | 1.39 | 6.16 | |
| FXR_14 | benzimidazole | 1.79 | 1.74 | 5.86 | |
| FXR_19 | benzimidazole | 0.46 | 1.23 | 6.93 | |
| FXR_2 | benzimidazole | 6.55 | 6.09 | 2.16 | |
| FXR_20 | benzimidazole | 0.91 | 1.54 | 5.12 | |
| FXR_21 | benzimidazole | 1.2 | 1.64 | 6.44 | |
| FXR_22 | benzimidazole | 1.78 | 1.51 | 7.45 | |
| FXR_24 | benzimidazole | 1.11 | 1.64 | 6.37 | |
| FXR_25 | benzimidazole | 1.1 | 1.46 | 7.76 | |
| FXR_26 | benzimidazole | 1.19 | 1.64 | 3.96 | |
| FXR_27 | benzimidazole | 1.04 | 1.62 | 6.74 | |
| FXR_28 | benzimidazole | 0.86 | 1.55 | 6.86 | |
| FXR_29 | benzimidazole | 0.86 | 1.65 | 2.73 | |
| FXR_30 | benzimidazole | 1.23 | 2.35 | 3.94 | |
| FXR_31 | benzimidazole | 1.05 | 2.18 | 4.45 | |
| FXR_32 | benzimidazole | 1.07 | 1.7 | 7.37 | |
| FXR_35 | benzimidazole | 0.96 | 1.68 | 1.75 | |
| FXR_36 | benzimidazole | 0.84 | 1.32 | 5.21 | |
| FXR_6 | benzimidazole | 1.19 | 1.76 | 4.18 | |
| FXR_7 | benzimidazole | 0.84 | 1.48 | 5.31 | |
| FXR_8 | benzimidazole | 6.33 | 6.59 | 6.1 | |
| FXR_9 | benzimidazole | 1.07 | 7.3 | 5.79 | |
| FXR_23 | isoxazole | 1.32 | 2.18 | 4.65 | |
| FXR_4 | isoxazole | 6.46 | 5.61 | 5.1 | |
| FXR_18 | other | 8.13 | 8.58 | 4.9 | |
| FXR_3 | other | 1.39 | 1.98 | 4.76 | |
| FXR_34 | other | 3.66 | 5.62 | 2.25 | |
| FXR_5 | other | 4.23 | 5.06 | 7.75 | |
| FXR_10 | spiro | 1.77 | 2.27 | 1.75 | |
| FXR_11 | spiro | 3.84 | 3.87 | 2.41 | |
| FXR_12 | spiro | 3.72 | 2.01 | 2.11 | |
| FXR_1 | sulfonamides | 5.98 | 3.53 | 5.41 | |
| FXR_15 | sulfonamides | 8.77 | 5.87 | 5.11 | |
| FXR_16 | sulfonamides | 1.78 | 5.23 | 1.92 | |
| FXR_17 | sulfonamides | 4.56 | 4.46 | | |
| Total under 2 Å | | 24 | 18 | 3 | |

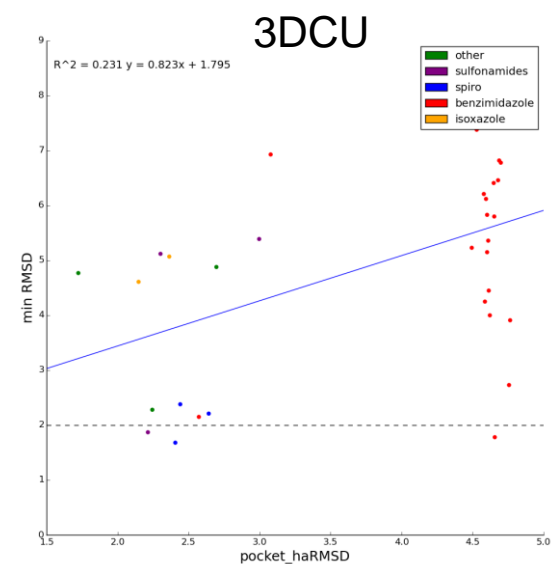
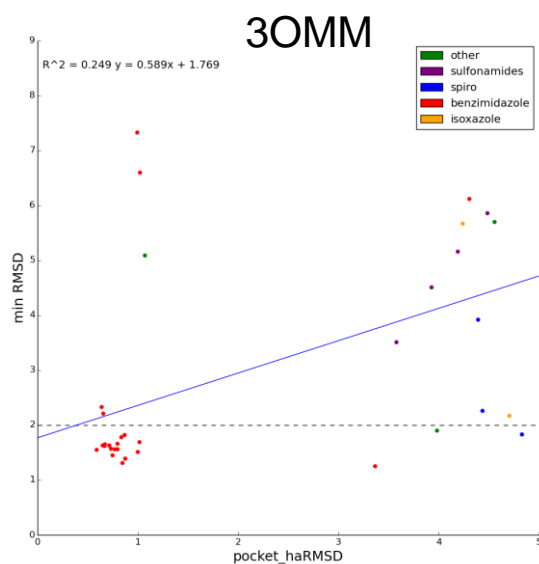
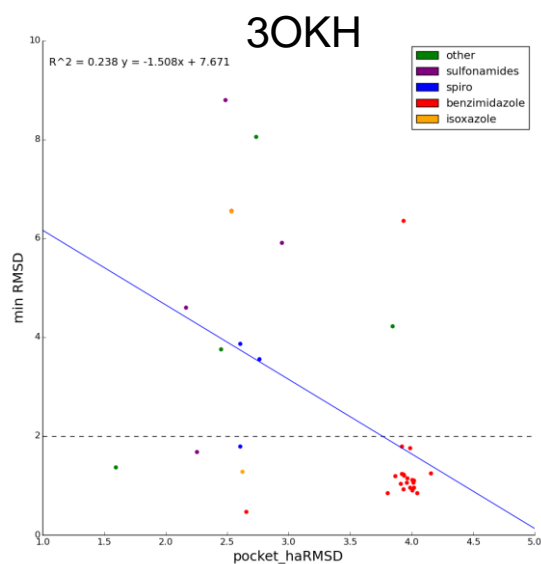
Examining the BPMetaD Score

- ◆ Seems to do pretty well at determining if a pose is bad
 - 96%+ True negative rate
 - 69.6% True positive rate



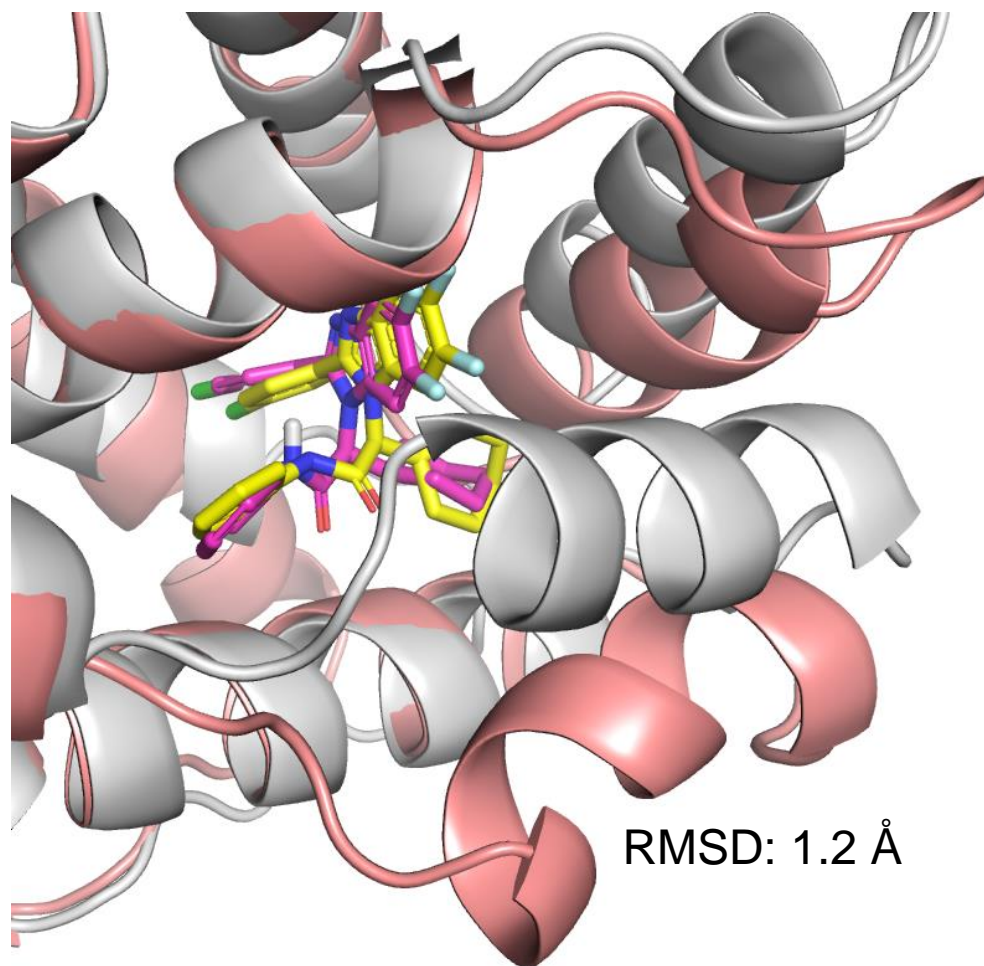
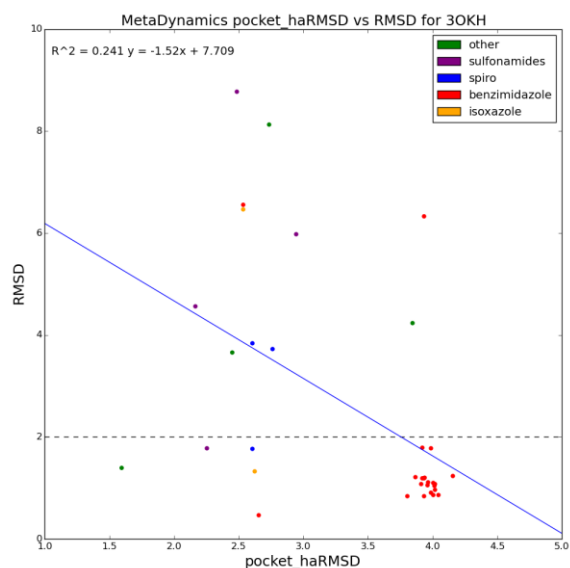
Worse structures generally give worse predictions

- ◆ Calculated the heavy atom RMSD of pocket residues between the three reference structure and the actual crystals
- ◆ Generally more changes lead to worse predictions
- ◆ Except for 3OKH where the benzimidazoles are tolerant of very different structures



Large differences in the pocket tolerated for the benzimidazoles

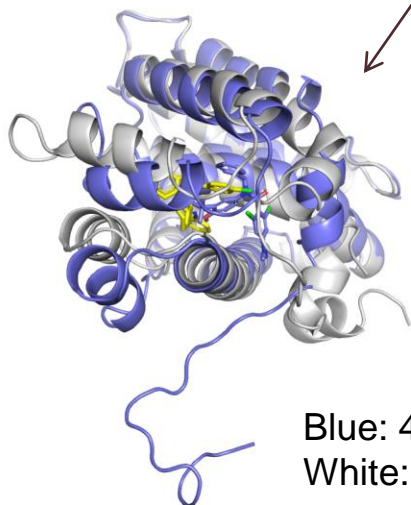
- ◆ Most of the benzimidazoles docked well to 3OKH using IFD + MetaD despite large differences in pocket residues
- ◆ The ligands site far enough back in the pocket further from the major difference



White/yellow: FXR_21 crystal
Peach: 3OKH crystal
Purple: Best predicted pose from MetaD

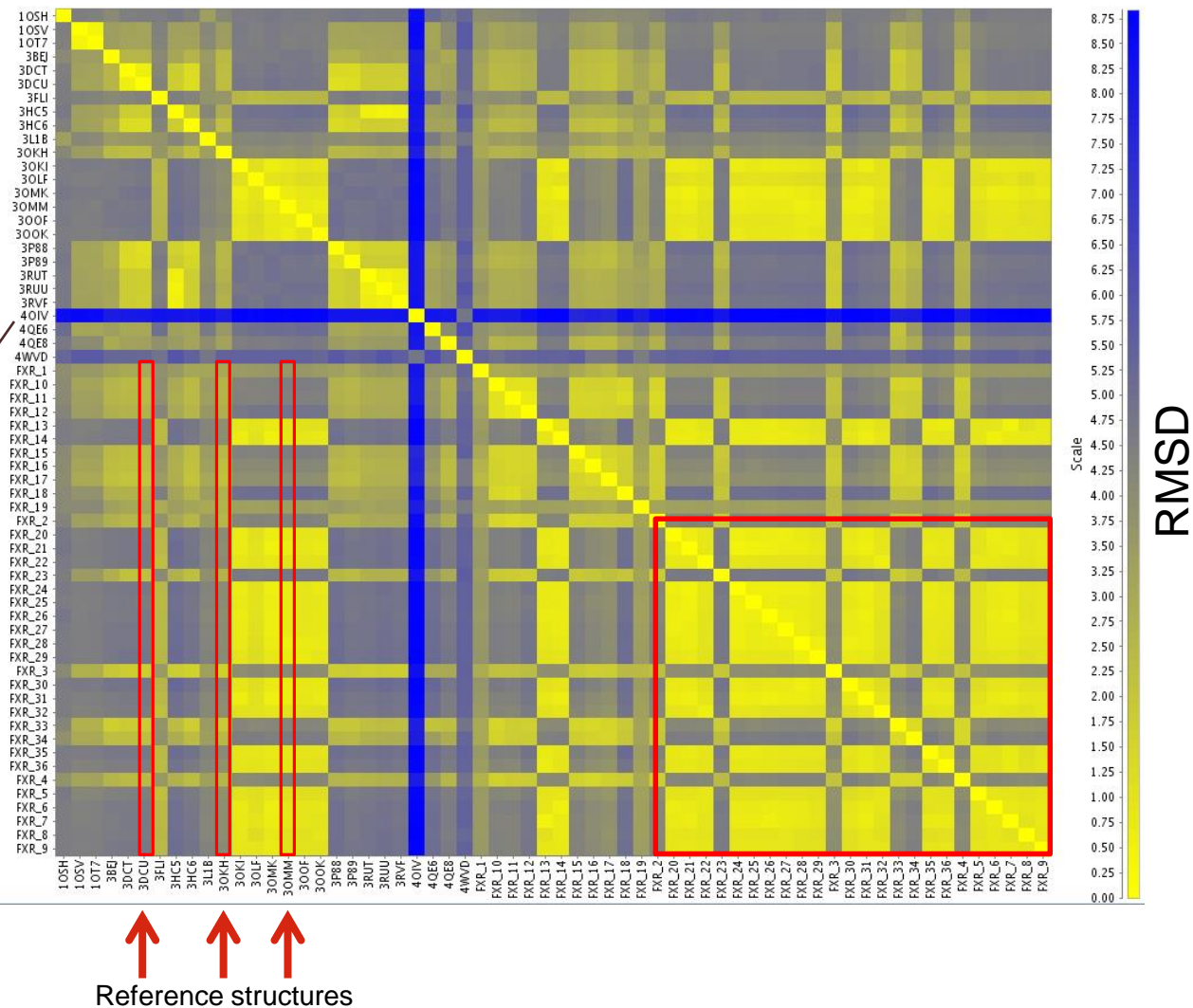
Diversity of Binding Sites

- ◆ A large number of the FXR compounds are self similar
- ◆ Despite inferior performance, 3OMM is more similar to the FXR structures than 3OKH



Blue: 4OIV
White: FXR_21

Pocket Residues Heavy atom RMSD



Visual inspection of the models provides accurate prediction of model quality

- ◆ For my final Manual method, I visually inspected the poses generated by all of the other models and manually chose poses
- ◆ I noted my assessment of the quality of the pose during the selection process
- ◆ My evaluation of the 3OKH Metadynamics poses compares favorably to the actual RMSDs

| Compound ID | Manual Evaluation | Min RMSD |
|-------------|-------------------|----------|
| FXR_19 | very good | 0.47 |
| FXR_29 | good | 0.84 |
| FXR_36 | good | 0.84 |
| FXR_28 | very good | 0.9 |
| FXR_7 | very good | 0.92 |
| FXR_20 | very good | 0.96 |
| FXR_35 | very good | 0.96 |
| FXR_32 | good | 1.03 |
| FXR_27 | very good | 1.05 |
| FXR_31 | very good | 1.05 |
| FXR_9 | very good | 1.1 |
| FXR_25 | very good | 1.11 |
| FXR_24 | very good | 1.14 |
| FXR_13 | good | 1.19 |
| FXR_21 | very good | 1.2 |
| FXR_26 | very good | 1.22 |
| FXR_6 | very good | 1.23 |
| FXR_30 | very good | 1.24 |
| FXR_23 | bad | 1.28 |
| FXR_3 | ok | 1.37 |
| FXR_16 | bad | 1.68 |
| FXR_22 | good | 1.76 |
| FXR_10 | bad | 1.79 |
| FXR_14 | very good | 1.79 |
| FXR_12 | bad | 3.55 |
| FXR_34 | ok | 3.76 |
| FXR_11 | bad | 3.87 |
| FXR_5 | bad | 4.22 |
| FXR_17 | bad | 4.6 |
| FXR_1 | bad | 5.91 |
| FXR_8 | bad | 6.35 |
| FXR_4 | ok | 6.54 |
| FXR_2 | bad | 6.55 |
| FXR_18 | ok | 8.05 |
| FXR_15 | bad | 8.8 |

Pose Prediction

- ◆ This was actually a quite tough challenge which is a departure from previous years
 - In the PL-2016-1 mini challenge, my average RMSD was 0.7 Å (N=5)
 - In the 2015 challenge our average RMSD was 0.32 Å (HSP90, N=6) and 1.32 Å (MAP4K4, N=30)
 - In the 2014 challenge my average RMSD was 1.2 Å (N=14)

Affinity Prediction

- ◆ Challenged to rank 102 compounds (including the 35 crystallized compounds) in 4 chemical series
- ◆ Able to submit predictions as part of Phase 1 and re-submit predictions for Phase 2 after receiving the crystal structures from Phase 1

My Affinity Prediction Methods

◆ Phase 1:

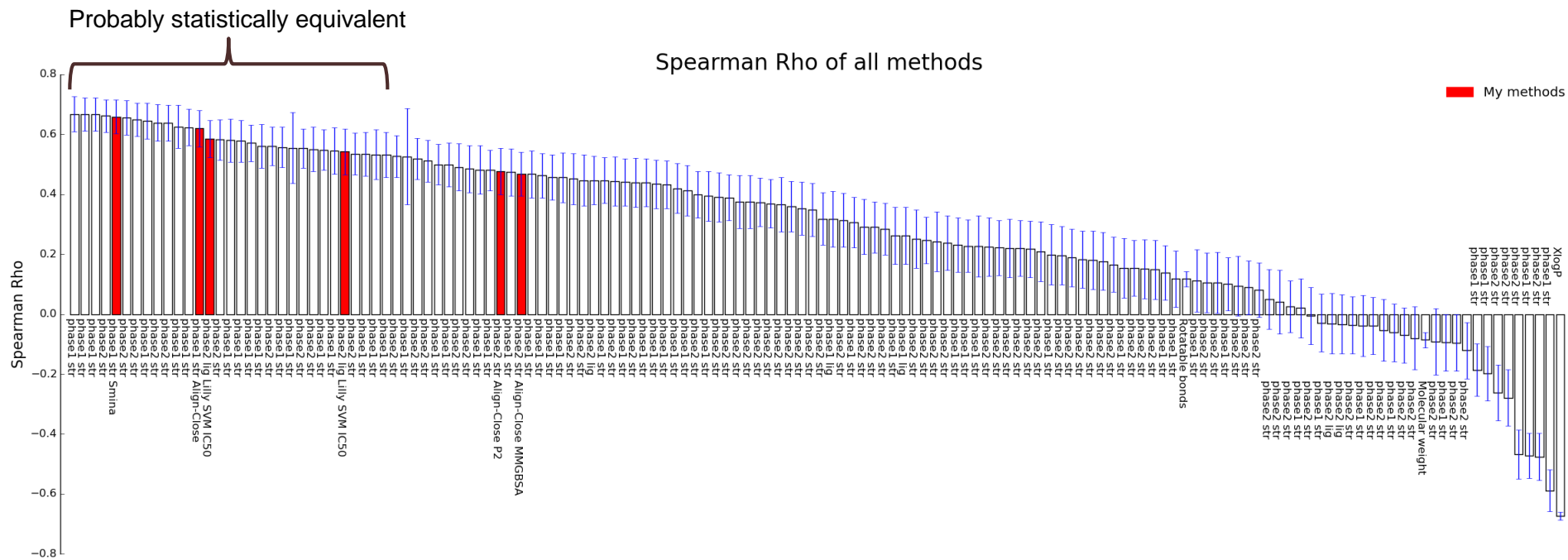
- **Smina ensemble docking** – Dock to the three reference structures and take the top scoring pose
- **Align-Close** – rank the compounds by the top scoring minimized pose (Vina)

◆ Phase 2:

- **Align-Close 2** – Same as for Phase 1 but includes 35 structures from Phase 1
- **Align-Close MMGBSA** – Minimize poses with MMGBSA
- **Lilly SVM QSAR** – 2D QSAR methods on trained using **IC50** or **EC50** data of FXR binders from ChEMBL

| | |
|--------------------|-------|
| Smina ensemble | gzd7a |
| Align-Close P1 | aaveo |
| Align-Close P2 | 6mjkt |
| Align-Close MMGBSA | vovuk |
| Lilly SVM IC50 | hj31e |
| Lilly SVM EC50 | naex2 |

All Affinity Prediction Results



- ◆ The Autodock Vina scoring function in Smina performed quite well
- ◆ Strangely, the Align-Close method did worse in Phase 2 when it had access to the additional structures from Phase 1
- ◆ The QSAR models were the best of the 7 ligand based methods and compared favorable to the structure based methods

Acknowledgements

- ◆ We thank the European Union's Seventh Framework Programme for Research, Technological Development, and Demonstration for funding the Diagnostic and Drug Discovery Initiative for Alzheimer's Disease (612347, 2014-2018)
- ◆ Schrödinger for providing temporary Desmond and Prime licenses for this competition and Davide Branduardi for critical advice on best usage of the Metadynamics simulations
- ◆ Ian Watson for assistance with the QSAR scripts