## Using the DeepAtom framework for DPI prediction Weekly update

Week 1 April, 2021

#### Overview

Data pre-processing: 3D grid creation

- i. Rasterization
- ii. Channelization

#### Network architecture

- i. Atom information integration block
- ii. Stacked feature extraction block
- iii. Global affinity regression block

#### Current progress

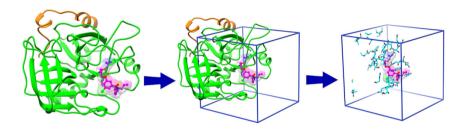


#### DeepAtom Overview

#### A 3D-CNN based approach to predicting binding affinity.

- 1. Input: "mostly" just the ligand part of the co-complex, rasterized into a 3D grid box.
- 2. Novel voxel representation: each voxel has 24 channels, "embedding the different raw information of atoms located around the voxel".
- 3. Grid processed using efficient 3D CNN implementation.
- 4. Final BAP created using "ensemble-like" training approach.

#### i. Rasterization (1/2)



A 3D grid of dimensions  $32 \times 32 \times 32$  ų with resolution 1 Å is created, centered on the ligand.

- 1. The greatest end-to-end distance of a ligand in PDBbind v.2016 is  $\sim$  32 Å.
- 2. The vdW radius of the heavy atoms found in PDBbind ligands is  $\geq 1.4$  Å.

#### i. Rasterization (2/2)

PCMax algorithm: when converting atoms to voxels, atoms-as-voxels have a continuous effect on their neighbors.

For each voxel, its occupancy taken as the maximum of

$$n(r_i) = 1 - \exp\left(-\left(\frac{r_{vdW,i}}{r_i}\right)^{12}\right)$$

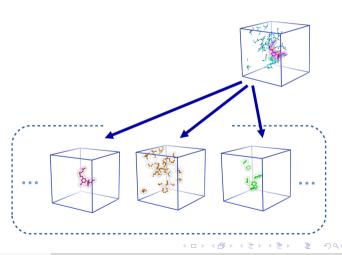
over all atoms i, where  $r_{vdW,i}$  is the  $i^{th}$  atom radius and  $r_i$  is the inter-voxel distance.

Reduces sparsity in input volumes, aids with convolution.

#### ii. Channelization

The input to the network is  $32 \times 32 \times 32 \times 24$ : 12 channels each created from the ligand and protein.

- ▶ 1 excluded volume channel.
- ► 11 channels of atoms involved in certain interatomic interactions.

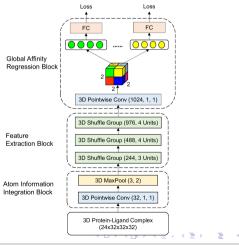


- i. Atom information integration block
- ii. Stacked feature extraction block
- iii. Global affinity regression block

#### The Three Block Overview

## DeepAtom is relatively shallow, and has three blocks:

- 1. Atom information integration block.
- 2. Stacked feature extraction block.
- 3. Global affinity regression block.



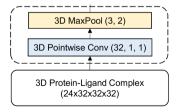
- i. Atom information integration block
- ii. Stacked feature extraction block

### i. Atom Information Integration Block

Atomic information across the 24 channels is aggregate in a pointwise fashion, and the output is then pooled.

- 1. PWConv $(W, h)_{(i,j,k)} = \sum_{m}^{24} W_m \cdot h_{(i,j,k,m)}$  is used to map the  $24 \times 32^3$  input to  $32^3$ .
- 2. The max-pooling downsamples the tensor to  $16^3$ .

Semantically, this block processes the input into non-linear function of the linear combination of the various channels (interaction types), parsimoniously.





- i. Atom information integration block
- ii. Stacked feature extraction block
- iii. Global affinity regression block

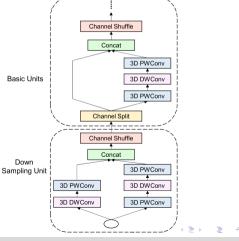
#### ii. Stacked Feature Extraction Block

Three 3D shuffle groups are stacked, each containing a parsimonious combination of pointwise and depthwise convolutions.

Shuffling, splitting, and then processing only certain channels encourages parsimonious while maintaining performance.

$$\mathsf{DWConv}(W,h)_{(i,j,k)} =$$

$$\Sigma_{s,t,r}^{S,T,R} W_{s,t,r} \odot h_{(i+s,j+t,k+r)}$$

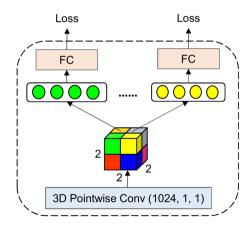


- i. Atom information integration block
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#### iii. Global Affinity Regression Block

# The output from the shuffle groups is downsampled and flattened into eight 1024-vectors.

- Each vector represents the output from a distinct channel subset, and is processed to create a BAP.
- ► The eight losses are used to train the network concurrently. During evaluation, the eight vectors are averaged before a single BAP is created.



#### Current Progress

- 1. Rasterize protein DUD-E PDBs joined to ligand MOL2 files.
  - ▶ Done using Python's HTMD and MoleculeKit packages.
- 2. Channelize 3D grids.
  - Done using an offline implementation of Arpeggio's web service.
- 3. Implement DeepAtom architecture.
  - ► Shuffle groups expected to be a particular challenge.
- 4. Train, tune, and validate DeepAtom on NSCC.
  - ▶ I have more "low-level" control over this DeepAtom implementation than I do of those of DrugVQA and GGNN.
  - ▶ Need to accommodate response class imbalance.



#### For More Details...

#### DeepAtom:



Li, Y., Rezaei, M. A., Li, C., & Li, X. (2019)

DeepAtom: A framework FOR Protein-Ligand binding affinity prediction *IEEE/ACM Transactions on Computational Biology and Bioinformatics*.

#### PCMax algorithm/rasterization of atoms:



Jiménez, J., Doerr, S., Martínez-Rosell, G., Rose, A. S., & De Fabritiis, G. (2017) Deepsite: Protein-binding site predictor using 3d-convolutional neural networks

Deepsite: Protein-binding site predictor using 3d-convolutional neural networks

Bioinformatics 33(19), 3036-3042.

#### For More Details...

#### Arpeggio interatomic interaction finder:



Jubb, H., Higueruelo, A., Ochoa-Montaño, B., Pitt, W., Ascher, D., & Blundell, T. (2017)

Arpeggio: A web server for calculating and visualising interatomic interactions in protein structures

Journal of Molecular Biology 429(3), 365-371.

#### ShuffleNet:



Zhang, X., Zhou, X., Lin, M., & Sun, J. (2018)

Shufflenet: An extremely efficient convolutional neural network for mobile devices 2018 IEEE/CVF Conference on Computer Vision and Pattern Recognition.