



Article

# DockingApp RF: A State-of-the-Art Novel Scoring Function for Molecular Docking in a User-Friendly Interface to AutoDock Vina

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## 1. Supplementary Methods

### 1.1. Energy terms

The energy terms used for Docking App RF's features selection can be found in the following 3 Eq. 1, Eq. 2, Eq. 3 and Eq. 4. These equations are taken from [1] and calculated by using the option 4 score\_only of AutoDock Vina.

Gaussian terms:

$$\begin{aligned} Gauss_1 &= e^{-((d-o_1)/s_1)^2} \\ Gauss_2 &= e^{-((d-o_2)/s_2)^2} \end{aligned} \quad (1)$$

Repulsion term:

$$Repulsion(d) = \begin{cases} d^2, & \text{for } d \leq 0 \\ 0, & \text{for } d > 0 \end{cases} \quad (2)$$

Hydrogen bond term:

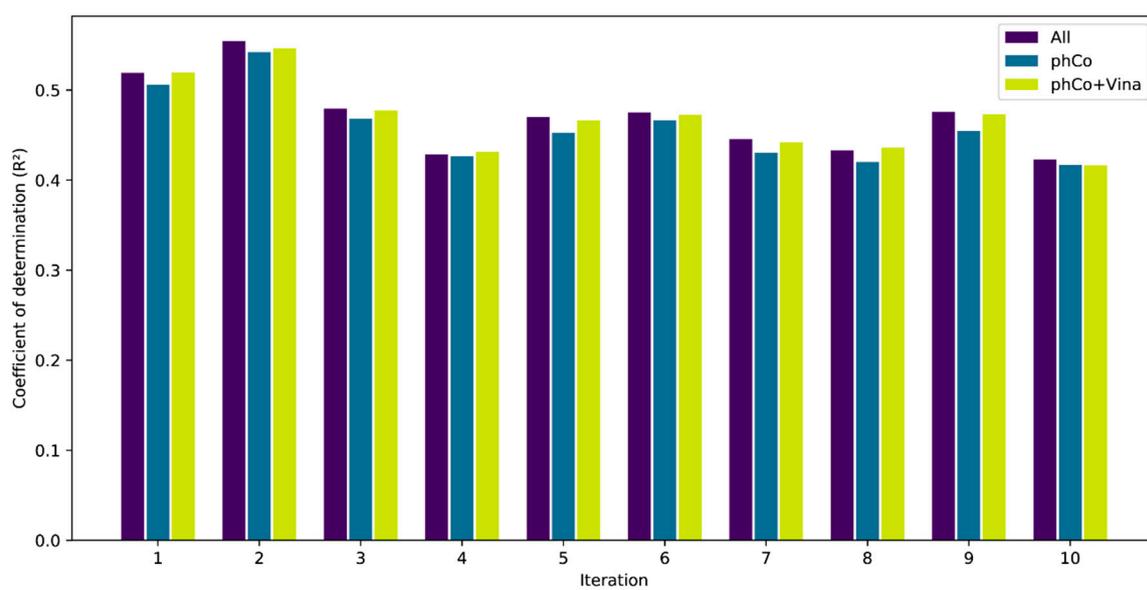
$$HBond(d) = \begin{cases} 1, & \text{for } d \leq h_1 \\ \frac{d}{-h_1}, & \text{for } h_1 < d < 0 \\ 0, & \text{for } d \geq 0 \end{cases} \quad (3)$$

Hydrophobic term:

$$Hydrophobic(d) = \begin{cases} 1, & \text{for } d \leq p_1 \\ p_2 - d, & \text{for } p_1 < d < p_2 \\ 0, & \text{for } d \geq p_2 \end{cases} \quad (4)$$

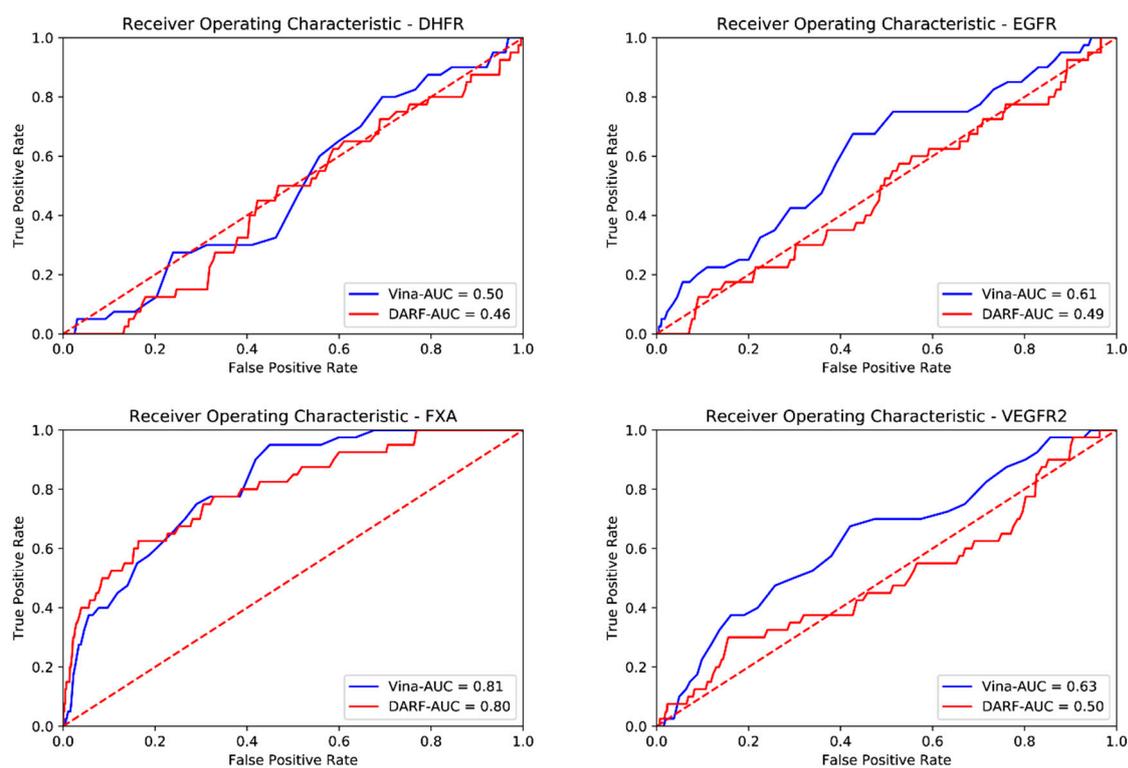
1. Trott, O.; Olson, A.J. AutoDock Vina: improving the speed and accuracy of docking with a new scoring function, efficient optimization, and multithreading. *J Comput Chem* **2010**, *31*, 455–61.

### 1.2. Ten-fold Cross-validation

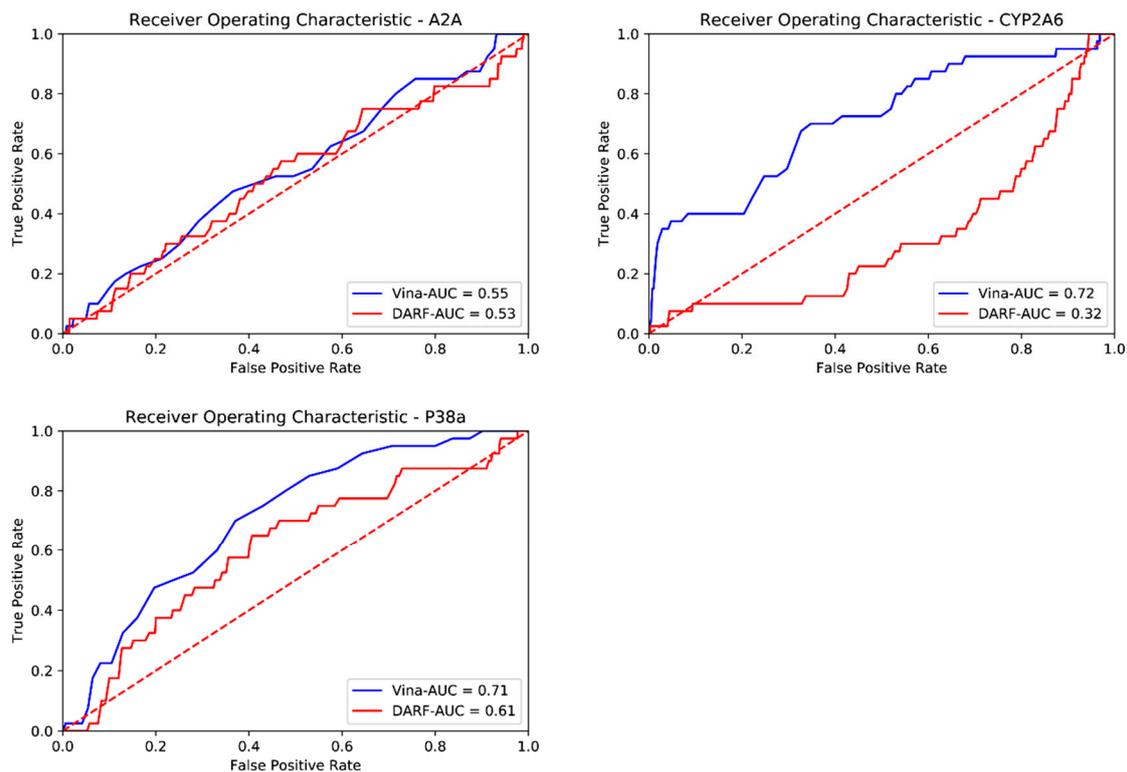


**Figure 1.** Results of the ten-fold cross-validation performed for three models trained with three different combinations of features: only intermolecular contacts (phCo, light blue), phCo plus Vina energy terms (green), phCo with Vina and solvent accessible surface area feature (dark blue, All).

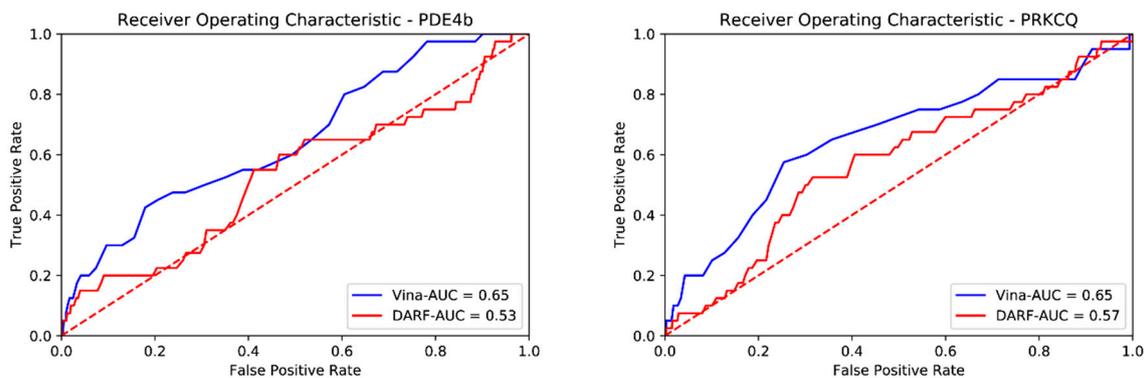
### 1.3. Docking power



**Figure 2.** ROC curve and AUC score for the four challenging complexes derived from DEKOIS2.0. The ROC curve for DockingApp RF (DARF) is depicted in red, while the one for Vina in blue.

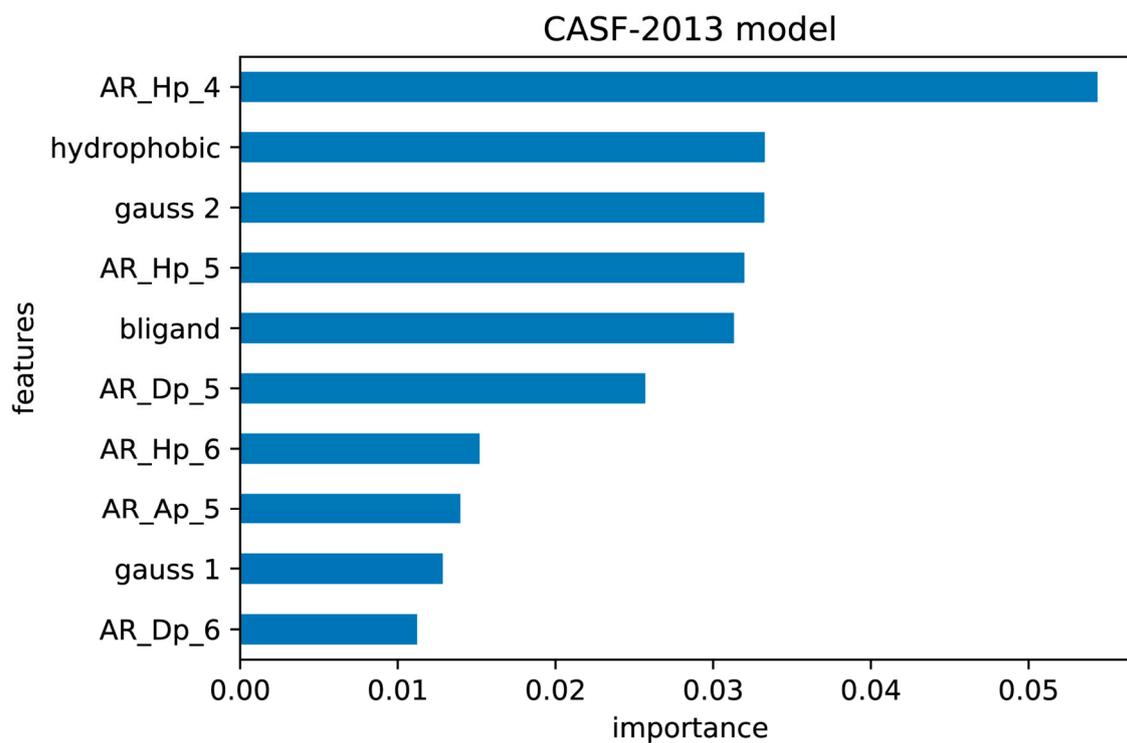


**Figure 3.** ROC curve and AUC score for the three moderately challenging complexes derived from DEKOIS2.0. The ROC curve for DockingApp RF (DARF) is depicted in red, while the one for Vina in blue.

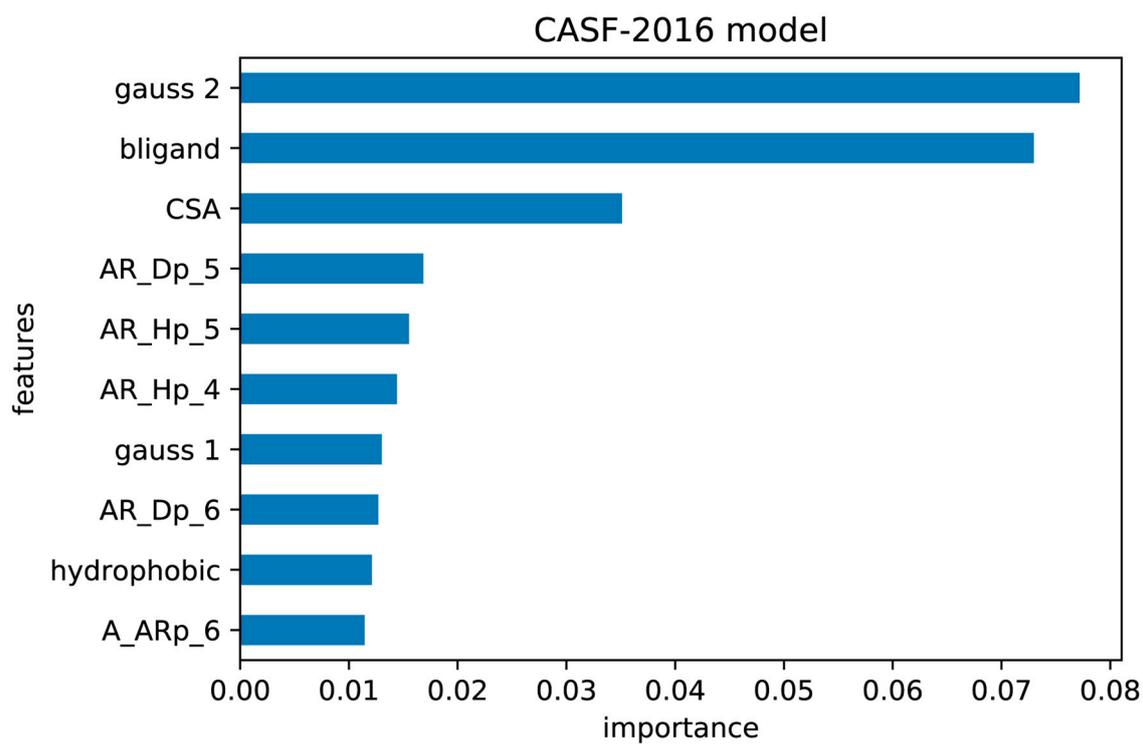


**Figure 4.** ROC curve and AUC score for the two less challenging complexes derived from DEKOIS2.0. The ROC curve for DockingApp RF (DARF) is depicted in red, while the one for Vina in blue.

## 1.4. Features importance

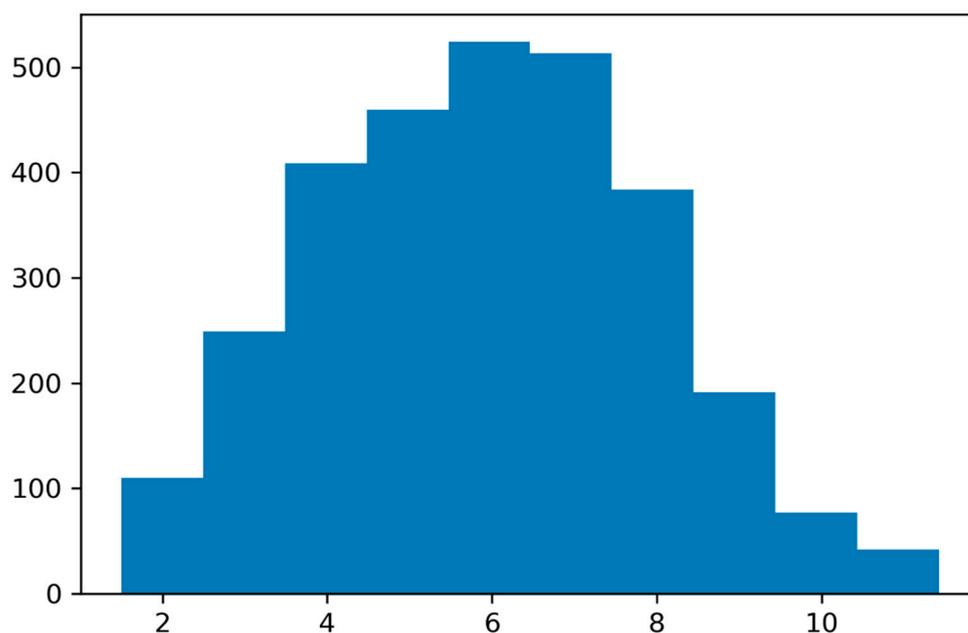


**Figure 5.** The ten most important features in the CASF-2013 model. PhCo features are presented with the format X\_Yp\_n, where X is the pharmacophoric feature calculated on a ligand atom, Y the pharmacophoric feature calculated on a protein atom, hence the p, and n is the shell at which the contact is registered.

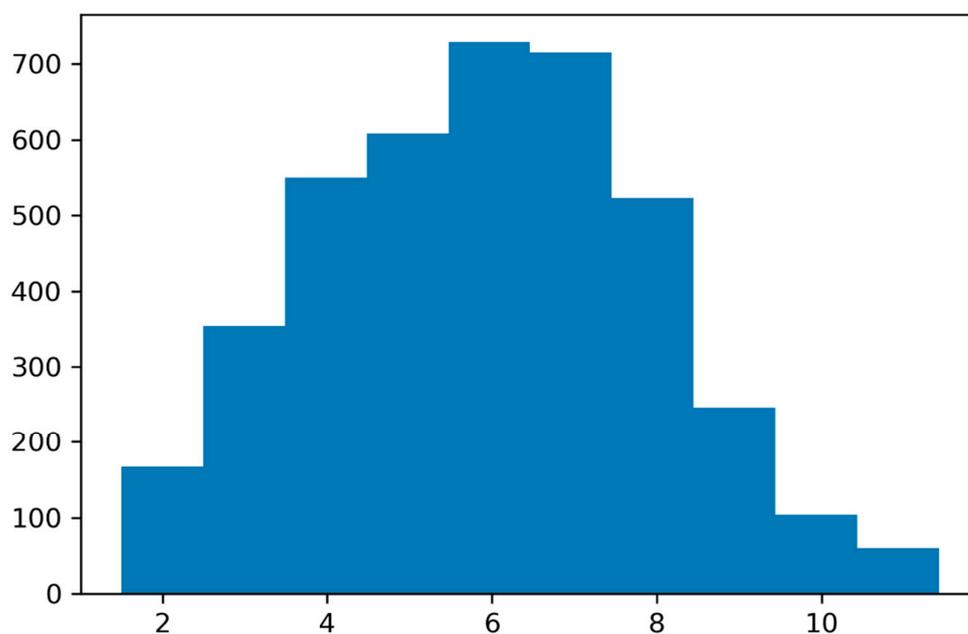


**Figure 6.** The ten most important features in the CASF-2016 model. PhCo features are presented with the format  $X\_Yp\_n$ , where  $X$  is the pharmacophoric feature calculated on a ligand atom,  $Y$  the pharmacophoric feature calculated on a protein atom, hence the  $p$ , and  $n$  is the shell at which the contact is registered.

## 1.5. Docking power



**Figure 7.** Distribution of the pKd values in the PDBBind2013 refined set.



**Figure 8.** Distribution of the pKd values in the PDBBind2016 refined set.

