

Identification of binding sites and favorable ligand binding
moieties by virtual screening and Self-Organizing Map
analysis

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Additional Tables

| Target | PDB entry | Cav number | Cav ids |
|--------|-----------|------------|------------------------|
| pa2ga | 1kvo | 8 | 1 2 3 5 14 15 16 19 30 |
| hmdh | 3ccw | 6 | 3 6 9 12 13 15 |
| braf | 3d4q | 5 | 2 3 4 5 6 |
| reni | 3g6z | 4 | 1 3 7 8 |
| prgr | 3kba | 4 | 1 2 3 4 |
| pgh2 | 3ln1 | 4 | 6 7 11 13 |
| glcm | 2v3f | 4 | 2 6 7 11 |
| esr2 | 2fsz | 4 | 3 8 9 11 |
| dpp4 | 2i78 | 4 | 15 17 20 24 |
| cxcr4 | 3odu | 4 | 3 4 6 9 |
| cp3a4 | 3nxu | 4 | 1 2 3 5 |
| mk01 | 2ojg | 3 | 2 4 5 |
| kpcb | 2i0e | 3 | 1 10 11 |
| kith | 2b8t | 3 | 1 8 9 |
| hivrt | 3lan | 3 | 2 3 6 |
| esr1 | 1sj0 | 3 | 1 4 5 |
| drd3 | 3pbl | 3 | 1 4 5 |
| cp2c9 | 1r9o | 3 | 3 5 6 |
| aofb | 1s3b | 3 | 2 5 11 |
| adrb1 | 2vt4 | 3 | 1 6 9 |
| aces | 1e66 | 3 | 2 3 7 |
| vgfr2 | 2p2i | 2 | 1 2 |
| thrb | 1ype | 2 | 2 3 |
| thb | 1q4x | 2 | 1 2 |
| tgfr1 | 3hmm | 2 | 1 2 |
| src | 3e18 | 2 | 5 6 |
| sahh | 1li4 | 2 | 6 8 |
| pyrd | 1d3g | 2 | 10 11 |
| pygm | 1c8k | 2 | 6 8 |
| pparg | 2gtk | 2 | 3 5 |
| ppard | 2znp | 2 | 1 10 |
| ppara | 2p54 | 2 | 1 2 |
| pgh1 | 2oyu | 2 | 2 3 |
| parp1 | 3l3m | 2 | 3 4 |
| nram | 1b9v | 2 | 1 3 |
| mcr | 2aa2 | 2 | 2 3 |
| lck | 2of2 | 2 | 3 6 |
| jak2 | 3lpb | 2 | 4 5 |
| inha | 2h7l | 2 | 1 6 |
| gria2 | 3kgc | 2 | 1 2 |
| gcr | 3bqd | 2 | 1 3 |
| fgfr1 | 3c4f | 2 | 3 4 |
| dhi1 | 3frj | 2 | 2 5 |
| bace1 | 3l5d | 2 | 1 2 |
| andr | 2am9 | 2 | 5 6 |
| ampc | 1l2s | 2 | 3 5 |
| adrb2 | 3ny8 | 2 | 1 2 |
| ace | 3bkl | 2 | 1 3 |
| abl1 | 2hzi | 2 | 1 6 |

Table S1 DUD-E targets sub-domains containing at least two cavities (detected with mkggrid) with a volume superior to 100 Å³. The last column contains the labels obtained with mkggrid for the detected cavities.

| | | | | | | | | |
|----------------------------------|----------------------------------|----------------------------------|--------------|--------------|-------|-------|--------------|--------------|
| HIV-RT | Cavity label | 1 | 2 | 3 | 4 | 5 | 6 | |
| | Cavity volume (\AA^3) | 15.4 | 338.5 | 957.4 | 83.8 | 34.8 | 294.9 | |
| | Neuron density (ADvina) | 0.260 | 3.070 | 0.976 | 0.0 | 0.0 | 0.058 | |
| | Neuron density (Dock) | 0.0 | 2.065 | 0.251 | 0.465 | 0.0 | 0.0 | |
| | Cavity label | 7 | 8 | 9 | | | | |
| | Cavity volume (\AA^3) | 15.4 | 15.4 | 15.4 | | | | |
| | Neuron density (ADvina) | 1.494 | 1.494 | 0.0 | | | | |
| | Neuron density (Dock) | 0.0 | 0.0 | 0.0 | | | | |
| | ABL1 | Cavity label | 1' | 2' | 3' | 4' | 5' | 6' |
| | | Cavity volume (\AA^3) | 257.1 | 15.4 | 52.8 | 46.8 | 15.1 | 615.5 |
| Neuron density (ADvina) | | 3.940 | 0.0 | 0.0 | 0.043 | 0.861 | 2.600 | |
| Neuron density (Dock) | | 6.410 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | |
| Cavity label | | 7' | 8' | 9' | 10' | 11' | 12' | |
| Cavity volume (\AA^3) | | 79.6 | 15.1 | 22.1 | 46.6 | 89.5 | 37.6 | |
| Neuron density (ADvina) | | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | |
| Neuron density (Dock) | | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | |

Table S2 Cavities detected by mgrid on HIV-RT and ABL1. Their label, volume and neuron density (neuron/ \AA^3) were calculated with the SOMs obtained with the EGF library. Cavities corresponding to AS and BS2 are in bold for each target.

| Method | Set size | Success rate (SR) | |
|--------------|----------|-------------------|------|
| | | Top3 | Top1 |
| SOM-BSfinder | 102 | 99% | 90% |
| FTsite | 48 | 98% | 94% |
| | 35 | 97% | 97% |
| Q-siteFinder | 134 | 86% | 80% |
| SiteHound | 77 | 95% | 77% |
| AutoLigand | 187 | – | 73% |

Table S3 Success rates for SOM-BSfinder and other energy-based algorithms when precision threshold is set to zero. These results are consistent with those obtained with various precision thresholds. SOM-BSfinder outperformed Q-siteFinder, SiteHound and AutoLigand, but presented lower success rates compared to FTSite.

| Target | SOM-BSfinder | SiteHound | FTSite |
|--------|--------------|-----------|--------|
| HIV-RT | 1 | 3 | 1 |
| ABL1 | 2 | 2 | 3 |
| REN1 | 1 | 2 | 1,2,3 |
| BRAF | 1 | 2,3 | 1 |
| HMDH | 1 | 6 | 1,2 |
| PA2GA | 1 | 1,2 | 1,2 |

Table S4 Active Site rank calculated for 6 targets in the DUD-E, chosen in different categories (table 1), using SOM-BSfinder, SiteHound and FTSite. With ABL1, SOM-BSfinder and SiteHound ranked the AS as the second position and FTSite ranked it in the third one. Otherwise, SOM-BSfinder ranked the AS as the first CC, which is not the case for FTSite and SiteHound. The latter algorithms showed more variability in the ranking.

| | E(EGFd) | E(AS) | $E(\overline{AS})$ | $E(AS)/E(EGFd)$ | $E(AS)/E(\overline{AS})$ |
|--------|---------|-------|--------------------|-----------------|--------------------------|
| HIV-RT | 0.08 | 0.22 | 0.05 | 2.85 | 4.64 |
| ABL1 | 0.06 | 0.18 | 0.04 | 2.87 | 4.44 |

Table S5 The enrichments in “active features” of the docked fragments ($E(EGFd)$), the AS ($E(AS)$) and the complementary of AS ($E(\overline{AS})$) for the test targets HIV-RT and ABL1.

| Target | Metric | Value | μ | σ | Z-score |
|--------|--------|-------|-------|-----------|---------|
| HIV-RT | Se | 0.49 | 0.17 | 10^{-2} | 23 |
| | Sp | 0.85 | 0.82 | 10^{-3} | 23 |
| ABL1 | Se | 0.46 | 0.16 | 10^{-2} | 20 |
| | Sp | 0.86 | 0.84 | 10^{-3} | 20 |

Table S6 Z-scores of the sensitivity (Se) and specificity (Sp) values obtained with the chemical features decomposition analysis for the test targets: HIV-RT and ABL1. For the sensitivity, we simulated a random sampling over features docking in the AS (F_{AS}) 1 million times. In a perfect scenario, all the active features (F_A) would dock in the AS, giving a sensitivity equal to 1. In the worst scenario, none of the active features would dock in the AS. The resulting samples were normally distributed $N(\mu, \sigma)$. The Z-score is the distance in terms of σ between the “experimental” value and the mean μ of the normal distribution. We did the same for the sensitivity, sampling randomly over features that would never dock in the AS ($\overline{F_{AS}}$).

Additional Figures



