

True landscape of drug-protein binding

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Abstract

“Docking” is a major tool in *in-silico* drug design and the method has been extensively developed. Simultaneously, computer hardware has been developed quite a lot. Nevertheless, poor performance of the docking methods are well known and its reason is not explained so far.

We have tried to understand the situation and reached a plausible explanation. The direct output of MO or MD calculation for a binding energy between drug and protein, is in a range of 40-60 kcal/mol. We subtract a constant to make calculated value comparable with experimentally observed value, 8-12 kcal/mol. On the other hand, drug activity measure, K_d , is defined as $kT \ln K_d = \Delta G$. And $\Delta G = \Delta H - T\Delta S$. MO and MD calculate only ΔH term and their output of binding energy either neglects $T\Delta S$ term or assumes $T\Delta S$ is similar for all compounds. The reason of the poor performance of docking methods seems to lie in this process.

Contribution of entropy must be significant if we consider released solvation water upon binding and restricted freedom of bound small molecules. We assume the observed binding energy of -9 kcal/mol must be the sum of -6 kcal/mol and -3 kcal/mol for ΔH and $-T\Delta S$, respectively. Although you can not judge 1 μ M activity based on calculated binding energy in a range over -6 kcal/mol, MO and MD programs output -9 kcal/mol after artificial adjustment. Consequently, the selected compounds by docking methods do not show activity in the most cases.

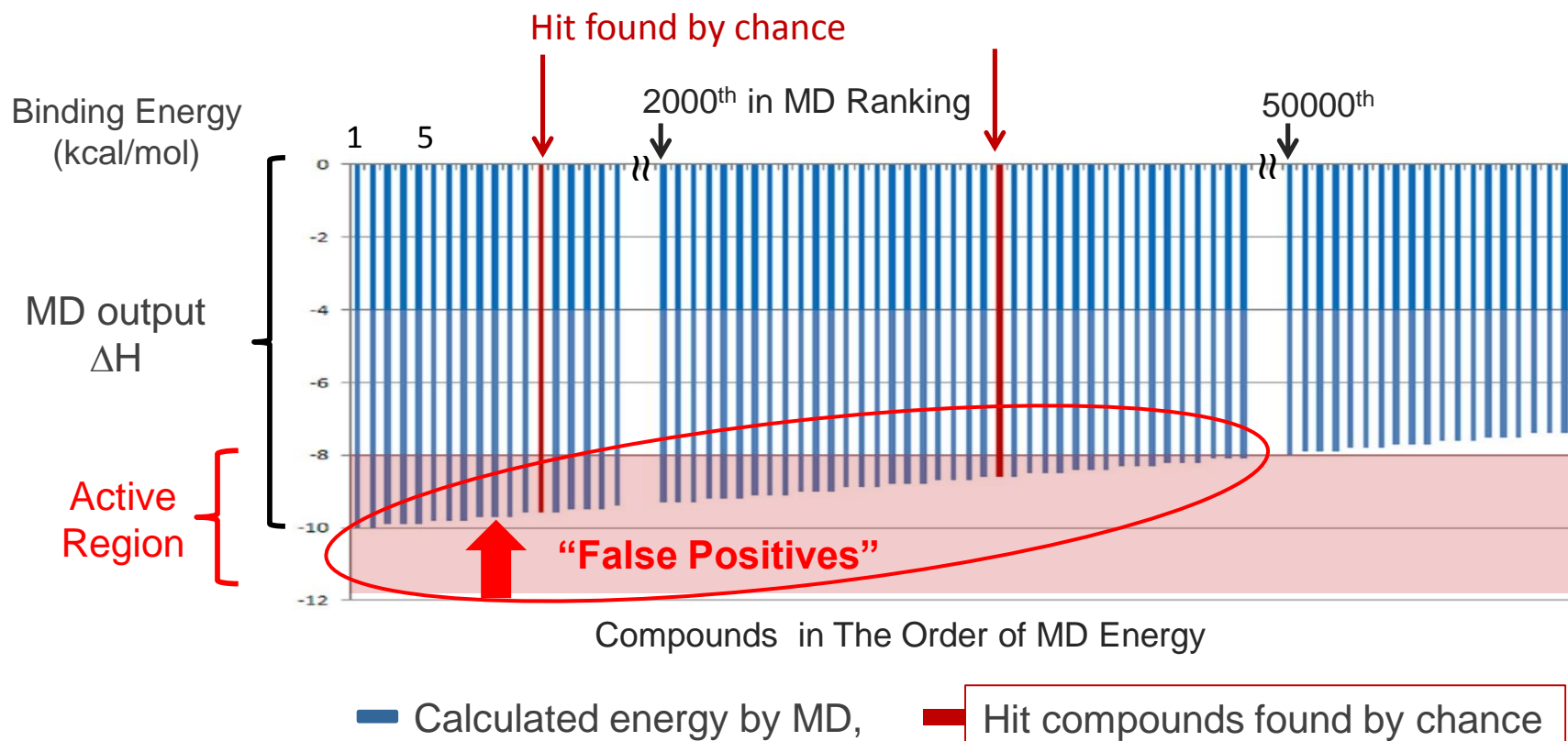
Interprotein has developed a system which consider entropy contribution and concentrates compounds with high total energy into higher rank. It has been successfully applied for 20 PPI targets and 10 enzyme targets. Details in Runx1-CBF β target is shown.

Conclusion

- The reason why elaborate calculations on best supercomputers do not succeed in finding active compounds is explained by the lack of considerations on “Entropy”.
- We assume correct active binding structures exist among top 50,000 computer output, but huge amount of false positives hide correct structures.
- Consideration on entropy will greatly help to find active structures and our INTENDD[®] system have continuous success on 10 enzymes and 20 PPI targets.
- Example on RUNX1-CBFb target is shown in detail.

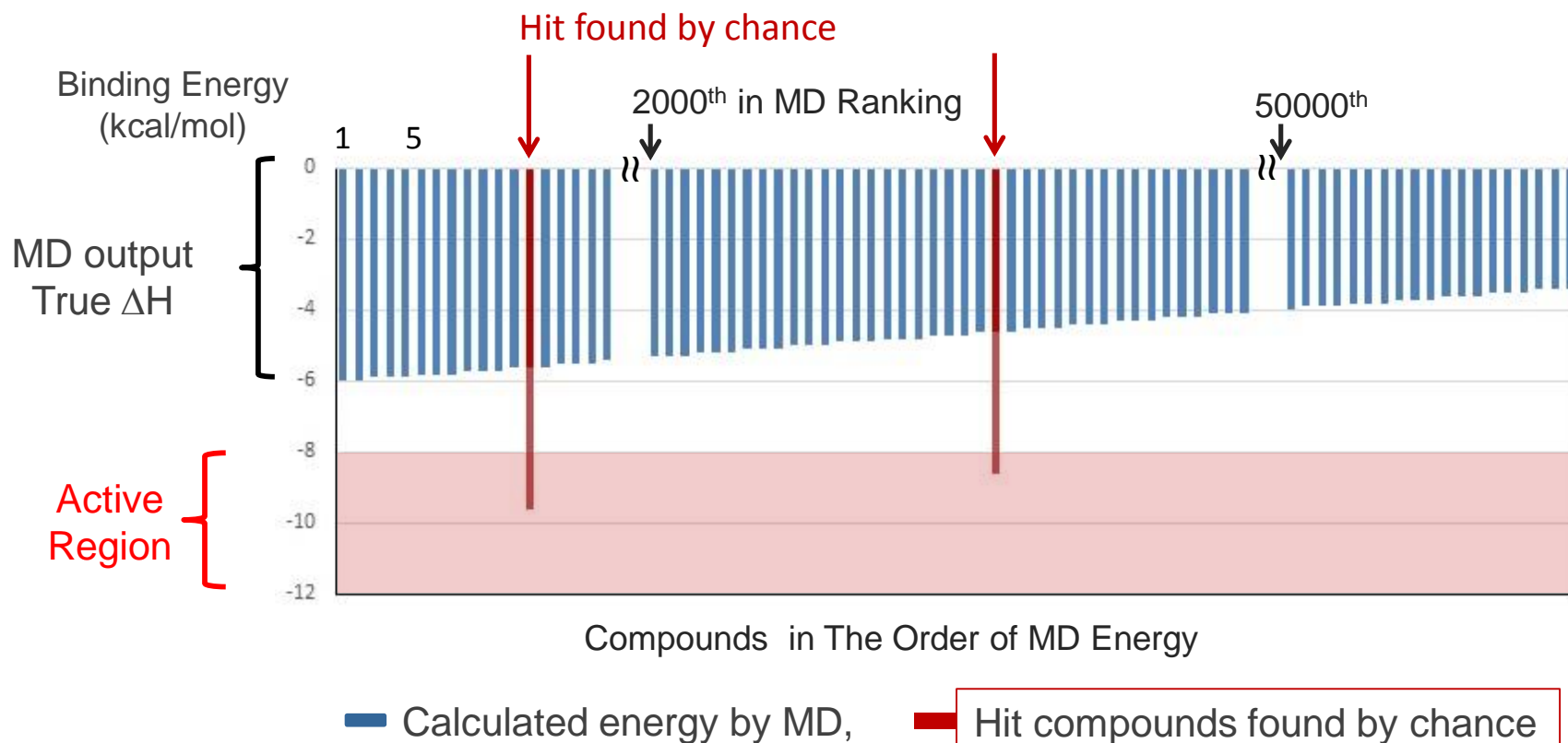
Docking problem | High false positive rate

In MD-based docking methods, binding energy is ΔH and most of the proposed compounds do not show activity. This leads to very high false-positive rate.



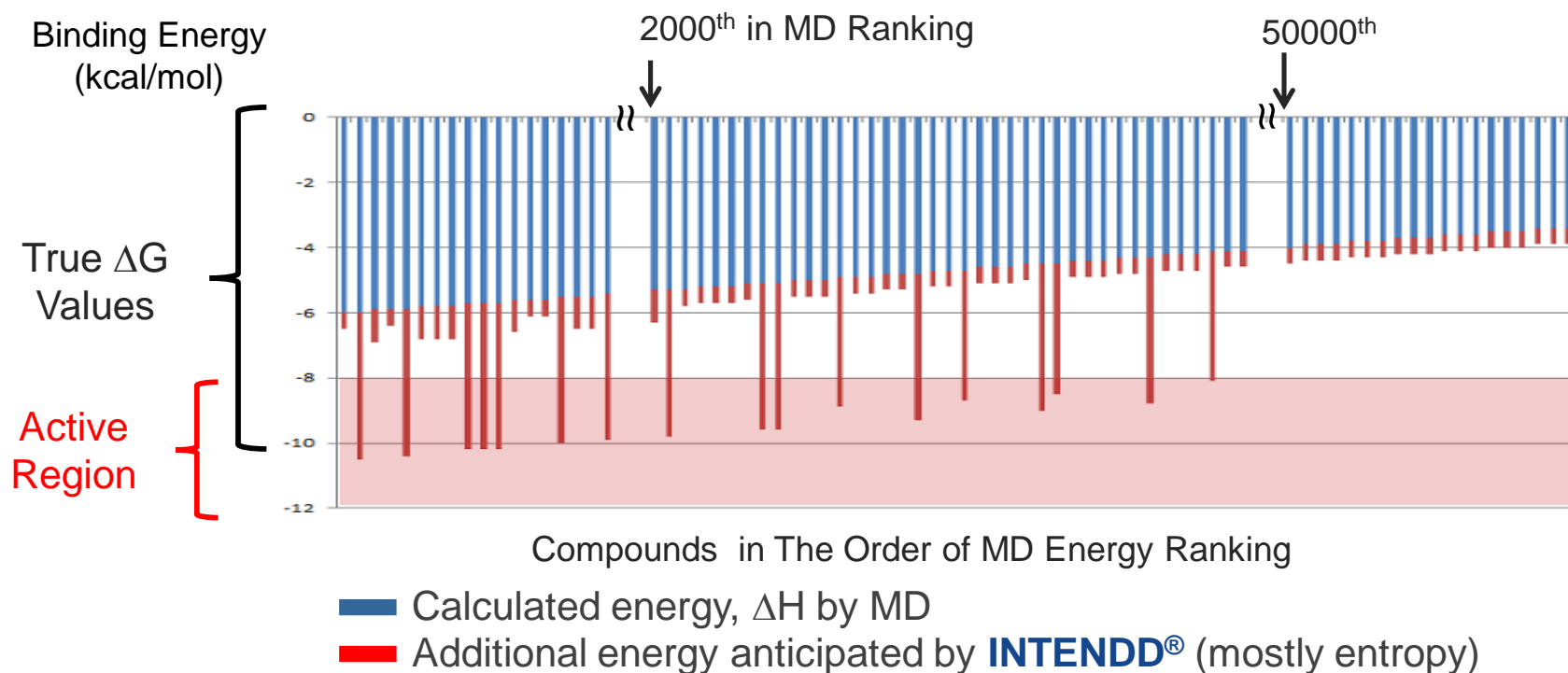
Docking problem | After removal of assumed $-T\Delta S$

In MD-based docking methods, binding energy, ΔH is adjusted to observed ΔG . When entropy has “Push” effect and subtracted from ΔH , remaining ΔH is small and can not be a measure of activity.



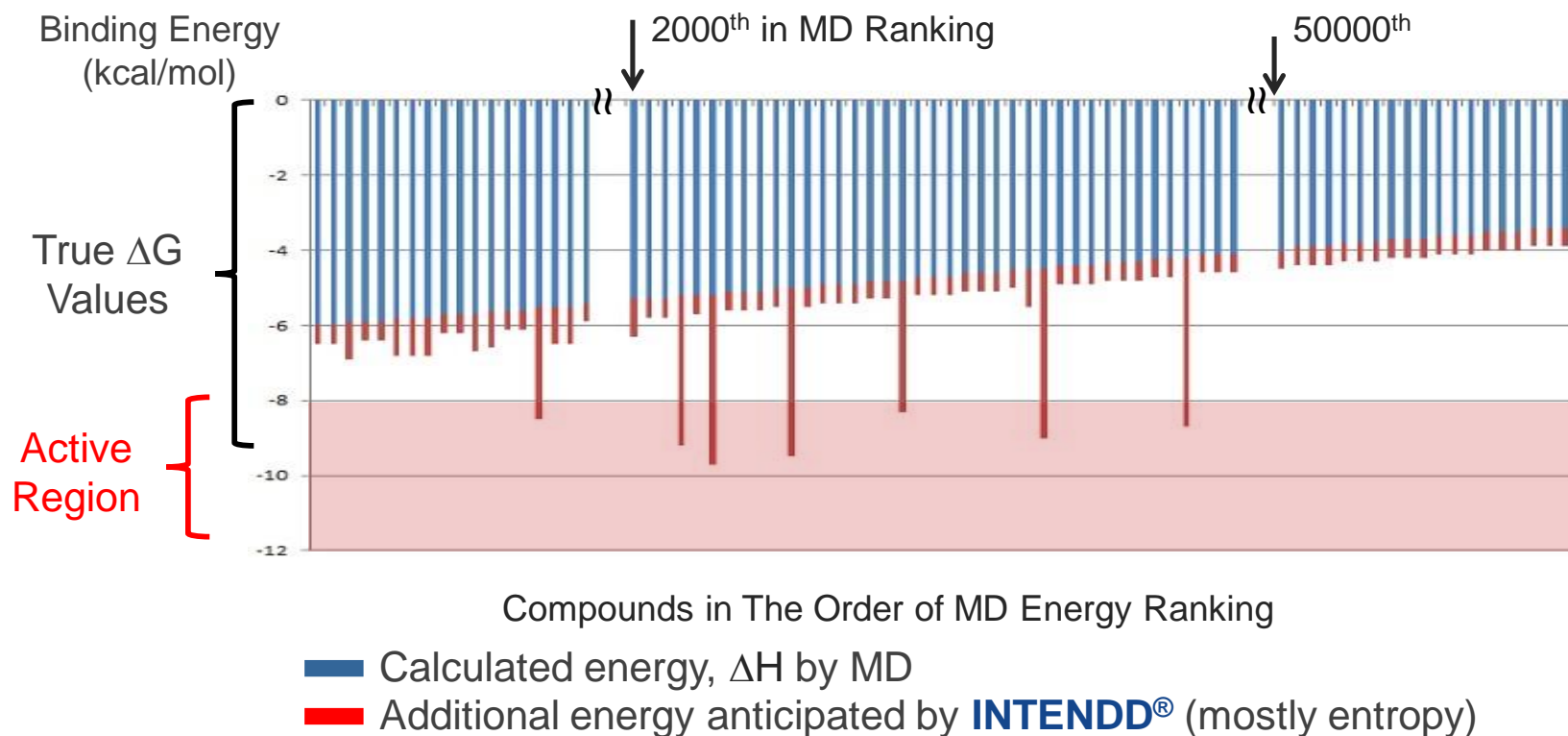
Kinase ATP sites are easy targets

ATP binding site of kinase is a flat pocket which accepts compounds with similar shape and size. Therefore, the contribution of entropy is similar among compounds. This is why the ΔH calculation can indicate hit compounds and population of hit in top group is high.



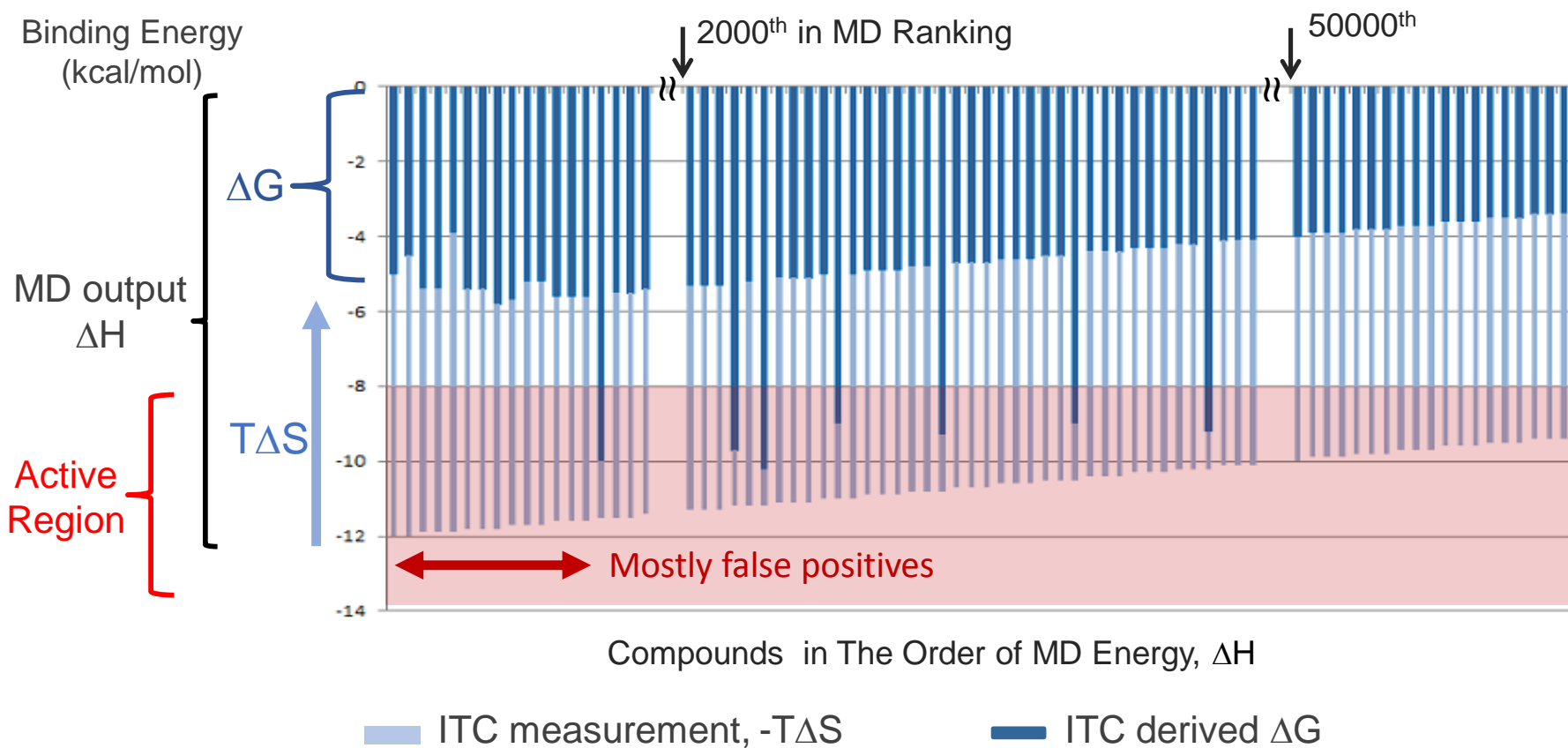
The real landscape of binding energy | ΔS “Push” case

- Activity Kd as $\Delta G = kT \ln Kd$
- Binding Energy $\Delta G = \Delta H - T\Delta S$
- Docking calculation can estimate most of ΔH .
- But ΔS contribution is more significant than was expected.



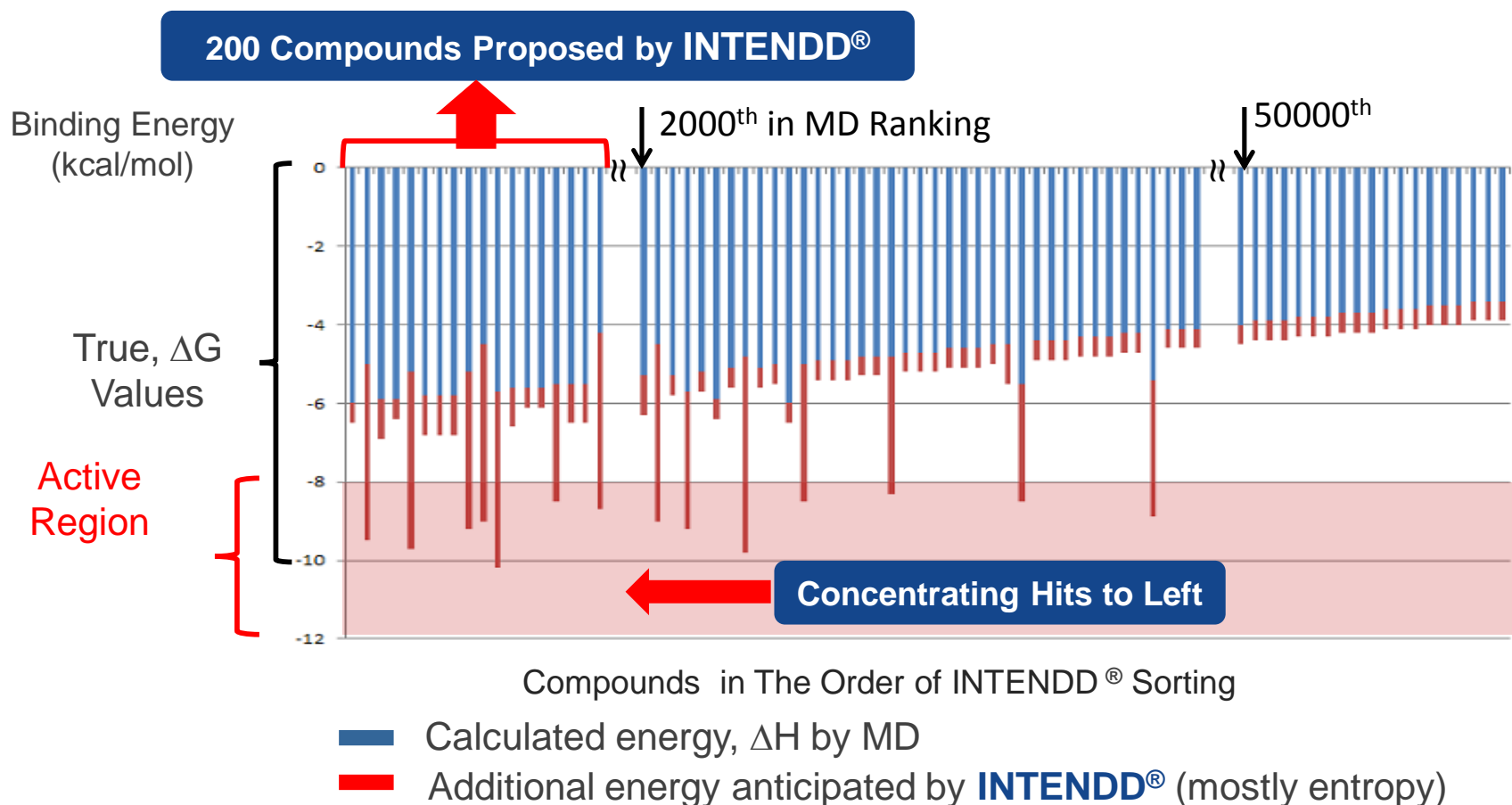
Alternative Case | Case of ΔS "Pull"

ITC measurements sometimes show negative entropy effect in drug-protein binding. Entropy effect is quite big and most compounds loose binding affinity because of decreased entropy.

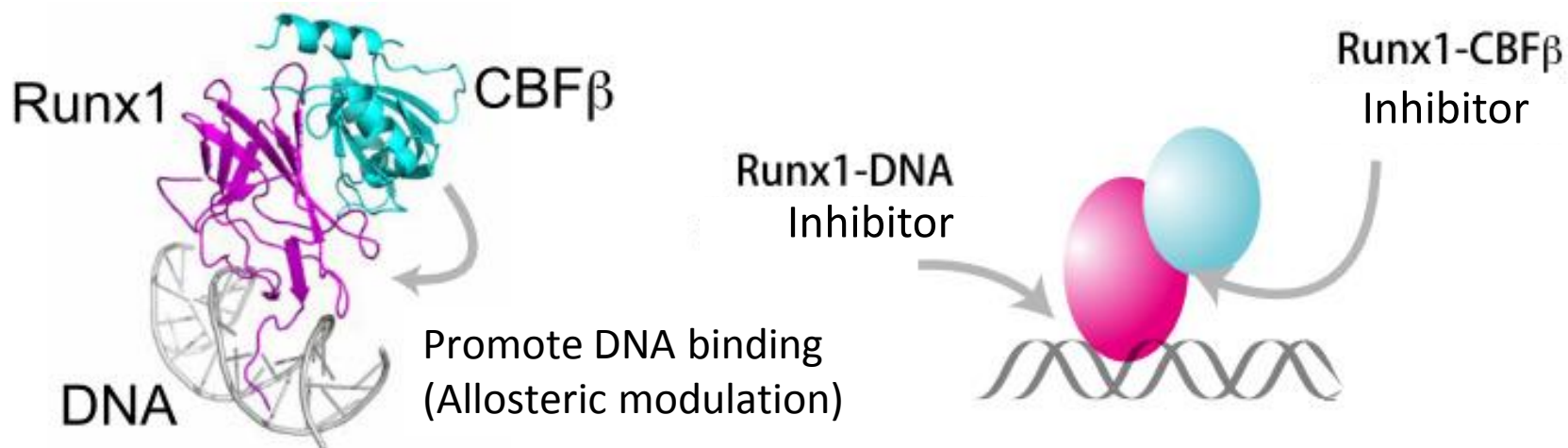


What our SBDD system, INTENDD[®] does.

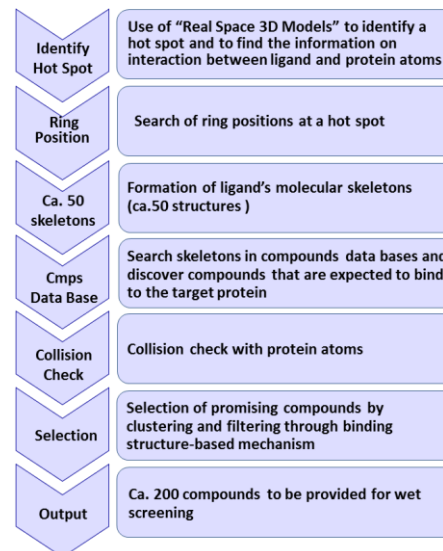
INTENDD[®] takes entropy into account, bringing promising compounds with big entropy to higher ranks.



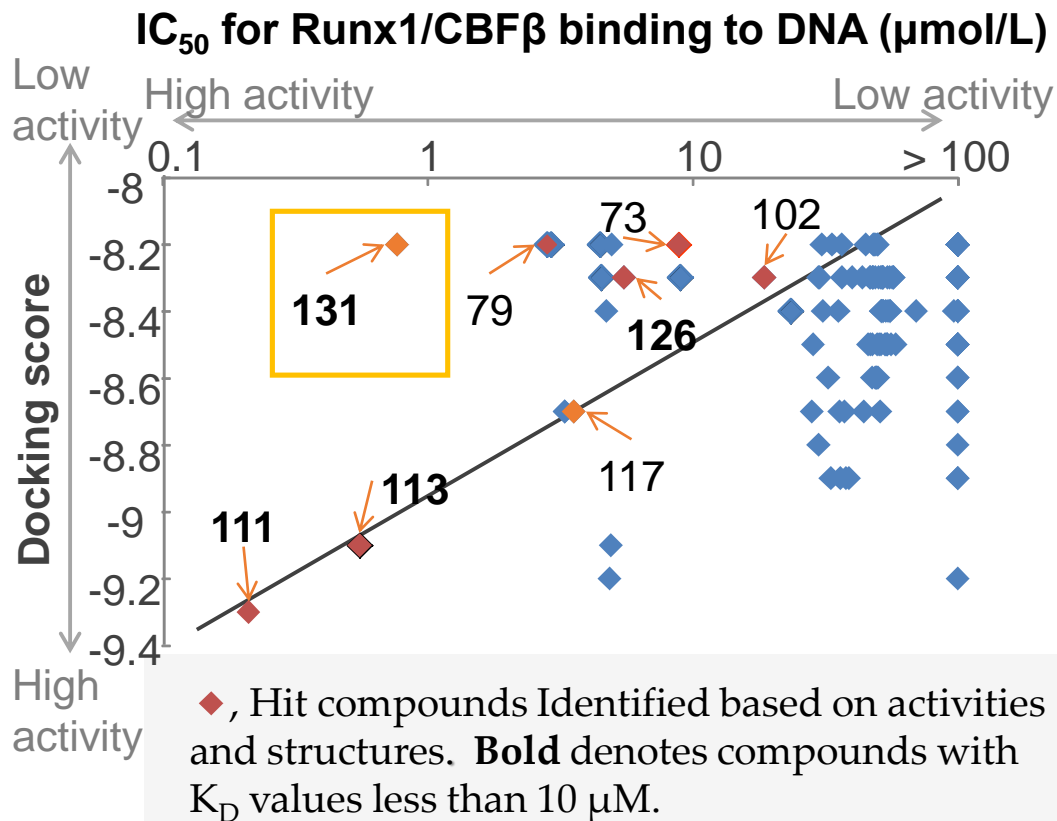
Strategy for Runx1 Inhibitor



Drug discovery for Runx1 Inhibitor by INTENDD[®]



Relationship between docking score and activity



Comps.	kcal/mol	IC ₅₀ (μM)	K _D (μM)
73	-8.2	8.9	ND
79	-8.2	2.9	35
102	-8.3	19	ND
111	-9.3	0.21	5.2
113	-9.1	0.55	1.6
117	-8.7	3.6	ND
126	-8.3	5.5	0.065
131	-8.2	0.77	9.5

ND: Not determined.

There is no linear relationship between MD-based docking score and activity. For instance, the lower right field contains comps with high docking scores and low activities, while the upper left field contains comps with low docking scores and high activities. **INTENDD**[®] can detect highly active comps even among comps with low docking scores (**131**). **131** analogs showed a good SAR, resulting in the production of many highly active comps ($\text{IC}_{50} < 1 \mu\text{M}$) in second screening.

Practical example | Runx1-CBF β interaction Inhibitor

Primary hits (IC₅₀ < 100 μ M)

Number of hit compounds that inhibit Runx1/CBF β -DNA binding over 50%

- 26 of 142* (hit rate: 18%) at the concentration of 100 μ M
- 7 of 142 (hit rate: 5%) at the concentration of 10 μ M
- 3 of 142 (hit rate: 2%) at the concentration of 1 μ M

* Number of compounds tested following proposal by INTENDD@SBSG®

- 1) Inhibition of Runx1/CBF β -DNA binding by compounds was assessed by SPR.
- 2) Binding affinity of compounds to Runx1 was determined by MST.
- 3) ND: not determined.

Compound	73	79	102	111	113	117	126	131
IC ₅₀ for Runx1/CBF β -DNA binding (μ M) ¹⁾	8.9	2.9	19	0.21	0.55	3.6	5.5	0.77
Binding affinity to Runx1(K _D , μ M) ²⁾	ND ³⁾	35	ND	5.2	1.6	ND	0.065	9.5

Number of tested compounds in secondary screening	1	0	2	2	6	10	17	35	Total
Number of secondary hit compounds	1	-	0	0	6	6	1	20	34
Hit rate in secondary screening (%)	100	-	0	0	100	60	5.9	57	46

Secondary hits (IC₅₀ < 10 μ M)

