

Volgograd State Medical University
Research Center of Innovative Medicines
Laboratory for Information Technology in Pharmacology
and Computer Modeling of Drugs



MULTI-TARGET NEURAL NETWORK MODEL OF ANXIOLYTIC ACTIVITY OF CHEMICAL COMPOUNDS BASED ON CORRELATION CONVOLUTION OF ENERGY SPECTRA OF MULTIPLE DOCKING

Vassiliev Pavel Mikhailovich

Research Center of Innovative Medicines



WHO – Anxiety disorders

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Anxiety disorders

27 September 2023

301 million people

Key facts

- Anxiety disorders are the world's most common mental disorders, affecting 301 million people in 2019.
- More women are affected by anxiety disorders than men.
- Symptoms of anxiety often have onset during childhood or adolescence.
- There are highly effective treatments for anxiety disorders.
- Approximately 1 in 4 people with anxiety disorders receive treatment for this condition.

Overview

Everyone can feel anxious sometimes, but people with anxiety disorders often experience fear and worry that is both intense and excessive. These feelings are typically accompanied by physical tension and other behavioural and cognitive symptoms. They are difficult to control, cause significant distress and can last a long time if untreated. Anxiety disorders interfere with daily activities and can impair a person's family, social and school or working life.

An estimated 4% of the global population currently experience an anxiety disorder (1). In 2019, 301 million people in the world had an anxiety disorder, making anxiety disorders the most common of all mental disorders (1).

Although highly effective treatments for anxiety disorders exist, only about 1 in 4 people in need (27.6%) receive any treatment (2). Barriers to care include lack of awareness that this is a treatable health condition, lack of investment in mental health services, lack of trained health care providers, and social stigma.

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Related

[Comprehensive mental health action plan 2013–2030](#) >

[Mental Health Gap Action Programme \(mhGAP\)](#) >

[More on mental health](#) >

Fact sheets



Mental disorders

8 June 2022



Depressive disorders 3

Purpose

To construct a multi-target model
of the dependence of the
anxiolytic activity of chemical
compounds on their multiple
docking energies using
correlation convolution and
artificial neural networks

Tasks

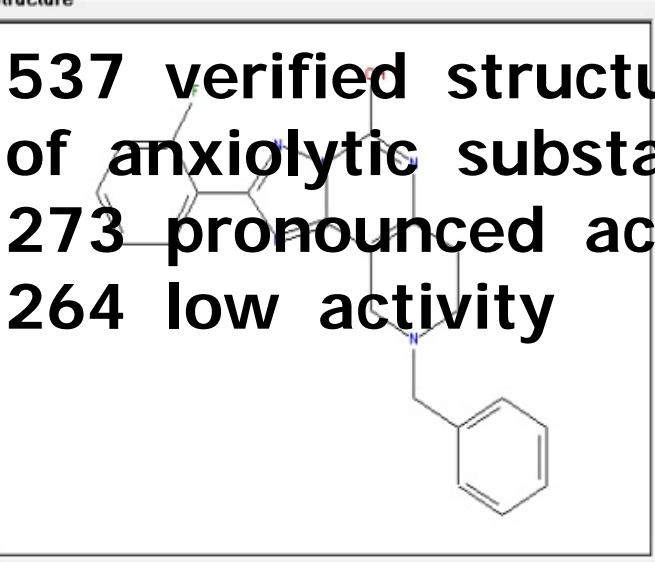
- Construction of optimized 3D models of ligands
- Selection and validation of relevant biotargets
- Formation of spaces for multiple docking
- Performing multiple ensemble docking to relevant biotargets
- Correlation convolution of multiple docking energy spectra
- Formation of training set
- Training of neural networks and selection of the best model

Verified database of anxiolytic substances

[Anxiolytic Substances v11 - Corrected](#)

Structure

537 verified structures
of anxiolytic substances:
273 pronounced activity
264 low activity



24 January 2022

Activity

Standard Type

ED50

pChEMBL Value

1

Assay

Assay ChEMBL ID

CHEMBL779993

Assay Description

Anxiolytic response measured to stimuli in rats.

Mol_ID	Brutto formula	Mol weight	Salt Component
1	C ₂₁ H ₁₈ FN ₅ O	375.3	

Compound Codes & Names

MolFilename	CODE - Molecule ChEMBL ID
CHEMBL99259	CHEMBL99259
Molecule Name	Compound Key
	8g

References	Document_ChEMBL_ID	Document_Journal	Document_Year
	CHEMBL1125833	J. Med. Chem.	1991

15.07.2022

РОССИЙСКАЯ ФЕДЕРАЦИЯ



СВИДЕТЕЛЬСТВО

о государственной регистрации базы данных

№ 2022621744

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Авторы: Васильев Павел Михайлович (RU), Мальцев Дмитрий Васильевич (RU), Перфильев Максим Алексеевич (RU), Спасов Александр Алексеевич (RU), Скрипка Мария Олеговна (RU), Кочетков Андрей Николаевич (RU)

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Руководитель Федеральной службы
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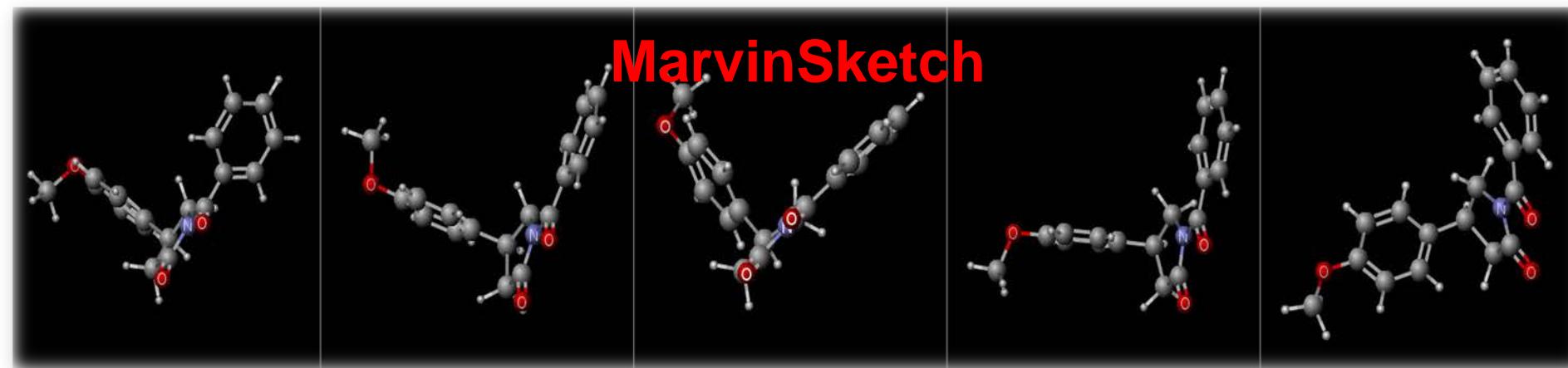
Ю.С. Зубов



Construction of optimized 3D models of ligands

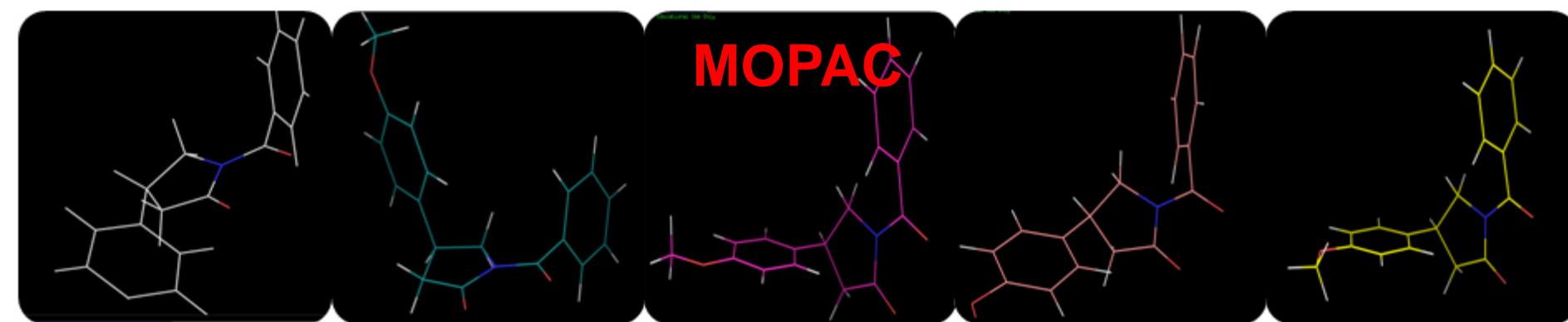
Molecular mechanics 10 conformers

MarvinSketch

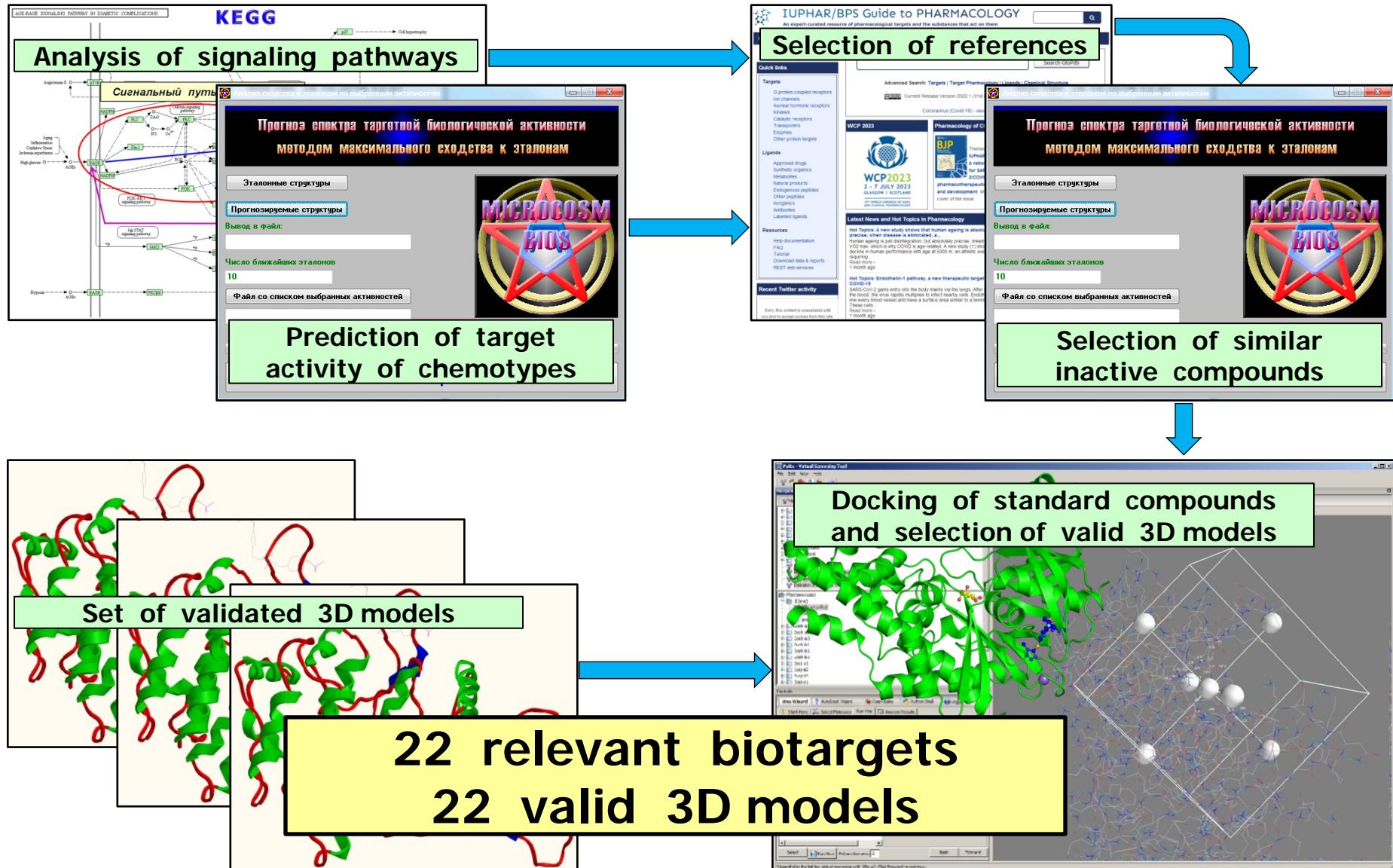


Quantum chemistry PM7

MOPAC

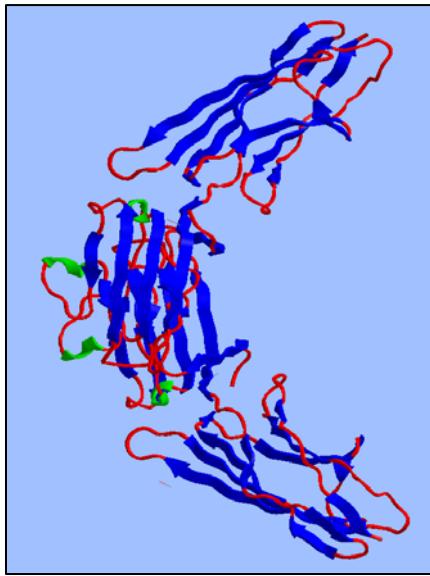


Selection and validation of relevant biotargets



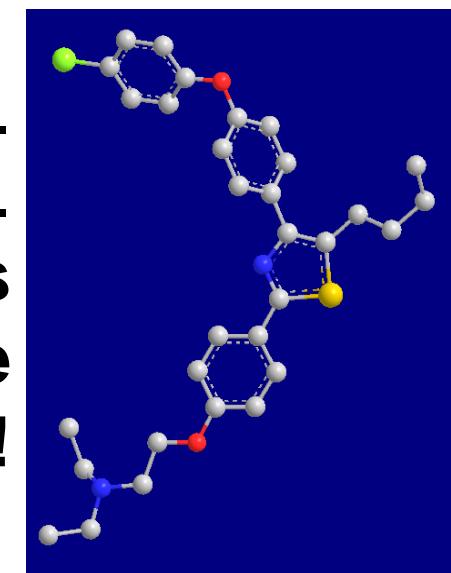
Multiple docking

Ligands interact with the protein throughout its entire volume

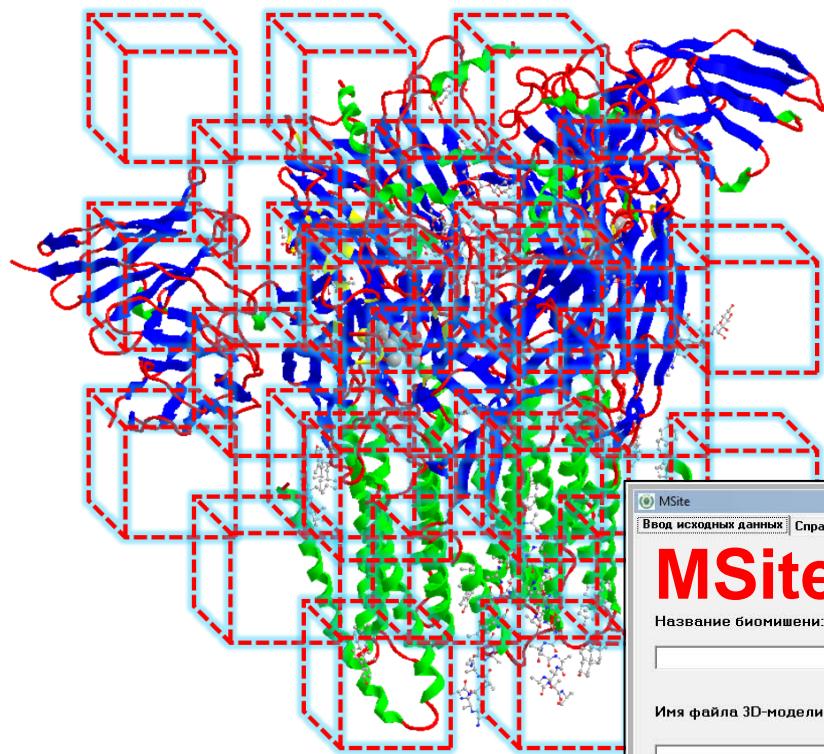


- Active concentration 10^{-10} M.
Avogadro's number $\sim 6 \cdot 10^{23}$.
 **$\sim 6 \cdot 10^{13}$ ligand molecules
interact with the target protein!**

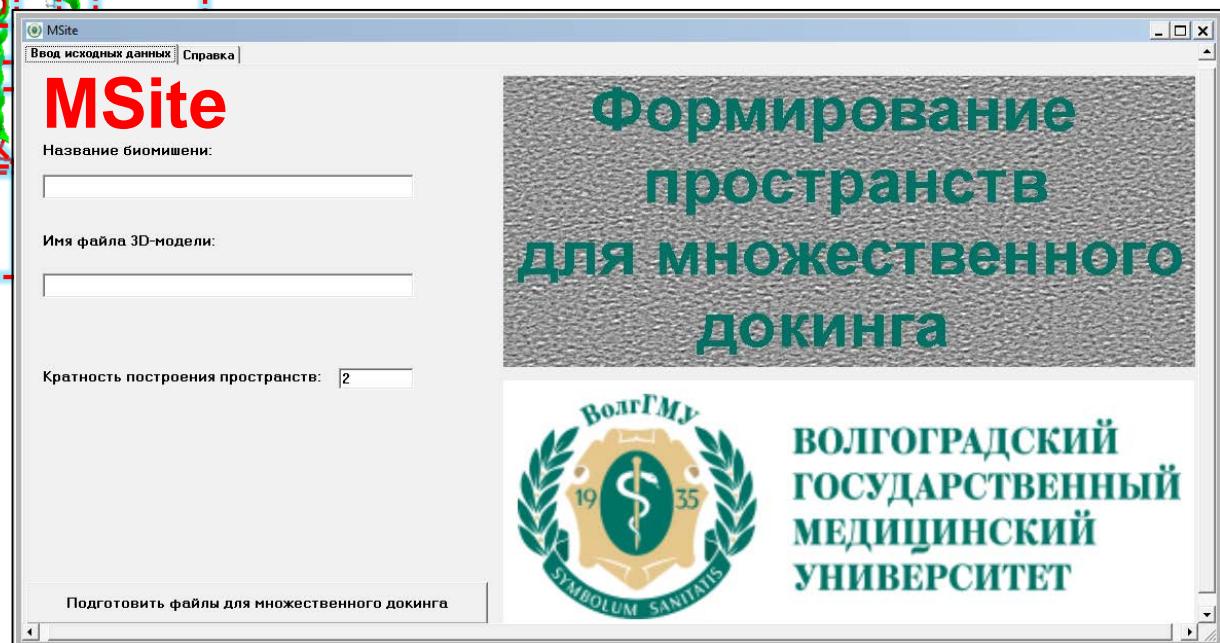
- RAGE volume $34\ 902\ \text{\AA}^3$.
Ligand volume $1\ 578\ \text{\AA}^3$.
**The ligand fits
into the receptor space
at least 22 times!**



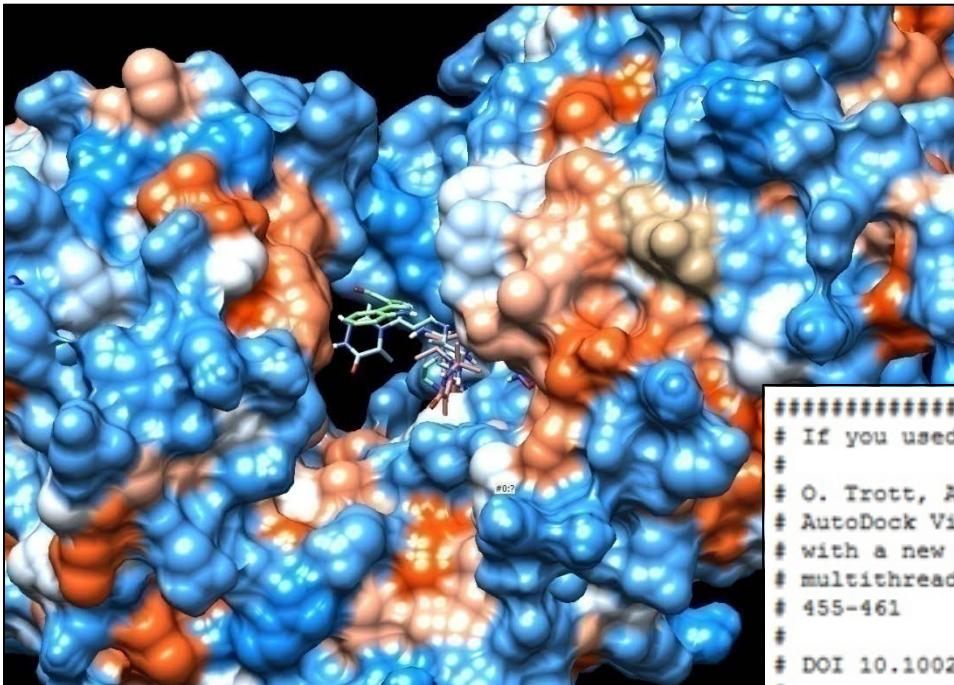
Formation of spaces for multiple docking



$$3 \cdot 3 \cdot 3 = 27 \text{ spaces}$$



Multiple ensemble docking to relevant biotargets



Each of 537 ligands
to each of 22 3D models
to each of 27 spaces
5 times in 10 conformations

$$\Delta E_{ijkl} = \min(\Delta E_{ijklm})$$

AutoDock Vina

```
#####
# If you used AutoDock Vina, please cite:
#
# O. Trott, A. J. Olson,
# AutoDock Vina: improving the speed and accuracy of docking
# with a new scoring function, efficient optimization and
# multithreading, Journal of Computational Chemistry 31 (2010)
# 455-461
#
# DOI 10.1002/jcc.21334
#
# Please see http://vina.scripps.edu for more information.
#####
Output will be ...\\Ligands\\Flavopiridol-PM7_out.pdbqt
Detected 12 CPUs
WARNING: at low exhaustiveness, it may be impossible to utilize all CPUs
Reading input ... done.
Setting up the scoring function ... done.
Analyzing the binding site ... done.
Using random seed: 58910256
Performing search ...
0% 10 20 30 40 50 60 70 80 90 100%
|----|----|----|----|----|----|----|----|----|
*****
```

Total processed
 $537 \cdot 22 \cdot 27 \cdot 5 \cdot 10$
 $\sim 16\,000\,000$
docking energies

Correlation convolution

$$R_{ijkl} = \frac{Cov_{ijkl}}{S_{ikl} \cdot S_{jkl}}, \quad i, j = 1 \dots M, \quad k = 1 \dots K, \quad l = 1 \dots L, \quad i \neq j$$

$$W_{kn} = \frac{1}{2} \sum_{\substack{l=1 \\ i \neq j}}^L \sum_{j=1}^M R_{ijkl} \cdot X_{ilkn} \cdot X_{jlkn}, \quad k = 1 \dots K, \quad n = 1 \dots N$$

$K = 22$, number of biotargets

$M = 27$, number of docking spaces

$L = 5$, number of docking energies for space i

$N = 537$, number of substances

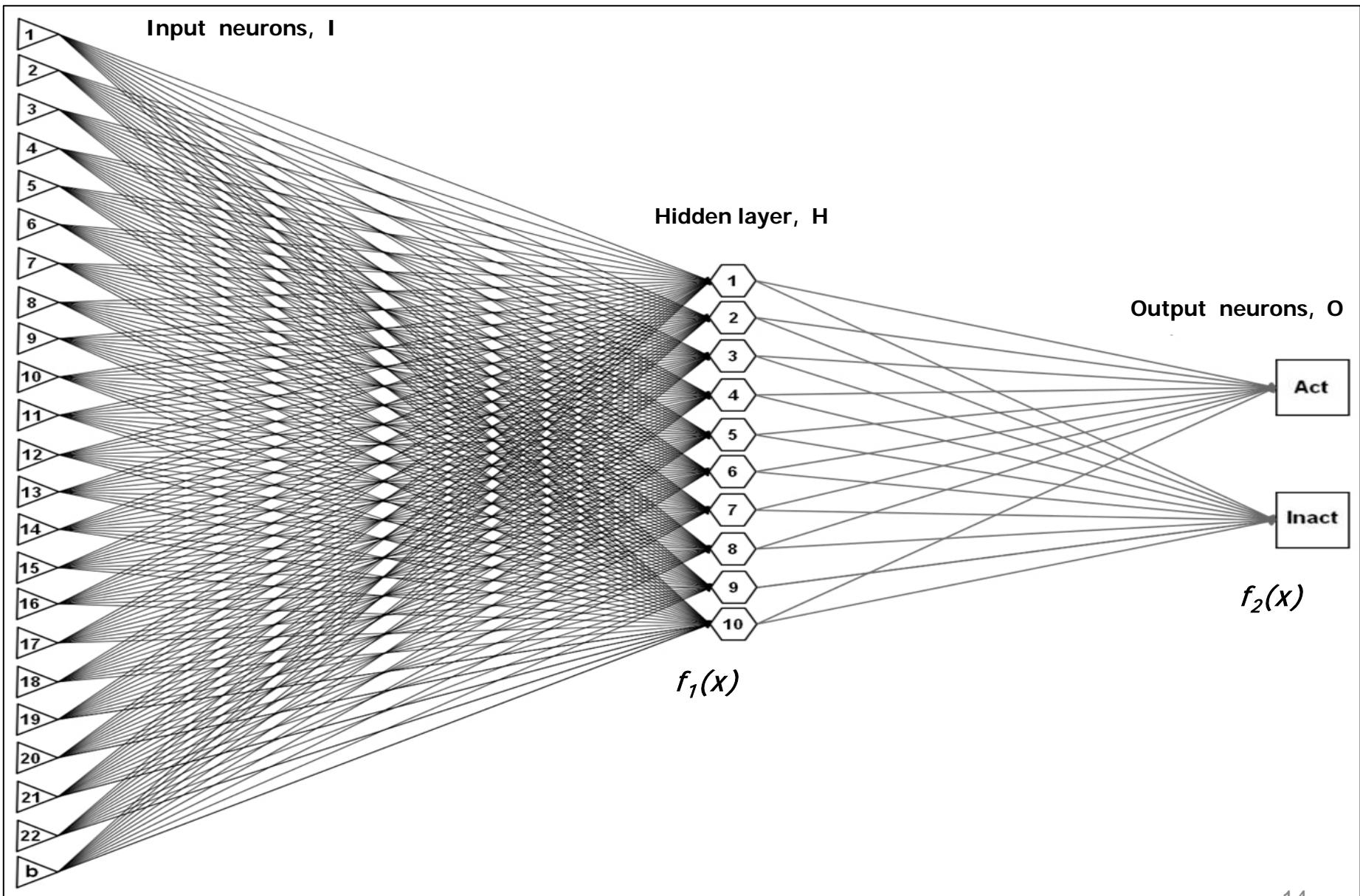
2970 docking energies —→ 22 convolution variables

Formation of training set

**537 Anxiolytics
273 with pronounced activity
264 with low activity
22 neurons
7 samplings**

Mol_ID	Code	LevHM	ADRA1A	ADRA1B	ADRA2A	ADRA2B	AGTR1	CA2	CA4	CNR1	GABAR	HTR1A	HTR1B	HTR1D	HTR2A	HTR2B	HTR2C	HTR4	HTR7	MTNR1A	MTNR1B	NMDA	SCN11A	SLC18A2	Samp1	Samp2	Samp3
1	CHEMBL99259	hm	16278	18509	17455	18446	18109	14903	12669	14127	20663	21214	16198	17404	19442	13662	19454	19865	15295	10879	18473	31531	22906	16151	train	val	test
2	CHEMBL513136	hm	14868	17099	14794	16137	15451	12078	10294	11251	16248	17422	12848	16467	18570	11539	16997	18216	11970	9607	15880	27271	20472	14342	train	train	val
3	CHEMBL3410223	hm	11852	13939	12749	14939	13927	11120	9701	9533	15033	15999	12326	13836	15856	10189	15055	16350	11869	8939	14991	27412	17231	13713	train	train	train
4	CHEMBL323519	hm	10932	11218	10318	12985	11799	10191	8659	8351	12785	15856	10832	12169	13155	8758	12770	13362	10077	7212	13512	21909	14430	11048	train	train	train
5	CHEMBL204240	hm	10088	11098	9952	11353	10711	9024	7117	7003	12571	12052	9311	17787	17133	7117	11372	2116	8943	7737	11629	19264	13112	10969	train	train	train
6	CHEMBL314608	hm	10056	11154	9986	11380	10779	9169	7195	7854	13112	12811	9983	12124	2316	1134	1185	13294	18753	7276	11685	21543	14357	10778	test	train	train
7	CHEMBL319178	hm	15755	17820	16585	17774	17640	14260	12273	15713	17336	20111	13467	16211	16545	15234	16073	15045	14862	10543	17765	30412	21758	15589	val	test	train
8	CHEMBL101045	hm	10159	11086	9850	12585	11113	10085	8762	8396	12079	15169	10256	11146	12901	8242	11878	11981	9819	7311	13254	20097	13973	11041	train	val	test
9	CHEMBL543621	hm	10644	10914	10052	12555	11158	9763	8392	8134	11871	14655	10046	10965	12494	8124	12326	12538	9293	7226	12988	19797	13608	10626	train	train	val
10	CHEMBL545731	hm	10239	10862	9665	12236	11070	9481	8237	8191	11803	14189	10021	11317	12112	7838	11833	12787	9229	7184	12139	19905	13106	10489	train	train	train
11	CHEMBL3409256	hm	1593	17118	15835	18932	1771	1338	1322	12425	20048	20790	15207	17841	19084	12438	19261	1753	14037	10866	15923	3229	2026	17211	train	train	train
12	CHEMBL153823	hm	12078	12285	11147	11744	1164	1155	9547	10747	1607	1137	1155	11337	1166	8980	11418	1290	10229	1684	12192	1993	14338	12414	train	train	train
13	CHEMBL466847	hm	14466	16420	14651	16115	15429	12512	10774	1516	14609	16862	2582	18320	16822	11238	15936	16170	12035	9410	16380	25673	100	14420	test	train	train
14	CHEMBL3084529	hm	13034	14415	13602	14811	15314	11688	9927	9810	15329	17164	11916	14686	16647	10625	14668	15677	11395	9014	14895	27452	20032	13602	val	test	train
15	CHEMBL3410224	hm	11747	14147	12607	15214	13653	11026	9401	9638	14955	16242	12059	13196	15475	9764	14485	16342	12004	8434	14444	25387	16959	12735	train	val	test
16	CHEMBL13662	hm	15243	15965	14150	15769	14374	12930	12157	11423	18667	17019	13803	15380	16634	11023	16270	17534	12992	11708	15470	26886	17962	15776	train	train	val
17	CHEMBL3246317	hm	13229	15026	13081	12668	13212	12142	10783	9920	1453	14471	1781	14655	15052	10221	14063	1616	10954	10387	14236	23029	17118	12373	train	train	train
18	CHEMBL3808873	hm	9778	11786	10670	11718	1159	8466	7160	7179	986	11727	1022	1019	11450	8425	1672	1357	170	6935	1590	20562	14043	9865	train	train	train
19	CHEMBL67100	hm	12898	14435	12205	14818	13432	10616	9347	10598	16075	15552	1987	14316	14250	10559	14184	15553	11253	9401	13684	24600	17597	14736	train	train	train
20	CHEMBL157522	hm	14875	15179	13821	15268	13823	12489	11467	11140	19337	16193	13134	14992	15574	10707	15799	16502	12868	10911	14782	26470	17931	14736	test	train	train
21	CHEMBL540812	hm	10418	11050	9974	12446	11001	9614	8231	8011	12255	14812	10343	11392	12327	8338	12007	12520	9604	6964	12904	20891	14109	10443	val	test	train
22	CHEMBL98258	hm	15543	18148	17201	18217	18314	14526	12590	13336	20207	20255	16127	17094	19634	13950	18494	19099	14831	10679	18614	31388	22355	16279	train	val	test
23	CHEMBL155270	hm	12207	12152	10890	12752	11726	10204	9411	9194	16741	13263	10848	12896	12850	8975	12968	13210	10370	9190	12525	21024	15422	12080	train	train	val
24	CHEMBL88606	hm	10904	11639	10782	11558	11549	10478	9167	8262	12927	1317	1338	1158	1708	1151	12677	9644	8762	13171	21299	14364	10199	train	train	train	
25	CHEMBL86611	hm	9245	10110	9300	10752	10161	8736	724	72	12049	1815	9327	1158	1897	1081	1159	12534	8604	6933	11065	18859	12761	10199	train	train	train
26	CHEMBL357362	hm	11888	13830	12086	14470	12832	11833	10451	9959	17209	16123	12118	13284	15113	10041	14893	16124	11065	9496	14883	23998	16409	13176	train	train	train
27	CHEMBL44515	hm	13115	15980	14560	14938	14428	12015	10593	10639	15365	17236	13396	16130	16277	10324	16437	19481	11785	9385	14946	27084	18447	13196	test	train	train
28	CHEMBL378298	hm	13736	14746	13252	15864	14743	12150	10803	11017	16776	18649	13469	15770	16682	11151	16054	16636	12515	9686	16004	27446	18379	14476	val	test	train
29	CHEMBL562219	hm	10807	12231	10826	11644	11287	9476	9099	8656	12461	13105	10539	11632	12787	8955	12724	13639	9546	8246	11916	21203	14752	11753	train	val	test
30	CHEMBL260870	hm	14191	15930	13129	15583	14738	11745	1073	11618	13787	1178	1380	119	1683	1034	1255	18123	12350	9663	16311	26784	18556	13936	train	train	val
31	CHEMBL87790	hm	8778	9286	8355	9454	9233	8059	7399	7357	1041	1042	852	542	1050	7348	103	11351	7764	6528	10139	16602	11650	9388	train	train	train
32	CHEMBL188335	hm	12826	15199	13889	15822	14908	11537	10414	11191	19263	17702	12889	15232	16305	1117	15839	16327	12212	9253	16770	26560	18048	14796	train	train	train
33	CHEMBL431298	hm	12756	15059	14106	15543	14609	12406	10098	10429	16614	18177	12861	13703	16421	10492	15973	17116	12296	8615	16389	28327	18777	13971	train	train	train
34	CHEMBL559437	hm	10026	11027	10221	10872	10559	8713	8216	7604	12631	12727	9603	11018	12122	8002	11896	13215	9096	7567	11865	18760	13289	11295	test	train	train
35	CHEMBL319966	hm	14176	16072	13920	16272	15452	12668	11624	11571	20377	17932	13269	15785	17368	11357	16155	17789	13109	10878	16436	28435	19776	15236	val	test	train
36	CHEMBL41	hm	11803	13027	11635	12743	12449	9943	9154	9423	14441	13549	10609	12520	13443	9602	13234	13866	10005	9400	11920	22693	15355	11640	train	val	test
37	CHEMBL40906	hm	13625	16599	13508	14176	14077	12267	10561	10253	14907	16017	13311	15028	16724	10310	14670	18695	11431	10325	15014	24632	17792	13498	train	train	val
38	CHEMBL465161	hm	13384	14197	10658	14399	13165	13360	10355	9366	14232	16762	11595	14843	16323	10218	14080	13511	11278	8549	15991	25592	19747	12489	train	train	train
39	CHEMBL88106	hm	9256	10248	9442	10774	10380	8757	7842	7792	12106	11711	9318	10561	11011	8012	11212	12535	8663	6888	10887	18117	12618	10224	train	train	train
40	CHEMBL86557	hm	9941	11026	10154	11463	10714	9409	8235	8262	12016	12569	9724	10956	11653	8050	12106	13468	9160	7486	11480	19973	13184	10885	train	train	train
41	CHEMBL70418	hm	12224	14630	11327	11652	11920	10445	9220	8870	13129	12974	10889	12727	14353	9722	13603	15299	10070	8867	12759	21937	15306	11389	test	train	train
42	CHEMBL2263293	hm	12840	13703	12143	15298	13616	12201	11071	10624	16386	17167	12739	14183	15288	10895	14828	15345	12110	8730	15122	24634	18509	13511	val	test	train
43	CHEMBL190772	hm	12598	14141	12959	15600	14644	11760	10286	10687	17909	17352	12803	14546	16289	10470	16731	15714	11558	9252	16395	25860	18291	13749	train	val	test
44	CHEMBL20042	hm	13304	15150	13958	15757	14863	12837	11193	11143</																	

Neural network with bottle-neck



Training of neural networks

SANN - Results: NetWorks RAGE-NFKB v02

Active neural networks

7 sampling options

4000 trained neural networks for each sampling option

50 automatically selected neural networks for each sampling option

Neural network training in progress...

Building network 145 (MLP 22-16-2, exp, logistic)

Cycle=40: 1 best neural network

Classification rate: Train=85.034, Test=75

~30 000 networks were trained

The screenshot shows the SANN software interface. At the top, there's a table titled 'Active neural networks' with columns for Net. ..., Net. name, Training ..., Test p..., Algorit..., Error fu..., Hidden ..., Output ... and rows for 6 different networks. Below this, a large text box displays the progress of building network 145, mentioning a classification rate of 85.034 for training and 75 for testing. A progress dialog box is overlaid on the interface, showing 'Neural network training in progress...', 'Building network 145 (MLP 22-16-2, exp, logistic)', 'Cycle=40: 1 best neural network', and 'Classification rate: Train=85.034, Test=75'. The dialog has buttons for Cancel, Next, and Finish. On the right side of the interface, there are tabs for Ensemble, Standalone and ensemble, Predictions, Targets, Standard res., Confidence Int., Accuracy, Variables, and Options. Under Options, there are checkboxes for Train, Test, Validation, and Missing.

Accuracy of the best neural network

Mol_ID	Code	Sampl1	F	F(h)	Sampl2	F	F(h)	Sampl3	F	F(h)	Sampl4	F	F(h)	Sampl5	F	F(h)	Sampl6	F	F(h)	Sampl7	F
1	CHEMBL99259	hm	0.985	0.985	hm	1.000	1.000	hm	0.808	0.808	hm	0.725	0.725	hm	0.596	0.596	hm	1.000	1.000	hm	1.000
2	CHEMBL513136	nhm	0.605	0.395	hm	0.504	0.504	hm	0.844	0.844	nhm	0.984	0.016	hm	0.938	0.938	hm	0.868	0.868	hm	1.000
3	CHEMBL3410223	hm	0.889	0.889	hm	0.808	0.808	hm	1.000	1.000	hm	1.000	1.000	hm	0.975	0.975	hm	1.000	1.000	hm	0.954
4	CHEMBL323519	hm	0.968	0.968	hm	0.816	0.816	hm	1.000	1.000	hm	1.000	1.000	hm	0.857	0.857	hm	0.878	0.878	hm	1.000
5	CHEMBL204240	nhm	0.413	0.47	nhm	0.90	0.46	nhm	0.859	0.41	hm	0.857	0.857	hm	0.72	0.225	nhm	0.938	0.062	nhm	1.000
6	CHEMBL314608	hm	1.000	1.000	hm	0.93	0.938	hm	1.000	1.000	hm	1.000	1.000	hm	0.830	0.67	hm	0.949	0.949	hm	0.954
7	CHEMBL319178	nhm	1.000	0.000	hm	1.000	1.000	hm	0.842	0.842	hm	0.780	0.780	hm	0.630	0.630	hm	1.000	1.000	hm	1.000
8	CHEMBL101045	hm	0.916	0.916	hm	0.572	0.572	hm	1.000	1.000	hm	1.000	1.000	hm	0.702	0.702	hm	0.538	0.588	hm	1.000
9	CHEMBL543621	hm	0.990	0.990	hm	0.649	0.649	hm	0.707	0.707	hm	1.000	1.000	hm	0.609	0.609	hm	0.699	0.699	hm	1.000
10	CHEMBL545731	hm	0.998	0.998	hm	0.721	0.721	hm	1.000	1.000	nhm	1.000	0.000	hm	0.783	0.783	hm	0.664	0.664	hm	1.000
11	CHEMBL3409256	hm	1.000	1.000	hm	0.535	0.535	hm	1.000	1.000	hm	0.954									
12	CHEMBL153823	hm	1.000	1.000	hm	0.510	0.510	hm	1.000	1.000	hm	1.000	1.000	hm	0.617	0.617	hm	0.823	0.823	hm	0.954
13	CHEMBL466847	nhm	1.000	0.000	hm	0.564	0.564	hm	1.000	1.000	hm	1.000	1.000	hm	0.633	0.633	hm	0.860	0.860	hm	1.000
14	CHEMBL3084529	hm	0.742	0.742	hm	0.765	0.765	hm	1.000	1.000	hm	0.991	0.991	hm	0.664	0.664	hm	1.000	1.000	hm	1.000
15	CHEMBL3410224	hm	0.734	0.734	hm	0.930	0.930	hm	1.000	1.000	hm	1.000	1.000	nhm	0.596	0.404	hm	0.901	0.901	hm	0.954
16	CHEMBL13662	hm	0.850	0.850	nhm	0.602	0.398	hm	0.656	0.656	nhm	0.908	0.092	nhm	0.739	0.261	hm	1.000	1.000	hm	0.954
17	CHEMBL3246317	hm	0.978	0.978	hm	0.961	0.961	hm	1.000	1.000	nhm	1.000	0.000	hm	0.937	0.937	hm	0.912	0.912	hm	0.954
18	CHEMBL3808873	hm	0.957	0.957	hm	1.000	1.000	hm	0.997	0.997	hm	1.000	1.000	hm	0.911	0.911	hm	0.974	0.974	hm	1.000
19	CHEMBL67100	hm	0.884	0.834	hm	0.30	0.30	hm	0.390	0.95	hm	1.000	1.000	hm	0.945	0.945	nhm	1.000	0.000	hm	0.954
20	CHEMBL157522	hm	0.836	0.836	nhm	0.334	0.335	hm	0.015	0.655	hm	0.987	0.58	nhm	0.714	0.286	hm	1.000	1.000	nhm	1.000
21	CHEMBL540812	hm	1.000	1.000	hm	0.603	0.603	hm	1.000	1.000	hm	0.961	0.961	hm	0.771	0.771	nhm	0.516	0.484	hm	0.954
22	CHEMBL98258	hm	1.000	1.000	hm	1.000	1.000	hm	0.845	0.845	hm	0.897	0.897	hm	0.862	0.862	hm	1.000	1.000	hm	1.000
23	CHEMBL155270	hm	1.000	1.000	hm	0.651	0.345	nhm	0.00	0.000	hm	1.000	1.000	hm	0.515	0.515	hm	0.656	0.656	hm	0.940
24	CHEMBL68606	nhm	0.680	0.320	hm	0.010	0.010	hm	1.000	1.000	nhm	0.952	0.018	hm	0.936	0.936	hm	0.837	0.837	hm	1.000
27	CHEMBL86611	hm	0.999	0.999	hm	0.959	0.959	hm	1.000	1.000	hm	1.000	1.000	hm	0.993	0.993	hm	1.000	1.000	hm	1.000
28	CHEMBL357362	hm	1.000	1.000	hm	0.761	0.761	hm	0.879	0.879	hm	0.655	0.655	hm	0.895	0.895	hm	0.600	0.600	hm	0.954
25	CHEMBL44515	hm	0.751	0.751	hm	0.874	0.874	hm	1.000	1.000	hm	0.998	0.998	hm	0.899	0.899	hm	0.898	0.898	hm	0.929
26	CHEMBL378298	hm	0.888	0.838	nhm	0.573	0.32	hm	0.153	0.65	hm	0.910	0.911	hm	0.880	0.880	hm	1.000	1.000	hm	0.954
30	CHEMBL562219	hm	1.000	1.000	hm	0.00	0.00	hm	1.000	1.000	hm	1.000	1.000	hm	0.984	0.984	hm	0.998	0.998	hm	0.954
31	CHEMBL260870	hm	0.767	0.767	nhm	0.755	0.245	hm	0.656	0.656	nhm	0.922	0.078	nhm	0.634	0.366	hm	1.000	1.000	hm	0.954
32	CHEMBL87790	nhm	0.649	0.361	hm	0.986	0.986	hm	1.000	1.000	hm	0.690	0.690	hm	0.573	0.573	hm	0.992	0.992	hm	1.000
33	CHEMBL188335	hm	0.892	0.892	hm	1.000	1.000	hm	0.846	0.846	hm	0.962	0.963	hm	0.990	0.990	hm	1.000	1.000	hm	0.954
34	CHEMBL431298	hm	0.691	0.691	hm	1.000	1.000	hm	0.841	0.841	hm	1.000	1.000	nhm	0.561	0.439	hm	1.000	1.000	nhm	1.000
35	CHEMBL559437	hm	1.000	1.000	hm	0.778	0.778	hm	1.000	1.000	hm	1.000	1.000	hm	0.641	0.641	hm	0.985	0.985	nhm	0.879
36	CHEMBL319966	hm	0.849	0.849	hm	0.999	0.999	hm	1.000	1.000	hm	0.986	0.986	hm	0.978	0.978	hm	1.000	1.000	hm	0.954
37	CHEMBL41	hm	1.000	1.000	nhm	0.526	0.474	nhm	0.765	0.235	hm	0.977	0.977	hm	0.835	0.835	hm	0.977	0.977	nhm	0.879
38	CHEMBL40906	hm	0.878	0.878	hm	0.932	0.932	hm	0.656	0.656	hm	0.892	0.892	nhm	0.502	0.498	hm	0.865	0.865	hm	0.954
29	CHEMBL465161	hm	0.654	0.654	hm	0.514	0.514	hm	1.000	1.000	hm	1.000	1.000	hm	0.713	0.713	hm	1.000	1.000	hm	1.000
39	CHEMBL88106	hm	1.000	1.000	hm	0.926	0.926	hm	1.000	1.000	hm	1.000	1.000	hm	0.980	0.980	hm	1.000	1.000	hm	1.000

MLP 22-14-2 (tanh, softmax)

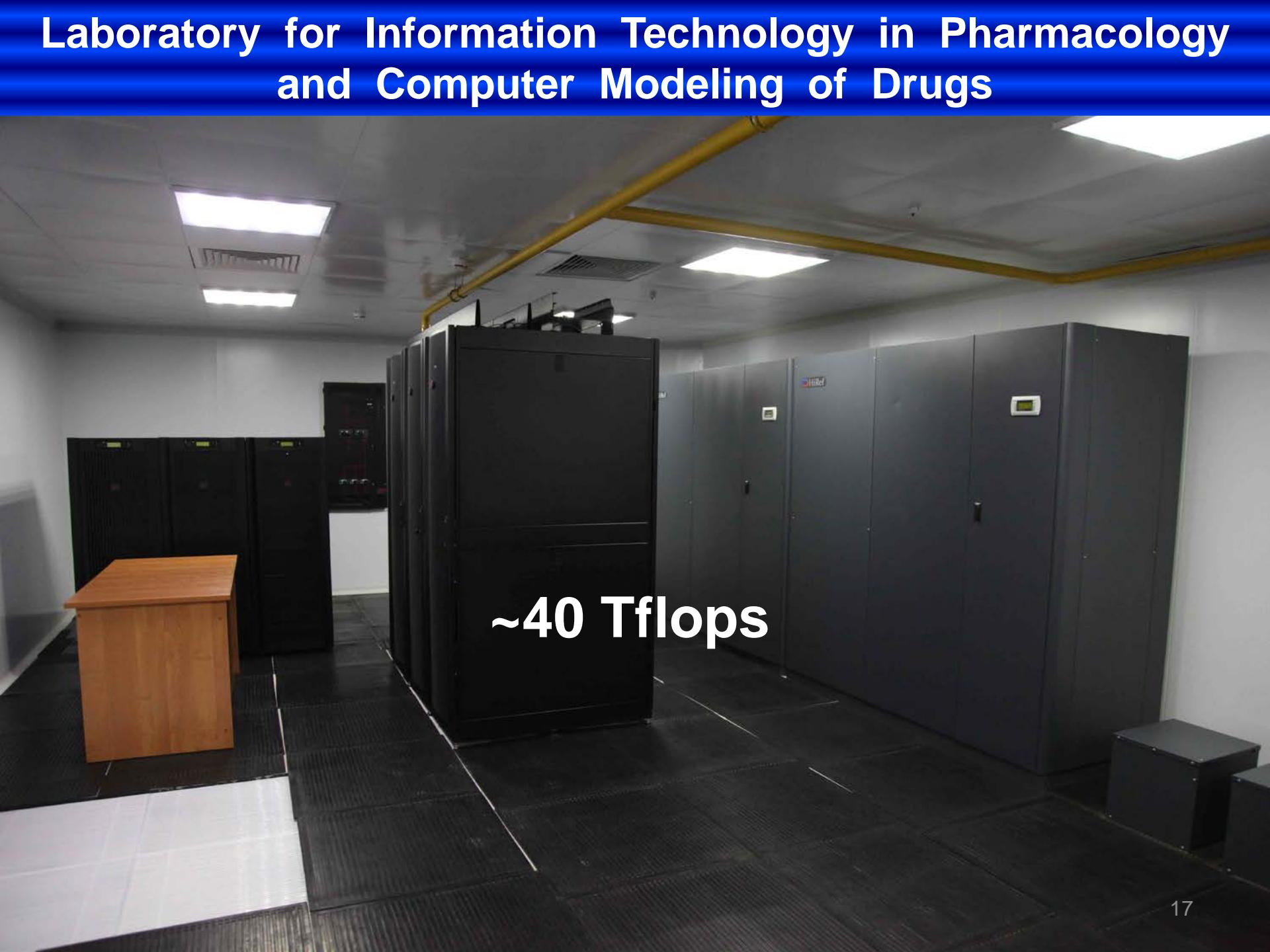
Accuracy $F_0 = 90.7\%$

Sensitivity $F_a = 90.9\%$

Specificity $F_n = 90.5\%$

AUC_{ROC} = 91.5%

Laboratory for Information Technology in Pharmacology and Computer Modeling of Drugs



~40 Tflops

Creative team

P.M. Vassiliev

M.A. Perfilev

A.V. Golubeva

A.N. Kochetkov

D.V. Maltsev

pvassiliev@mail.ru

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A night photograph of the Motherland Calls statue in Volgograd, Russia. The statue depicts a woman holding a sword aloft, standing on a hill. The sky is dark with scattered clouds. In the foreground, the Volga River is visible with some lights reflecting on the water. The statue is illuminated from below, and there are other smaller statues and lights in the background.

Thank You for Your
attention!