# Drug candidate prediction using machine learning techniques

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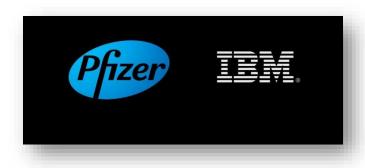
#### Backgrounds - Drug development process

# Prediction of drug-target interactions using deep neural networks model





#### **Pharmaceutical company**



Pfizer & IBM Watson : IBM and Pfizer to Accelerate Immunooncology Research with Watson for Drug Discovery

Janssen & BenevolnetAI (2016) : BenevolentAI signs exclusive license agreement with Janssen for clinical-stage drugs





GSK & Exscientia (2017) : GSK Launches Up-to-\$43M AI-Focused Collaboration with Exscientia





#### Why NOW? Drug discovery?

#### Why is Deep Learning Hot Now?









CPU		GPU
CORES CONTROL CACHE RAM	Internal Structure	CORES
<ul><li>Few complex cores</li><li>Specialized in Serial processing</li></ul>		<ul> <li>Many simple cores</li> <li>Built for Parallel processing (ex. Image)</li> </ul>
4~8	Cores	3000+
3~4 GHz	Clock Speed	~1.5 GHz
~1000 GFLOPS	Throughput	~10000 GFLOPS*
GILST Gwangju Institute of Science and Technology * FLOPS : Floating-Point Operations Per Second		ComSysBioLAB

#### **TOP 3 HIGH-END NVIDIA GPUs**



Name	GeForce GTX 1080 Ti	NVIDIA TITAN Xp	NVIDIA TITAN V
Architecture	Pascal	Pascal	Volta (Brand New)
CUDA cores	3584	3840	5120
Clock Speed	1582 MHz	1582 MHz	1455 MHz
VRAM Size	11GB GDDR5X (bus interface : 352 bit)	12GB GDDR5X (bus interface : 384 bit)	12GB HBM2 <sup>1</sup> (bus interface : 3072 bit)
Memory Bandwidth	484.4 GB/s	547.6 GB/s	652.8 GB/s
Price	\$699 USD	\$1,199 USD	\$ 2,999 USD
Tensor Cores <sup>2</sup>	<b>n/a</b> bosed of higher bus interface than G	n/a DDR5X.	640 ComSysBioLAE

Gwangiu Institute of science and Technology <sup>2</sup> Exclusive optimized cores for Deep learning operations.

#### **Drug discovery**?

#### Why is Deep Learning Hot Now?





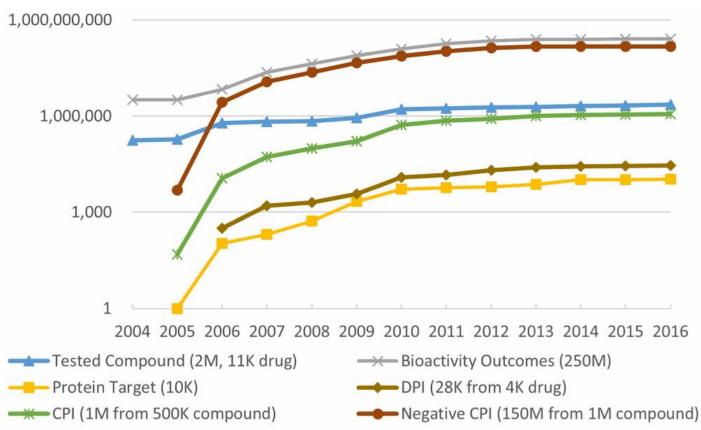


#### **DATA DRIVEN DRUGS**





### Big Data ! (chemical compound, target)



**Fig. 2.** The growth of biological data in PubChem BioAssay including biologically tested compounds, bioactivity outcomes, protein targets, drug-protein interactions (DPIs), compound-protein interactions (CPIs), negative compound-protein interactions (CPIs). The number in parenthesis is the total count of each data category. DPI and CPI are counted based on the confirmatory and literature-based assays



Cheng et al., The AAPS journal 2017



### Big Data ! (beyond the compounds)

#### Opinion

Trends in Pharmacological Sciences September 2014, Vol. 35, No. 9

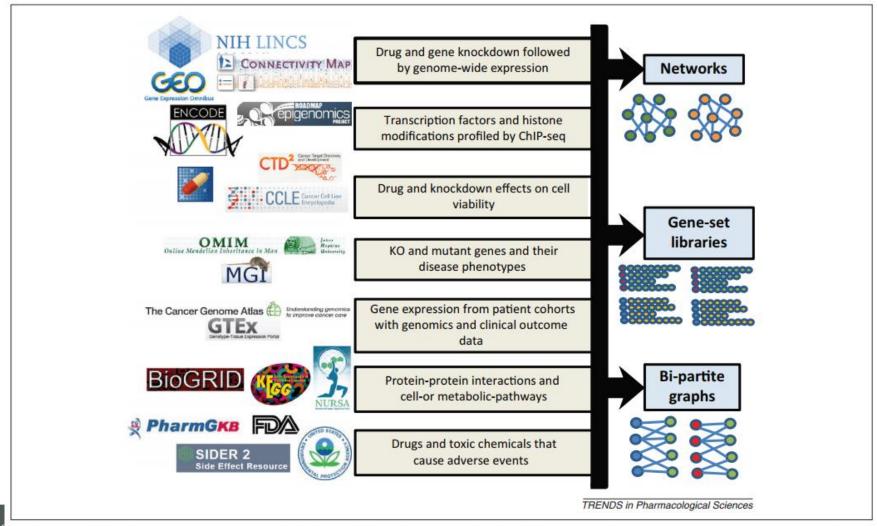
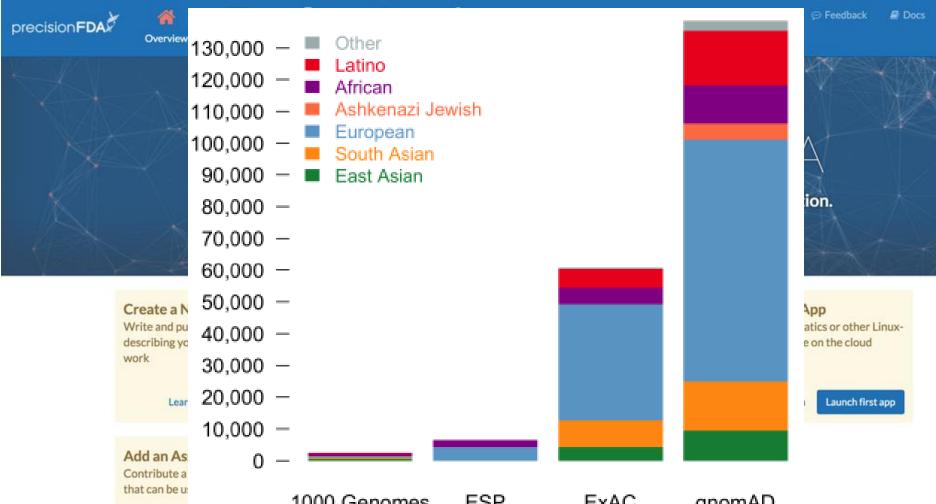


Figure 3. Resources from systems biology and systems pharmacology can be integrated by first identifying the various objects, their relations, and their data types and their data types and the converting the data into single-entity weighted networks, fuzzy-set libraries, or weighted multipartite graphs.

#### **Considering individual genomes**



1000 Genomes ESP ExAC gnomAD The long-term goal of the platform is to streamline the process of evaluating tests leading to medical patients being able to get precise care based on their own individual genomic data

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#### **Drug discovery**?

#### Why is Deep Learning Hot Now?







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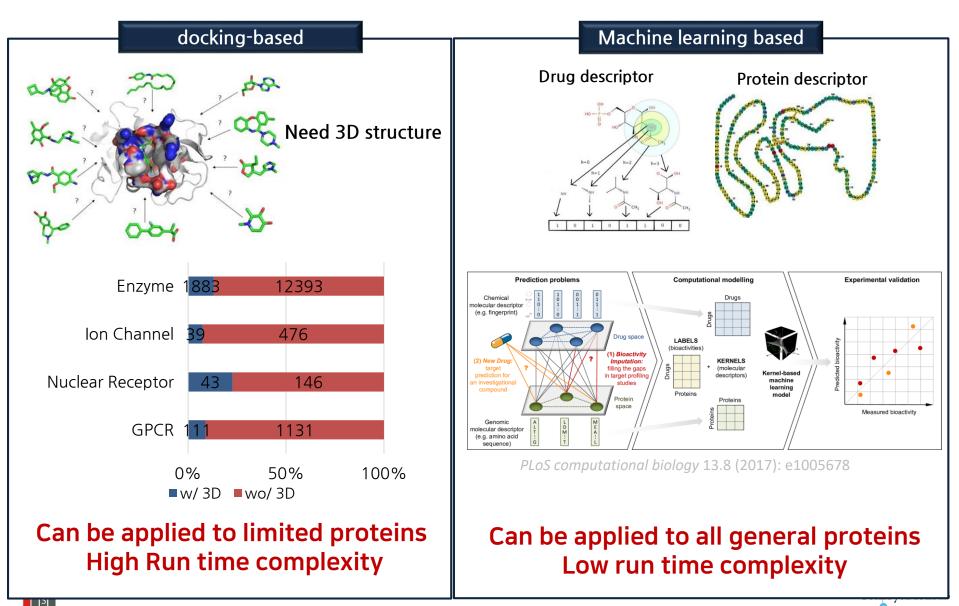
Ingoo Lee, Jongsoo Keum, Hojung Nam<sup>\*</sup>, "DeepConv-DTI: Prediction of drug-target interactions via deep learning with convolution on protein sequences", *Bioinformatics*, Under review.

### **DRUG-TARGET INTERACTION**



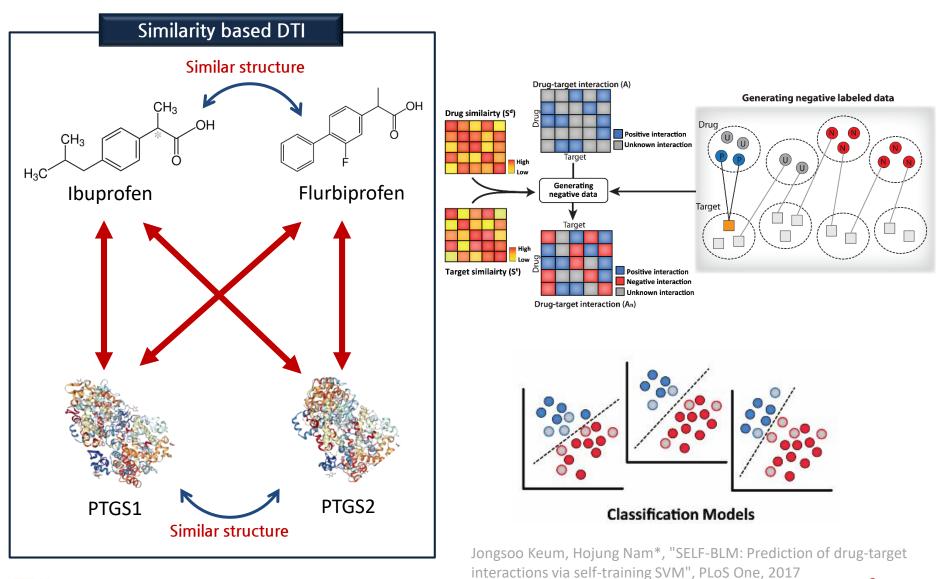


#### in-silico based DTI approaches



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#### **Similarity-based method**

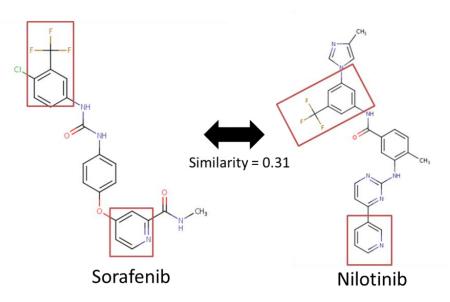






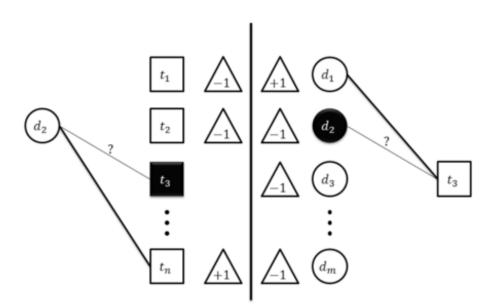
#### **Problems of Similarity-based method**

Miss prediction



- Common important substructures, but low similarity because of their proportion
- Does not work well (Low performance)

• High time complexity



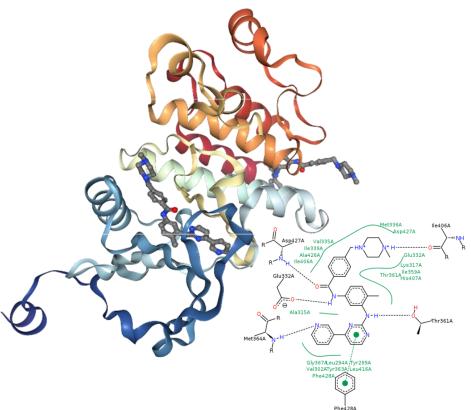
- Large scale data for performance
- Time complexity:  $O(n_d n_t + n_t n_d) \approx O(n^2)$

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Hard to train (High time cost)

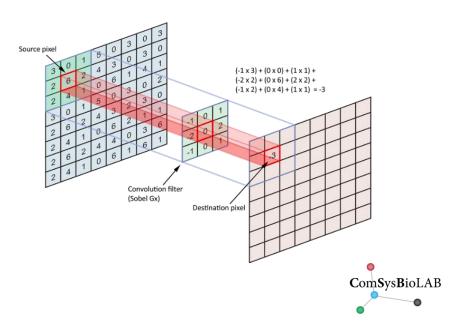


#### DeepConv-DTI : CNN based model to detect binding regions



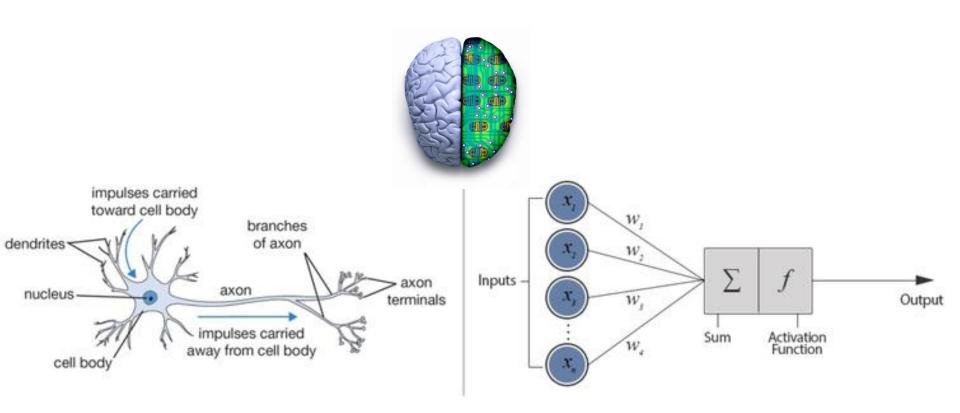
• 3D co-crystal structure of gleevec and its interaction with target protein

- Binding region of target protein have a pattern to interact with drug
- We can use patterns of binding regions to predict DTI for machine learning model
- Recently, convolutional neural network have received attraction to extract local patterns





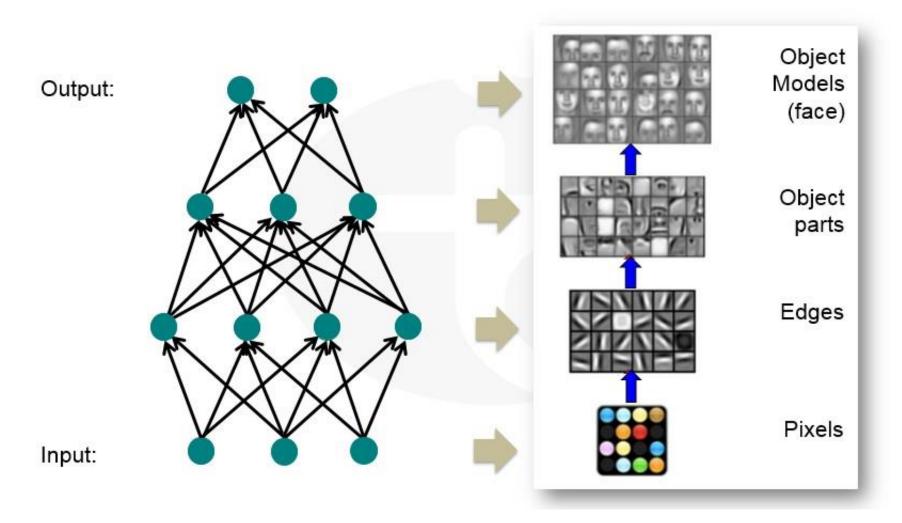
#### Neuron vs. Perceptron







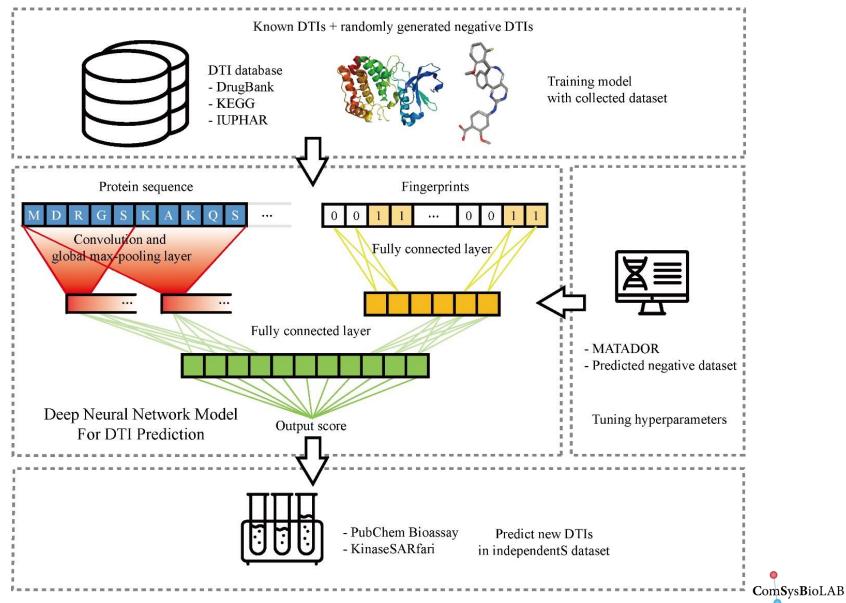
#### **Deep Neural Network**







#### **Overview**

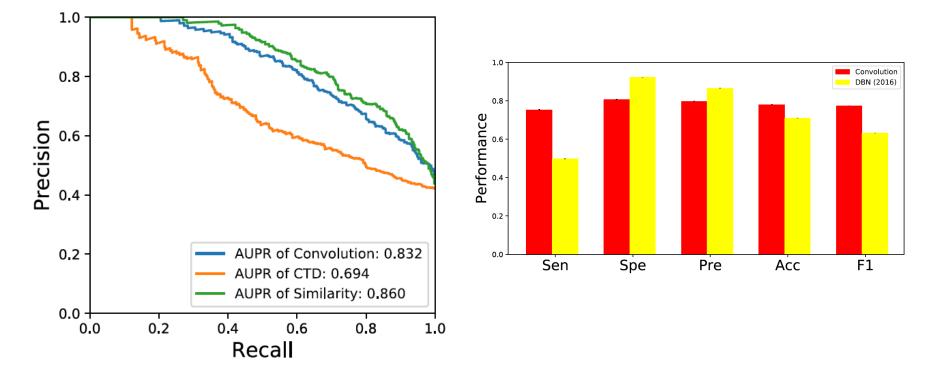


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#### **Performance evaluation**

 Comparison with other protein descriptor (vs CTD, similarity-based)

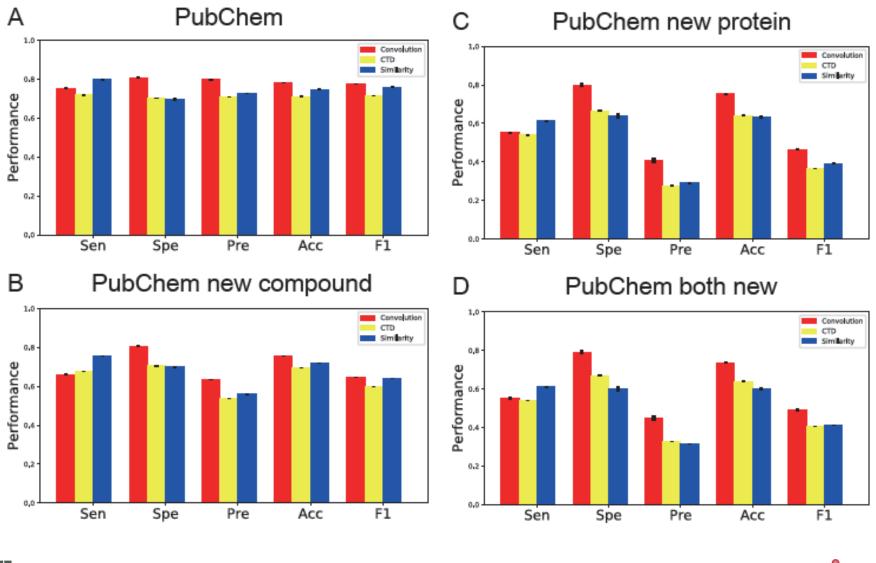


#### Precision-recall curves



Comparison with other DNN model
 (vs DBN)

#### **Performance evaluation**

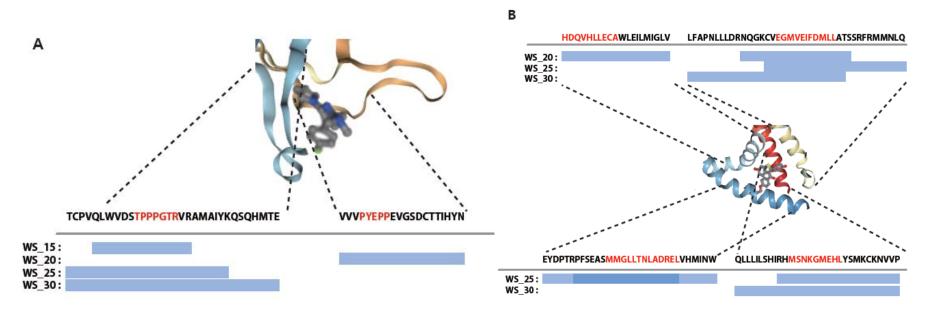


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- Comparison with other protein descriptor (vs CTD, similarity-based)

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#### Validation of extracted patterns

Compare pooled convolution result with binding sites from sc-PDB



- cellular tumor antigen protein (P04637, P53\_HUMAN)
- Estrogen receptor protein (P03372, ESR1\_HUMAN)
- Pooled convolution results covers actual binding site





## Summary (DTI)

- Propose a CNN based DTI prediction model
- The model show high prediction power in independent data sets
- The proposed model captures informative binding sites that contribute DTIs





## Thank you -Q&A-





