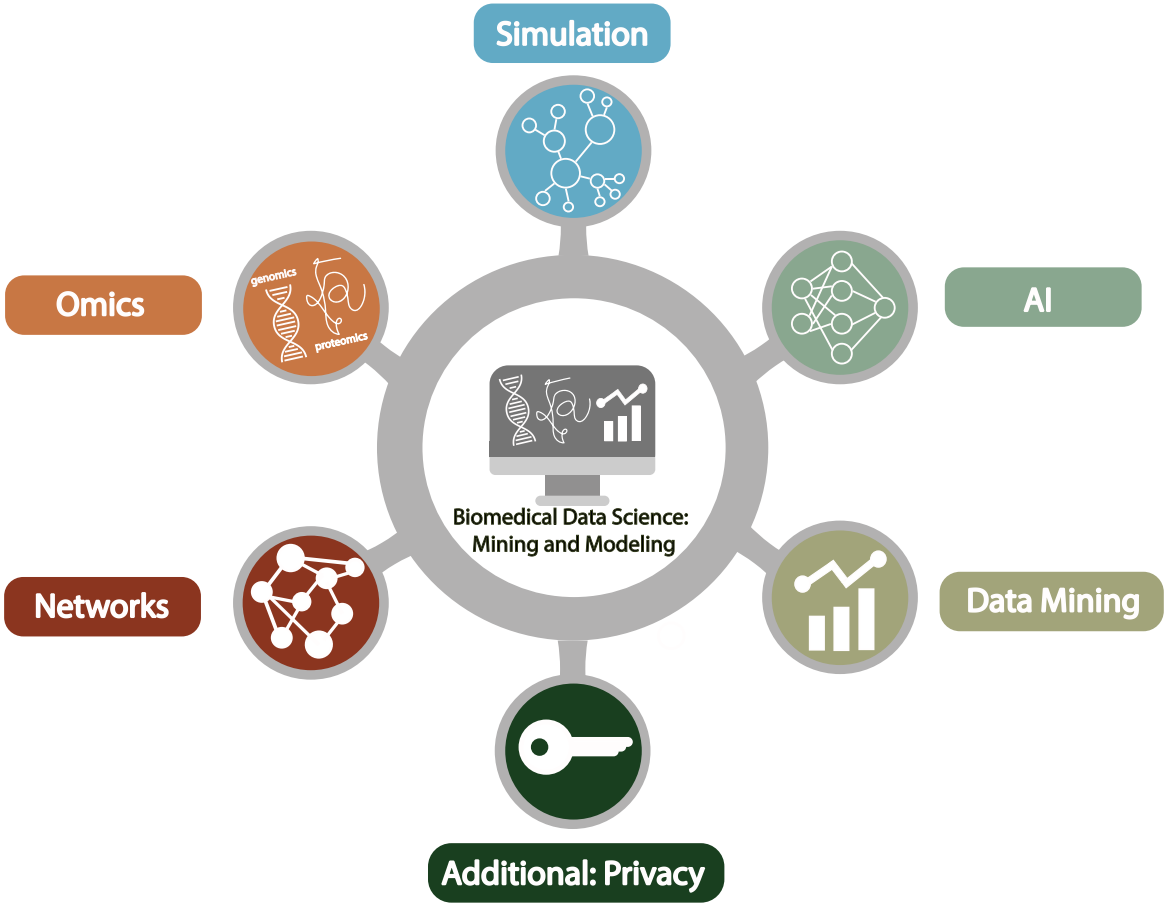
Biomedical Data Science (GersteinLab.org/courses/452)

Transition from Mining to Modeling (23i3)



Mark Gerstein
Yale U.

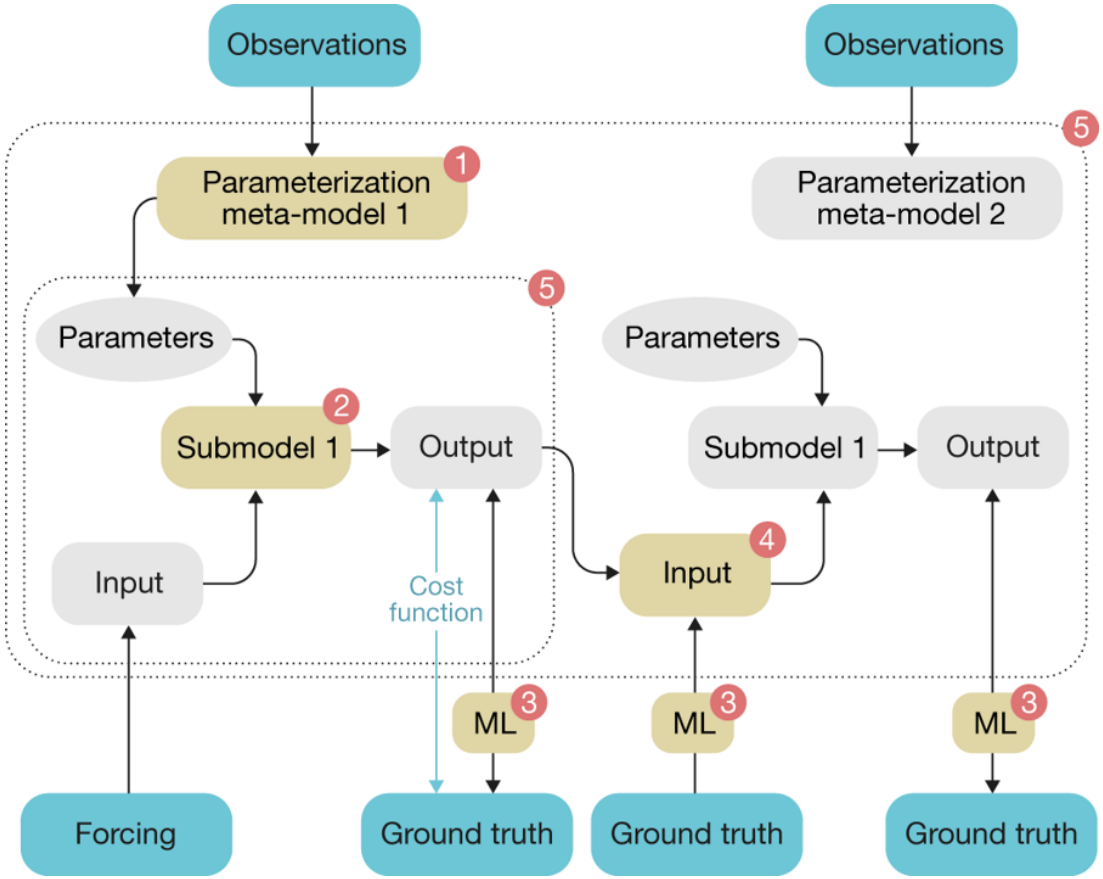
Last edit in spring ’23.
Very similar to 2019’s pack #14.
New video in ’23.

Combining Mining & Modeling

- Complementarity of physical & ML approaches
  - "Physical approaches in principle being directly interpretable and offering the potential of extrapolation beyond observed conditions, whereas data-driven approaches are highly flexible in adapting to data"
- Hybrid #1: ML into physical
  - e.g., Emulation of specific parts of a physical model for computational efficiency
  - More..
- Hybrid #2:
  Physical knowledge can be integrated into ML framework
  - Network architecture
  - Physical constraints in the cost function
  - Expansion of the training dataset for under sampled domains (i.e., physically based data augmentation) [More….]

https://www.nature.com/articles/s41586-019-0912-1
2019 Nature volume 566, pages195–204 (2019)

Hybrid #1: ML into physical models



(1) Improving parameterizations

(2) Replacing a 'physical' sub-model with a machine learning model

(3) Analysis of model–observation mismatch

(4) Constraining submodels

(5) Surrogate modelling or emulation

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Example of Hybrid #2: Integrating Physical Knowledge into Machine Learning

Physical Data Augmentation for Hybrid Physical-Statistical Model Construction: Example for Building a Model to Predict Sensitivity of Drug Binding to a SNV in a Protein

The Major Hurdle: Highly Scant Ligand Binding Assay Data for ΔBA

>10M SNV — Number of sequenced exonic SNVs

~2.6K SNV+Drug — ~2.6k drug ligands associated with ~60k SNVs

~175K SNV+PDB — ~175k SNVs mapped with at least one human PDB (>2.8Å)

~10K SNV-protein-drug — SNV-drug-structure tuples in GenoDock Dataset

~0.1k — experimental ΔBA records of human protein from Platinum Dataset (gold-standard for ΔBA)

Too scant to train or to parameterize a statistical learning model

The Physically-based Data Augmentation Approach: Leveraging Physical Calculations of ΔBA to Fill the Gap

(Reichstein *et al.*, **Nature**, 2019 & Xie et al., preprint, 2018)

- Expansion of the training dataset for under sampled domains
- Data augmentation is crucial to avoid overfitting

~10K — SNV-protein-drug tuple in GenoDock

Physically-based Data Augmenting

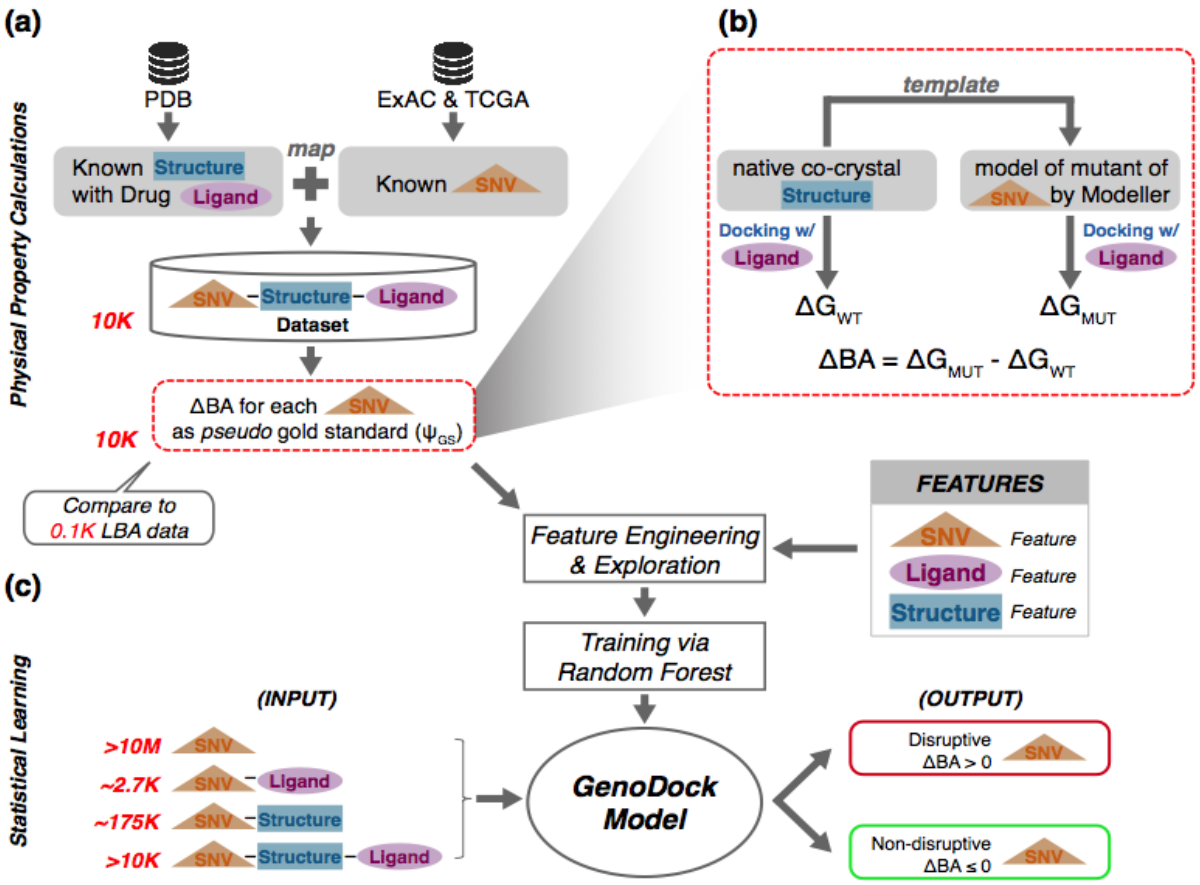
~0.1K — experimental ΔBA records (under-sampled)

(*87 human protein records from Platinum)

~10K — ΔBA for each SNV-protein-drug tuple (*pseudo* gold-standard)

[Wang et al. Structure ('19)]

Framework of the GenoDock Project - from Dataset Preparation to Model Construction



Wang *et al.*, accepted by *Structure*, 2019

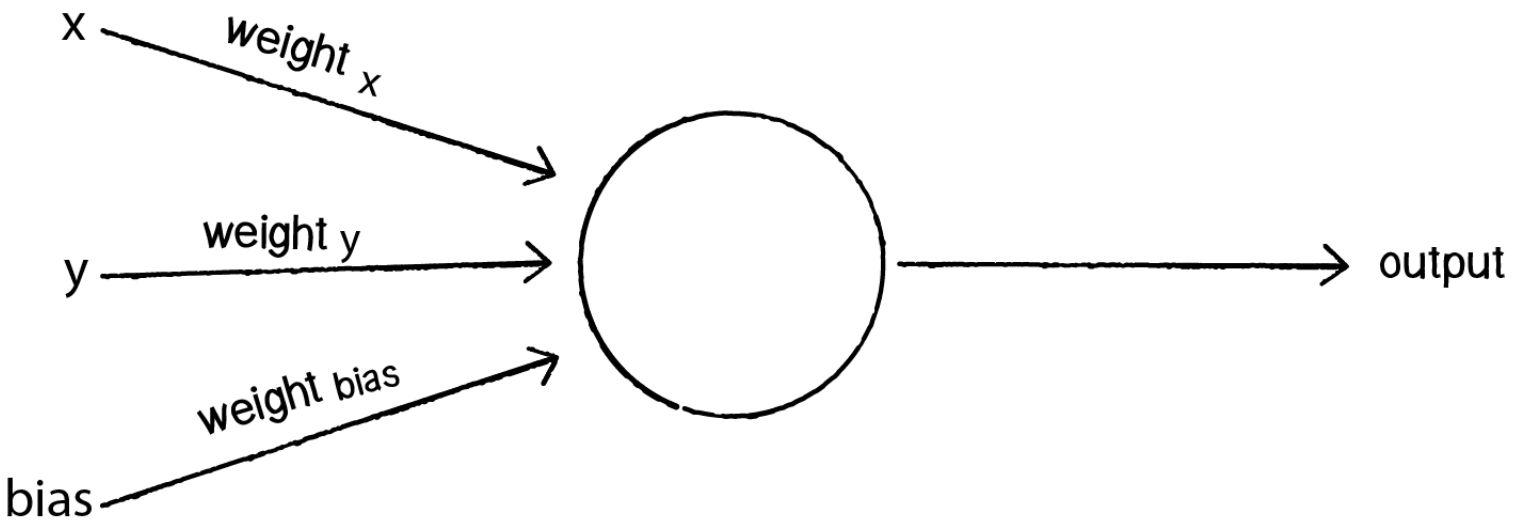
a collecting and processing raw data

b Ligand binding model for Δ BA calc.

c Statistical model training and testing

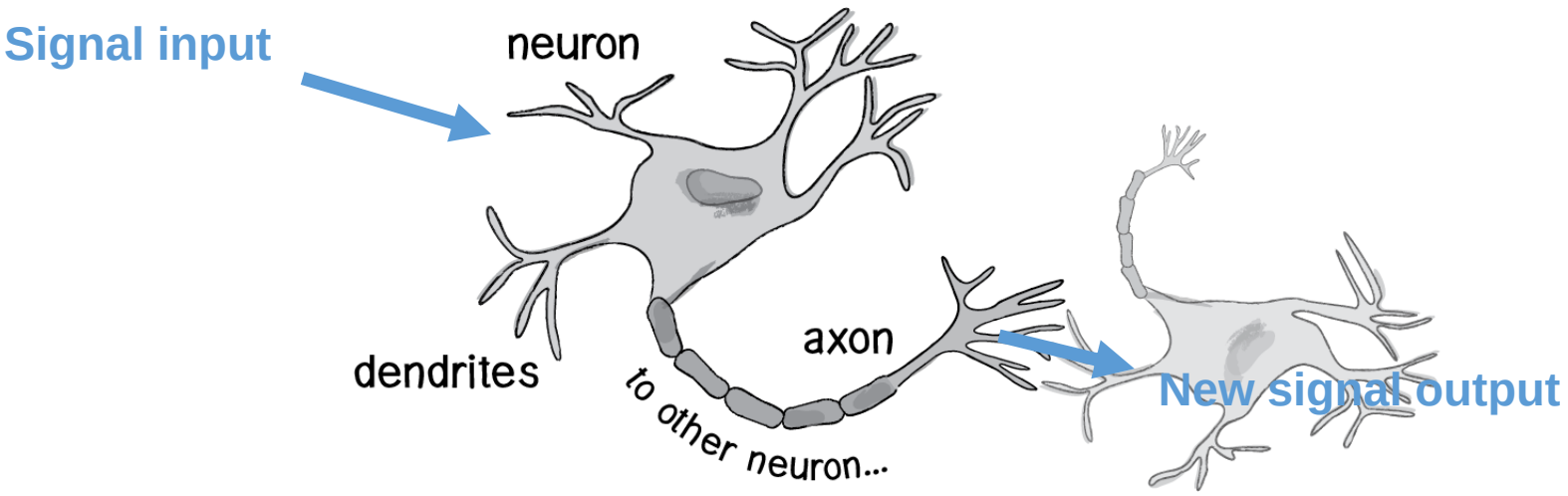
[Wang et al. Structure ('19)]

Perceptron



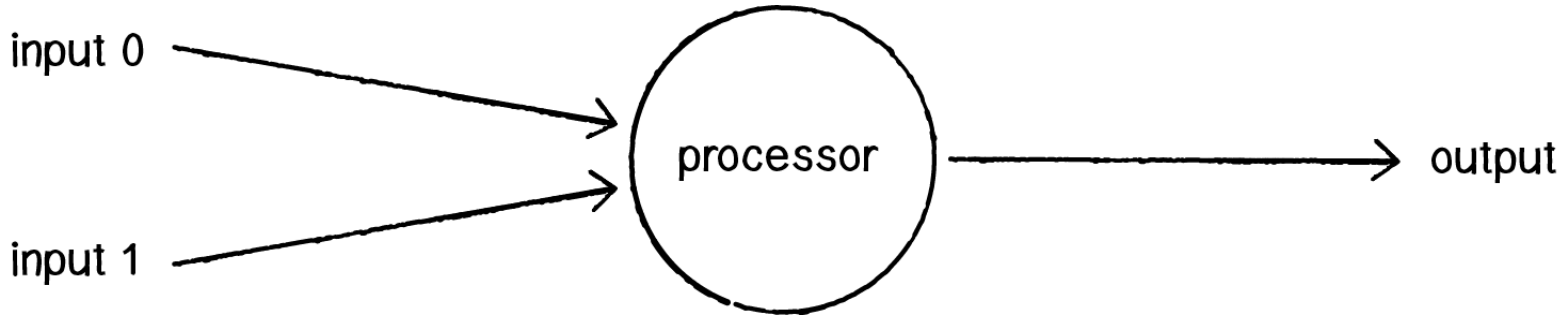
- A perceptron is the **simplest neural network** possible: a computational model of a single neuron
- Works as a linear classifier with n-1 dimension hyperplane
- It computes the elementary logical functions we usually think of as underlying computation, functions such as AND, OR, and NAND

Neural Networks



- Artificial Neural Network is loosely connected to biological neural networks
- A simple artificial node known as “neuron” (or “unit”) has
  1. Inputs
  2. Outputs
  3. Adaptive weights
  4. Activation

Example



1. **Multiply each input by its weight**.
2. **Sum** all of the weighted inputs.
3. Compute the output of the perceptron based on that sum passed through an activation function (e.g., the **sign of the sum**).