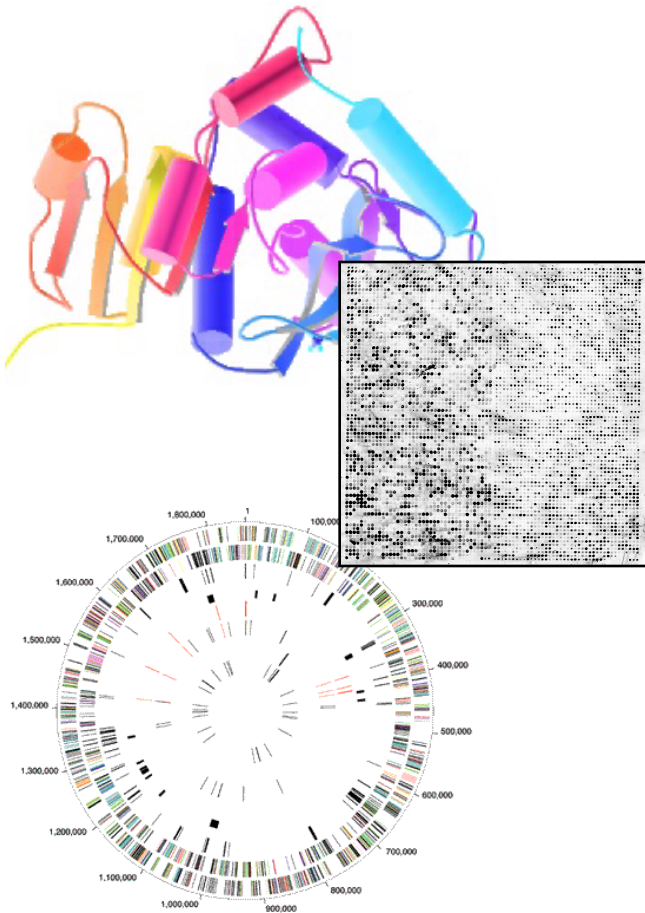


Biomed. Data Science:
Transition from Mining to Modeling

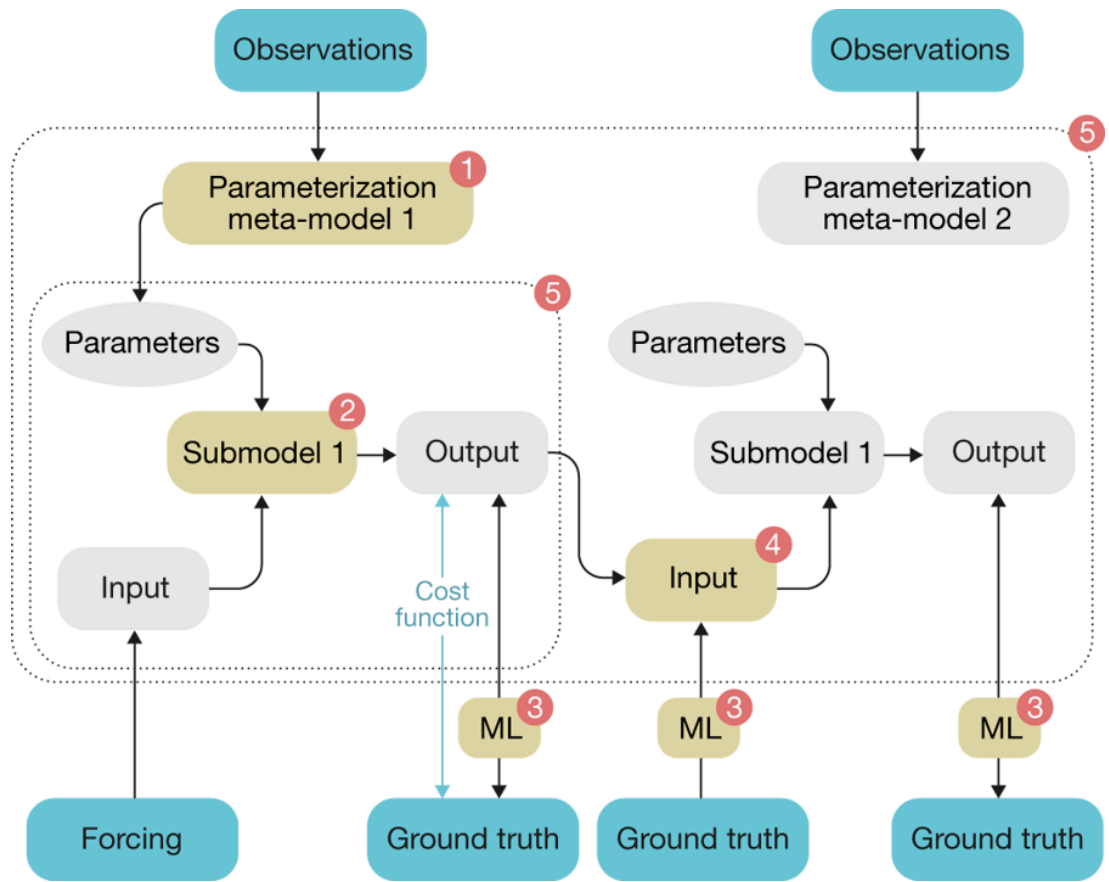


Mark Gerstein, Yale University
gersteinlab.org/courses/452
(last edit in spring '19, pack #14)

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 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 (ie physically based data augmentation) [More....]

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

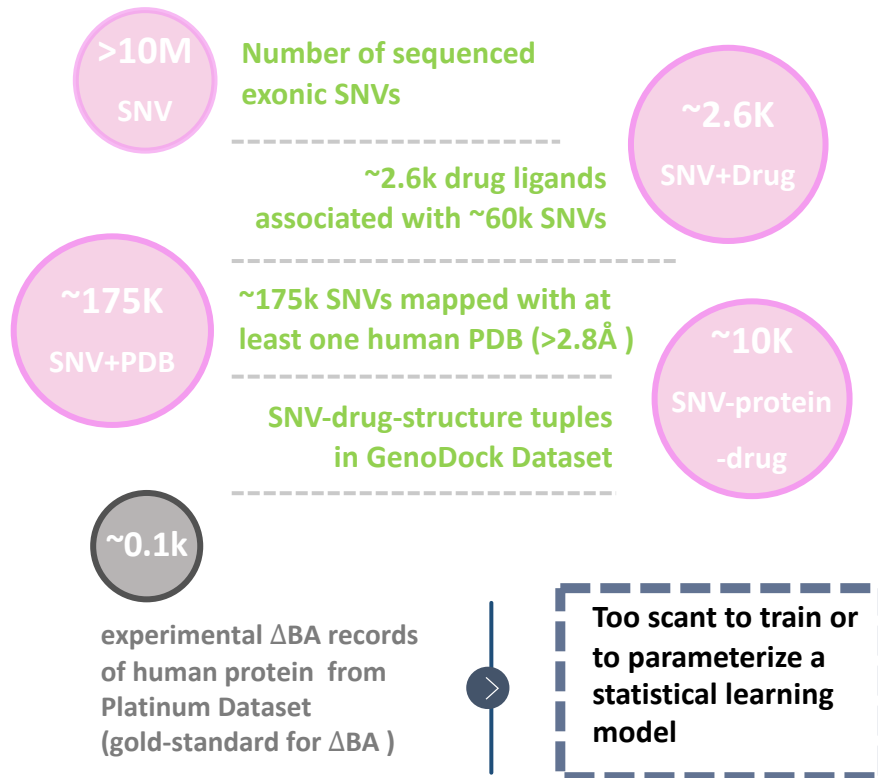
(4) Constraining submodels

(5) Surrogate modelling or emulation

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

Physical Data Augmentation for Hybrid Physical-Statistical Model Construction

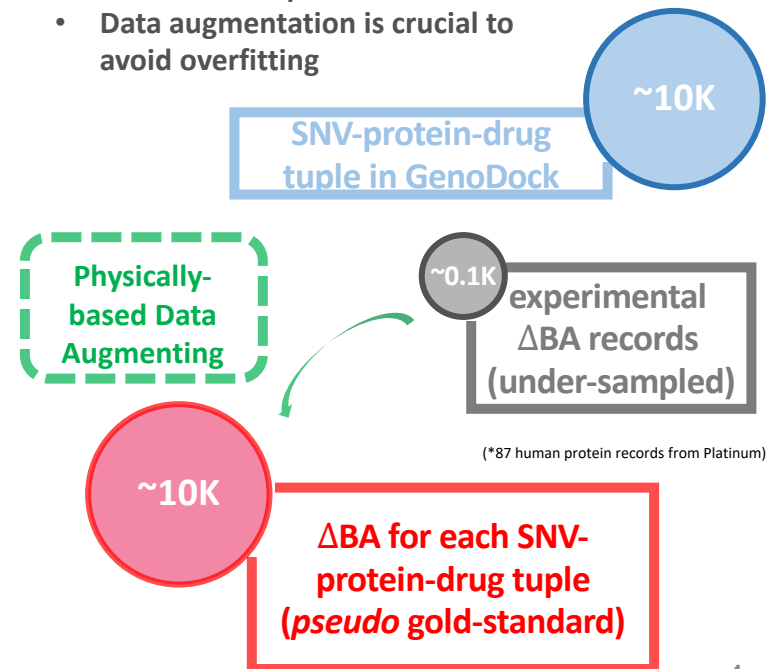
The Major Hurdle:
Highly Scant Ligand Binding Assay Data for Δ BA



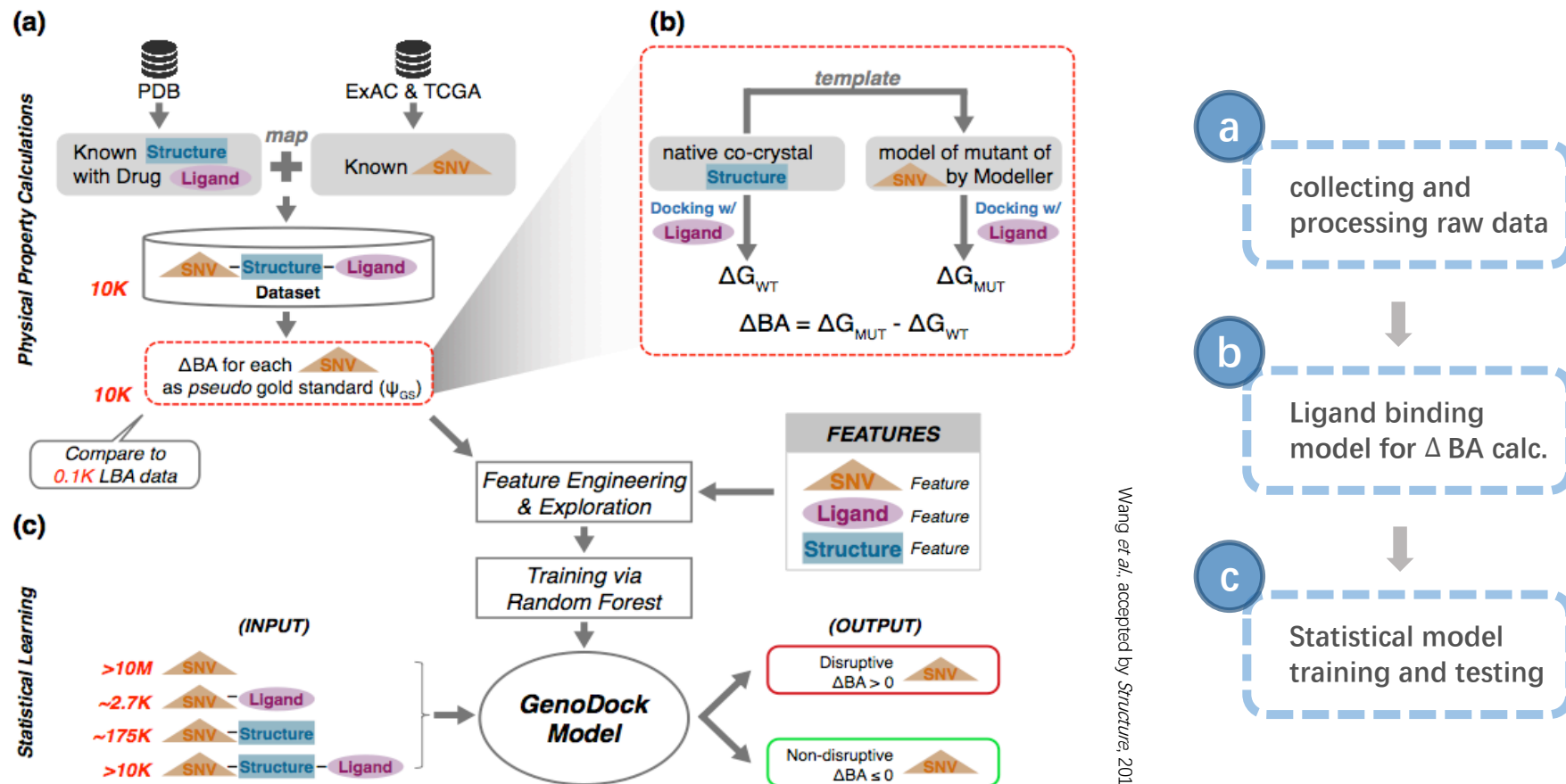
The Physically-based Data Augmentation Approach:
Leveraging Physical Calculations of Δ BA to Fill the Gap

(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



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



Wang et al., accepted by Structure, 2019