

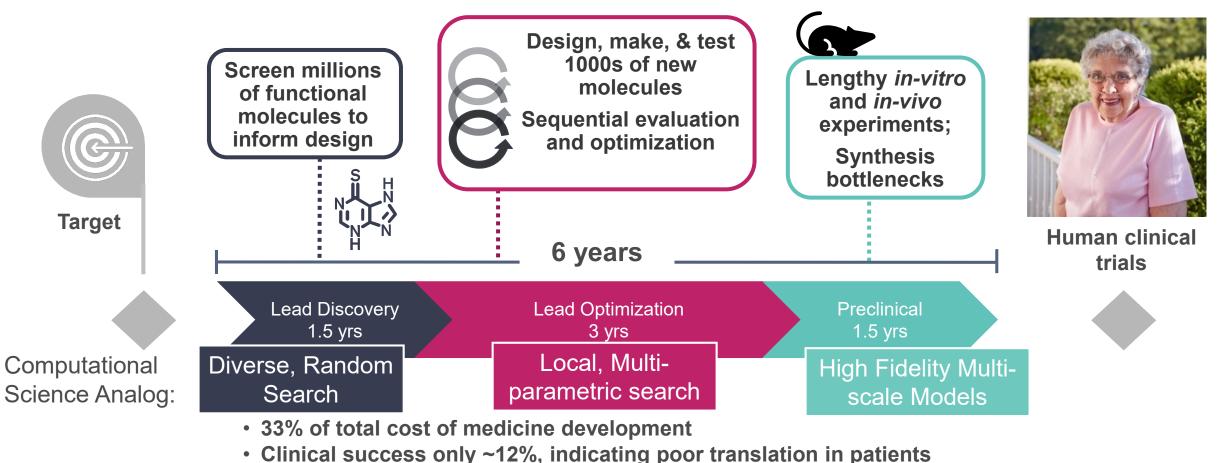
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SC 2019 Computational Approaches for Cancer Workshop

# Current drug discovery: long, costly, high failure

Is there a better way to get medicines to patients?

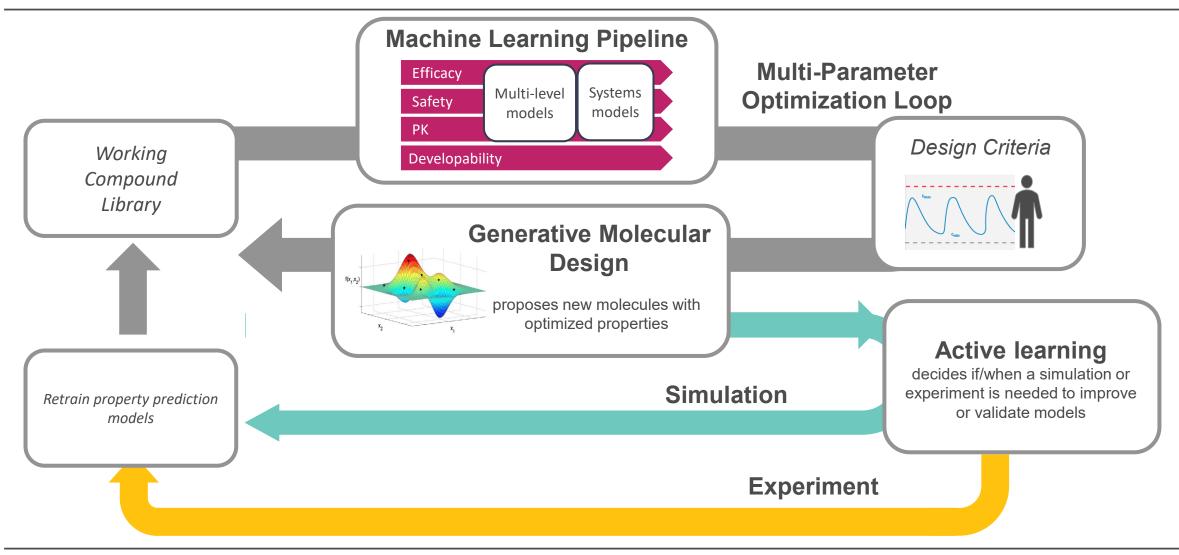


Source: http://www.nature.com/nrd/journal/v9/n3/pdf/nrd3078.pdf



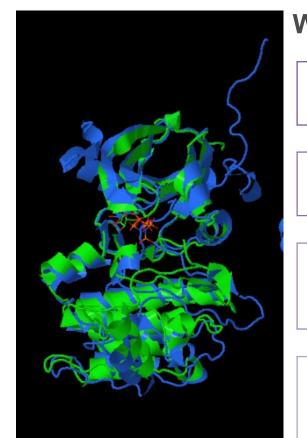
# The ATOM Platform

Active Learning Drug Discovery Framework



#### Application Generative Molecular Design to Aurora Kinase Case Study

**Case Study:** Can the ATOM generative design framework develop a novel, potent, selective Aurora Kinase B inhibitor, while maintaining select developability properties (i.e. PK/safety)



#### Why Aurora Kinase?

#### **Oncology Relevant**

A family of serine/threonine kinases involved in cell division pathways

#### Validated Target

≥12 inhibitors of AURK A and or AURK B have progressed to clinical trials

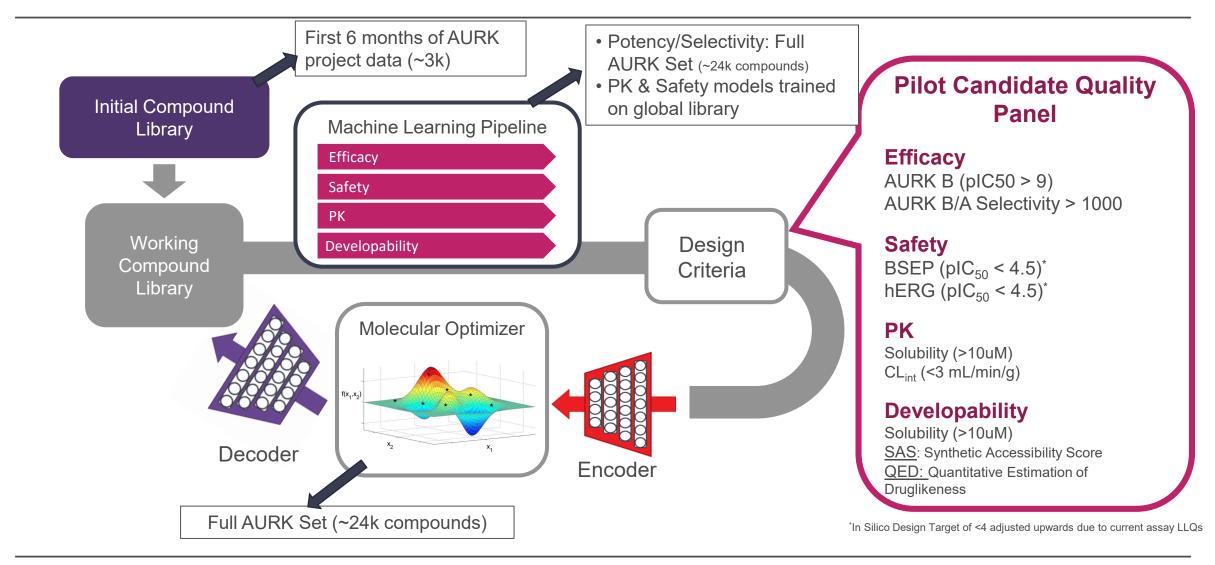
#### **Representative Drug Discovery Problem**

Selectivity between kinases is a challenging, but not intractable problem for drug discovery

#### Availability of Target-Specific Data

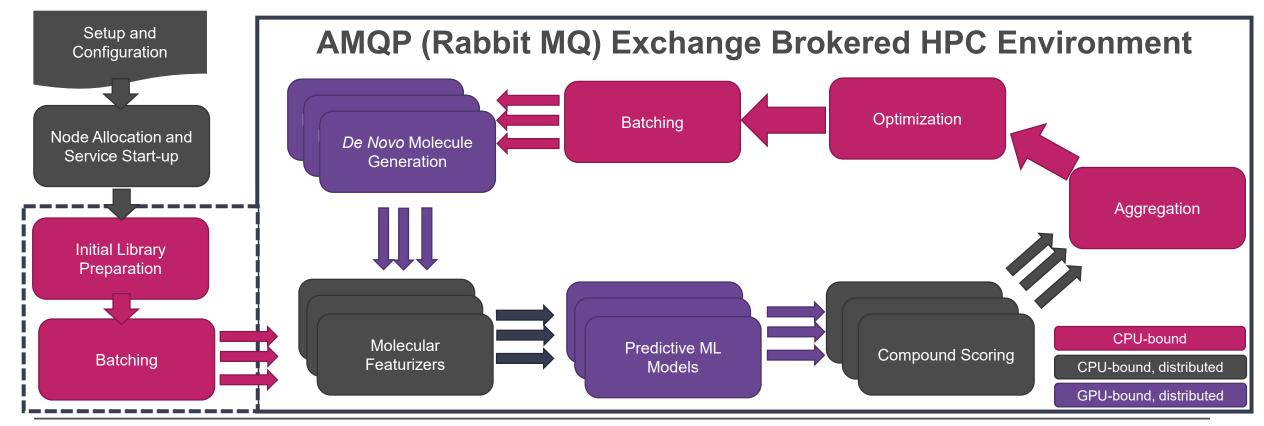
Potency data on ~24k compound available for AURK A and/or AURK B, which allows for assessment of problem at different phases of discovery

# Case Study Design and Targets



### High Performance Compute Facilitates Large Scale Search Enables Scalable Management of Heterogeneous Compute Tasks

- Facilitated ideation and evaluation of >3 million compounds in 24 hour run time
- Future scaling by 10x or more achievable on currently utilized ~100 node clusters

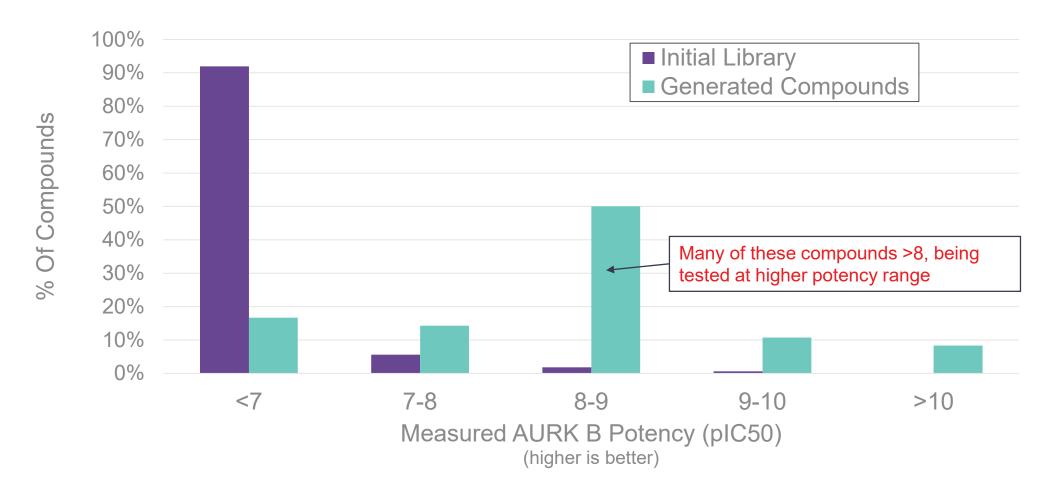


#### In Silico Optimization Proposed Library of Compound With High Quality *Predicted* Properties



Compound	AURK B plC50	AURK B/A Selectivity (fold)	hERG plC <sub>50</sub>	BSEP pIC <sub>50</sub>	hLM Clearance (mL/min/g)	Solubility (uM)	SAS
	•						
Compound 1	9.2	5287	3.3	4.3	3.6	1096	2.5
Compound 2	9.3	3233	3.2	4.2	2.5	399	2.4
Compound 3	9.6	11512	3.6	4.4	2.2	412	2.6
Compound 4	9.6	2449	3.2	4.3	2.5	60	2.3
Compound 5	9.7	3068	3.3	4.3	2.0	1155	2.5
Compound 6	9.6	5756	3.7	4.5	4.3	232	2.3
Compound 7	9.3	3296	3.3	4.4	2.6	33	2.4
Compound 8	9.1	1197	3.3	4.2	2.4	268	2.5
Compound 9	9.2	7724	3.3	4.3	2.3	733	2.7
Compound 10	10.1	2270	3.2	4.5	2.6	139	2.4

### Make Test Results Confirm High On-Target Potency 70% Of Tested Compounds with pIC<sub>50</sub> > 8



De Novo Synthesis & Testing Confirms Enrichment of High Potency Compounds

### Where To Next?

- Increase in problem scope
  - Scaling evaluation criteria: full safety and PK panels
  - Scaling of generative framework: Recently increased unique search space by >5x
- Application on realistic lead optimization scenarios
  - Reduction in target specific data
  - Integration of uncertainty quantification
  - Active learning for explore/exploit selection of compounds and model retraining
- Integration of structure based design techniques for virtual hit finding
- Integration of systems modelling for therapeutic window-based optimization

# Acknowledgements

### ATOM Generative Molecular Design Team

- Jason Deng
- Kevin McLoughlin
- Tom Sweitzer
- Jeff Mast
- Juliet McComas
- Margaret Tse
- Derek Jones

# **ATOM Joint Research Committee**

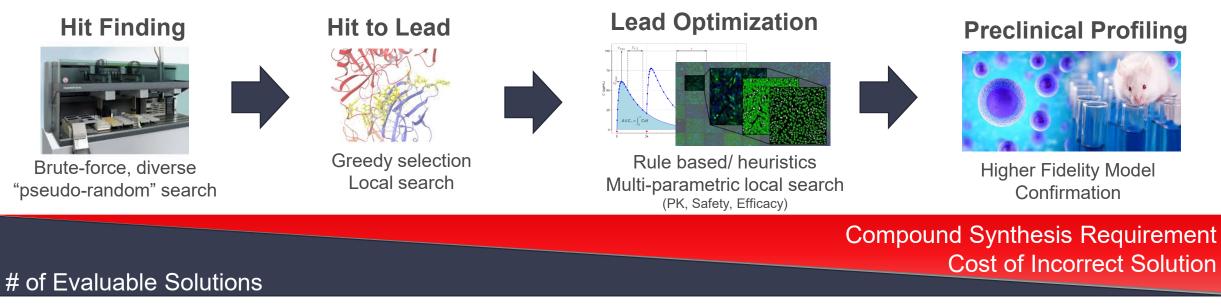
- Tom Rush
- Jim Brase
- Stacie Calad-Thomson
- Michelle Arkin
- Dwight Nissley



#### Backup Slides

# Drug Discovery As a Search Problem

Physical, multi-parametric, multi-fidelity process leads to long, expensive process

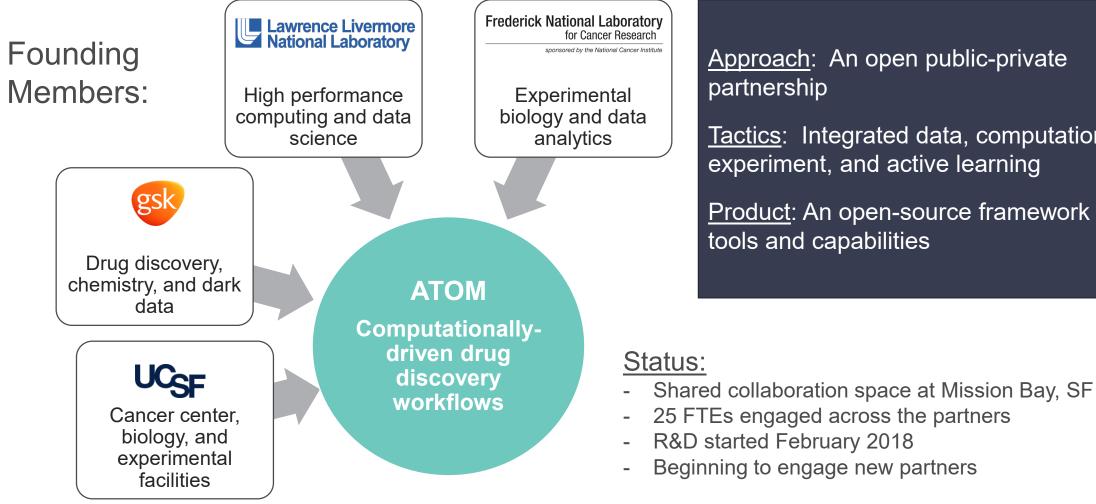


#### **Issues:**

- Large search space(10<sup>20</sup> 10<sup>60</sup> options)
- Long cycle times for *de novo* chemistry
- Difficult multi-factorial decisions

**Hypothesis:** Significant acceleration through *in silico* generative search with integration of automated QSAR/CADD and systems modelling techniques

# Accelerating Therapeutics for Opportunities in Medicine **ATOM Consortium**

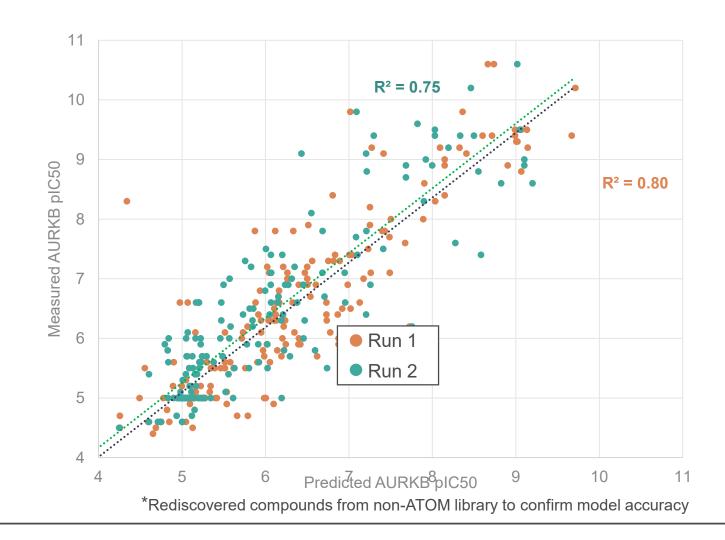


Approach: An open public-private

<u>Tactics</u>: Integrated data, computation, experiment, and active learning

Product: An open-source framework of tools and capabilities

# Additional "Rediscovered" Compounds with Existing Data Validates Predative Accuracy of The Models<sup>\*</sup>



ATOM