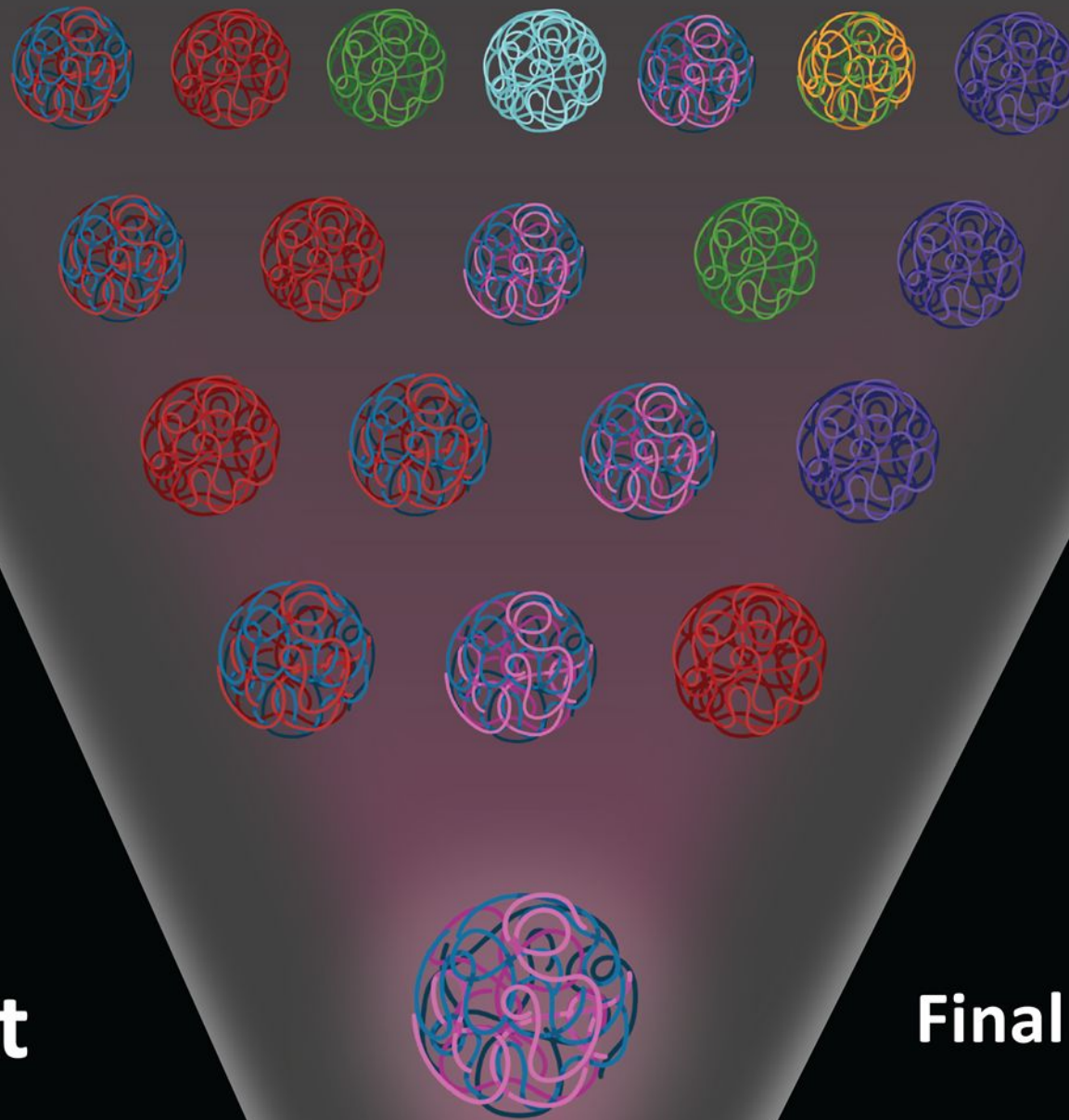


Machine Learning Directed Drug Formulation

Development

Zeqing Bao
Leslie Dan Faculty of Pharmacy
University of Toronto

Formulation Development Process



Particle Size

Stability

Drug Loading

Drug Release

Final Formulation

Formulation Development



Automation

Artificial Intelligence

Prediction Machines



The Simple Economics of
Artificial Intelligence

AJAY
AGRAWAL

JOSHUA
GANS

AVI
GOLDFARB

“

Artificial intelligence is a technology
of low-cost
prediction and discovery

”

What prediction problems do we have
in pharmaceutical sciences ?



AI in Pharmaceutical Sciences

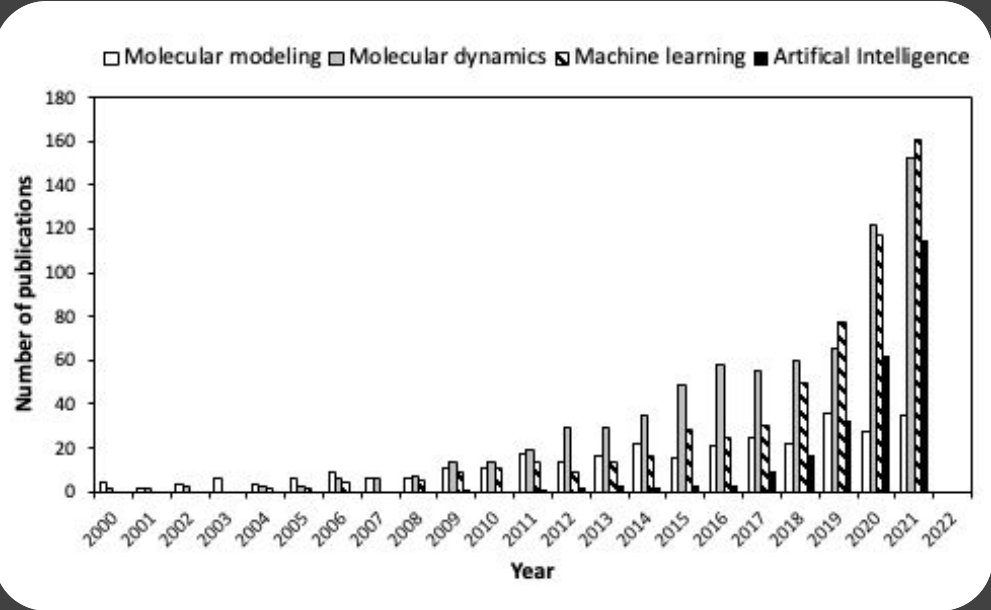


Journal of Controlled Release
Volumes 311–312, October 2019, Pages 326–327

Cover Story

Does artificial intelligence have the potential to transform drug formulation development?

Christine Allen , Paurie Bannigan





VECTOR INSTITUTE



Roche



NOVARTIS



Microsoft

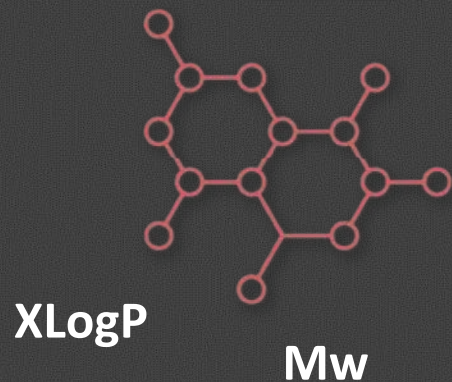
Creating an Intelligent and Connected Digital Workforce



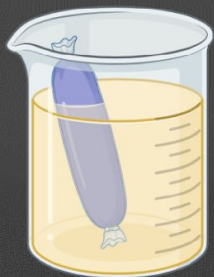


Identifying New Treatments for Chronic Kidney Disease & Idiopathic Pulmonary Fibrosis

What would a ML approach look like?

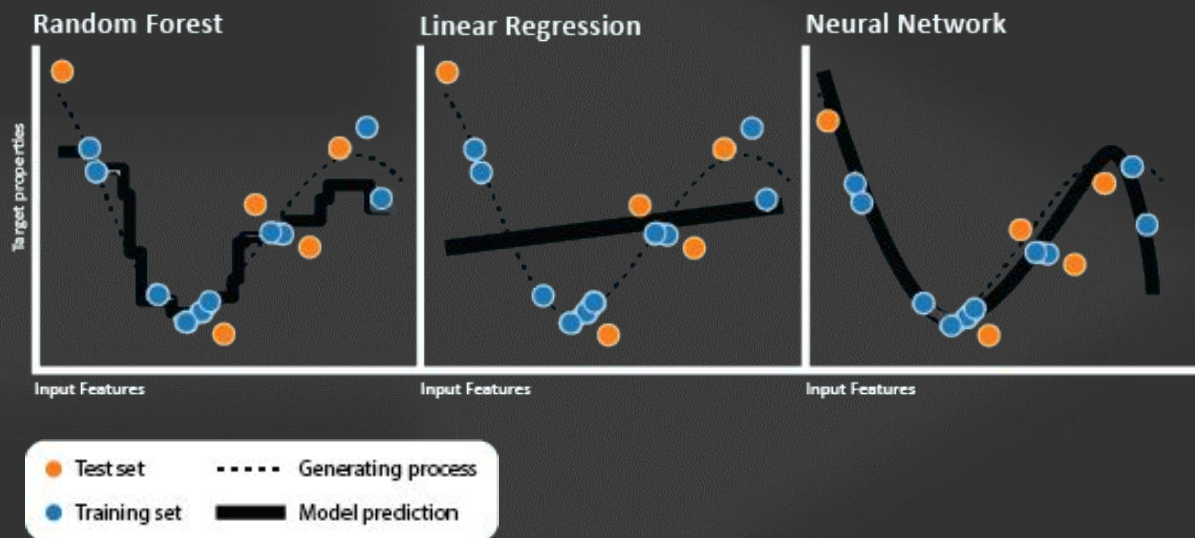


Features



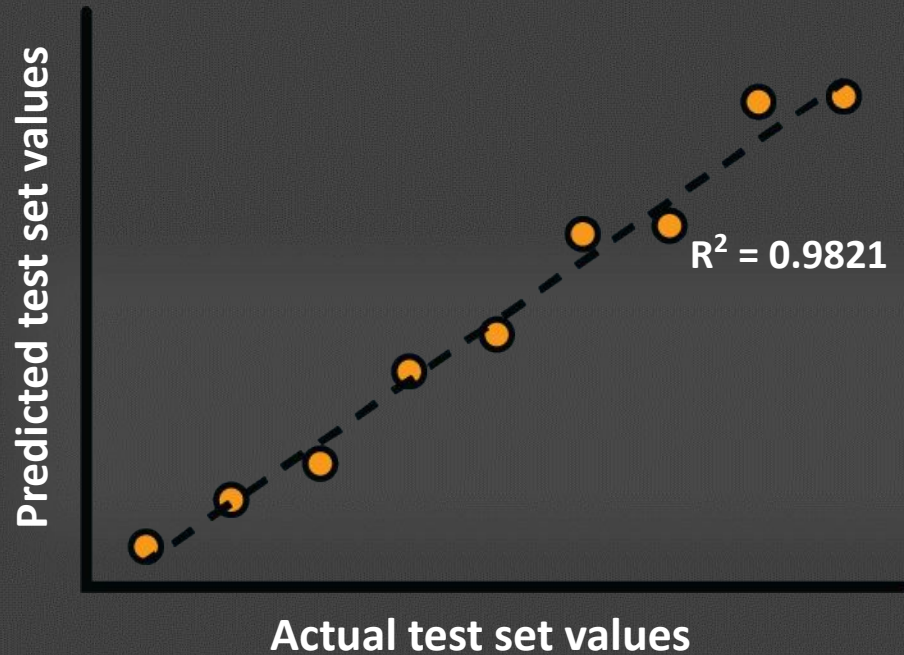
Release media
composition

Dataset Construction

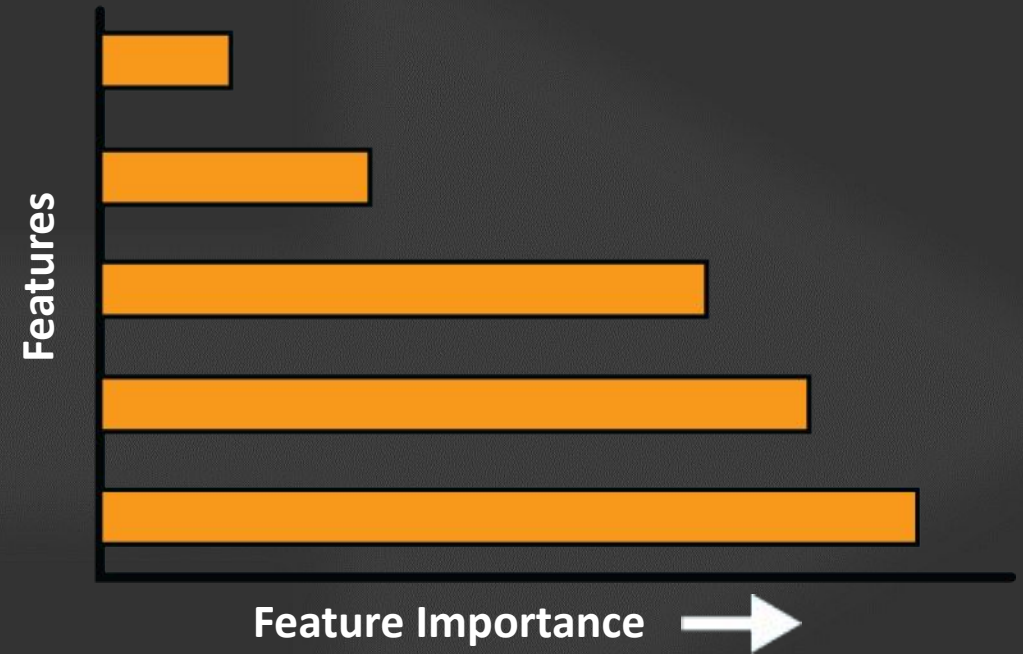


Model Selection

What would a ML approach look like?



Model Evaluation



Model Interpretation

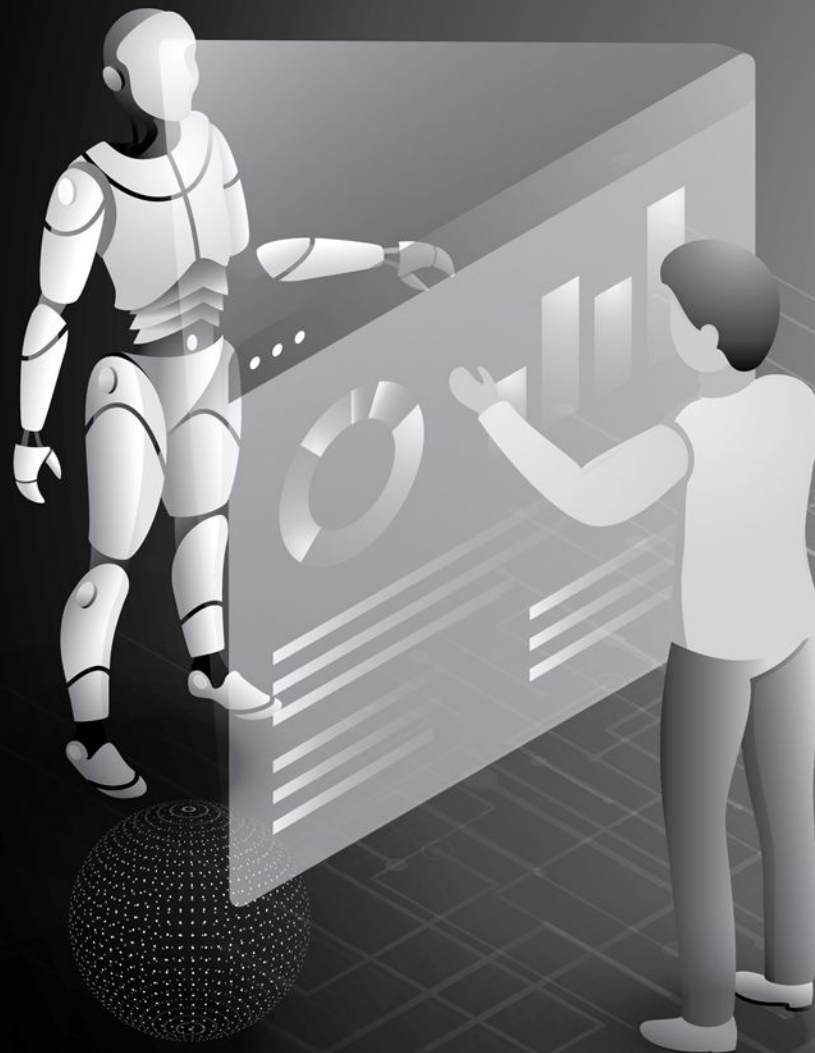
 the
matter lab

+

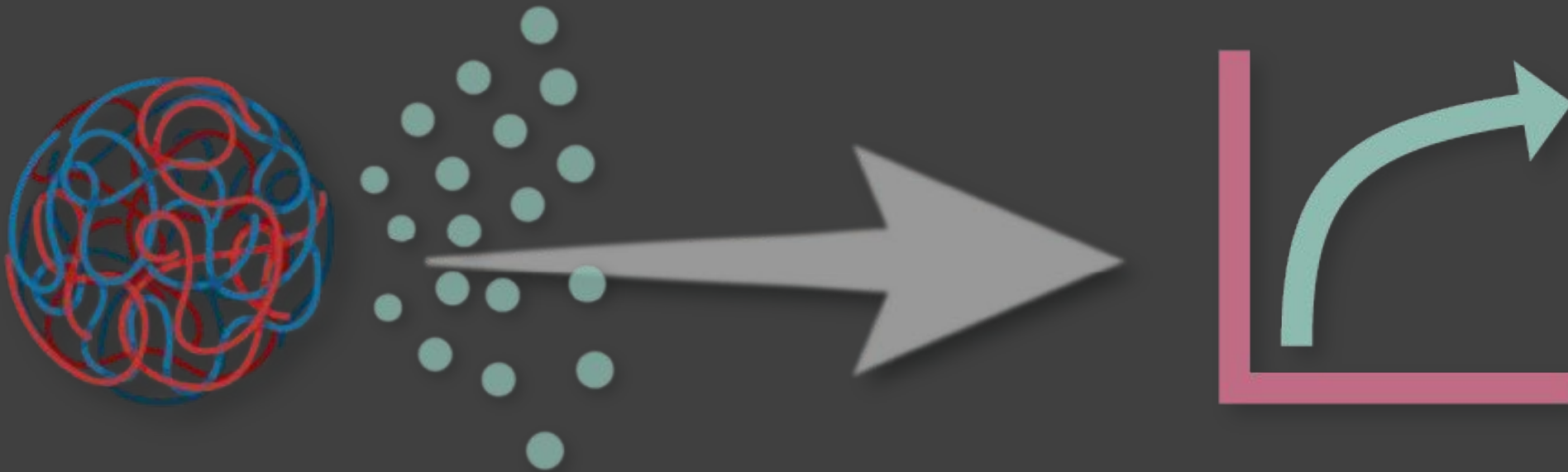
 ALLEN
LAB

Prediction of experimental conditions

Prediction of formulation performance



Prediction of Drug Release from MPs



 **Journal of Pharmaceutical Sciences**
Volume 110, Issue 7, July 2021, Pages 2771-2777

Pharmaceuticals, Drug Delivery and Pharmaceutical Technology

Poly(δ -valerolactone-co-allyl- δ -valerolactone) cross-linked microparticles: Formulation, characterization and biocompatibility

Zeqing Bao ^{a,1}, Sungmin Jung ^{a,1}, Jack Bufton ^a, James C. Evans ^a, Dean J. Aguilar ^b, Christine Allen ^{a,*} 

 **European Journal of Pharmaceutical Sciences**
Volume 162, 1 July 2021, 105808

Cross-linked valerolactone copolymer implants with tailorable biodegradation, loading and *in vitro* release of paclitaxel

Jack Bufton ^{a,1}, Sungmin Jung ^{a,1}, James C. Evans ^a, Zeqing Bao ^a, Dean Aguilar ^b, Christine Allen ^{a,*} 

Data Collection



WEB OF SCIENCE™

Data extracted from 40 papers



~200 release profiles



Training data:

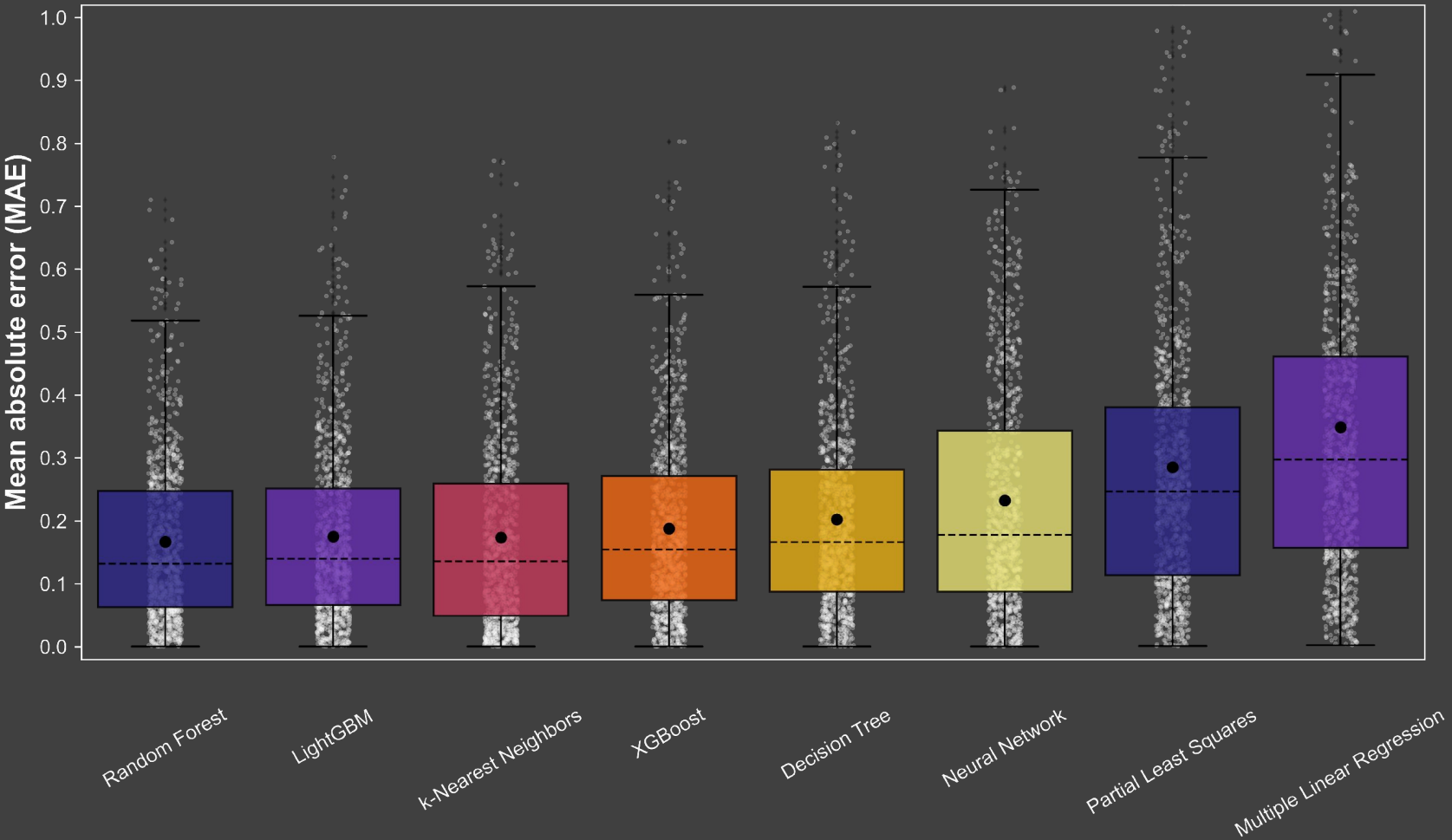
~ 100 drug release



Validation data:

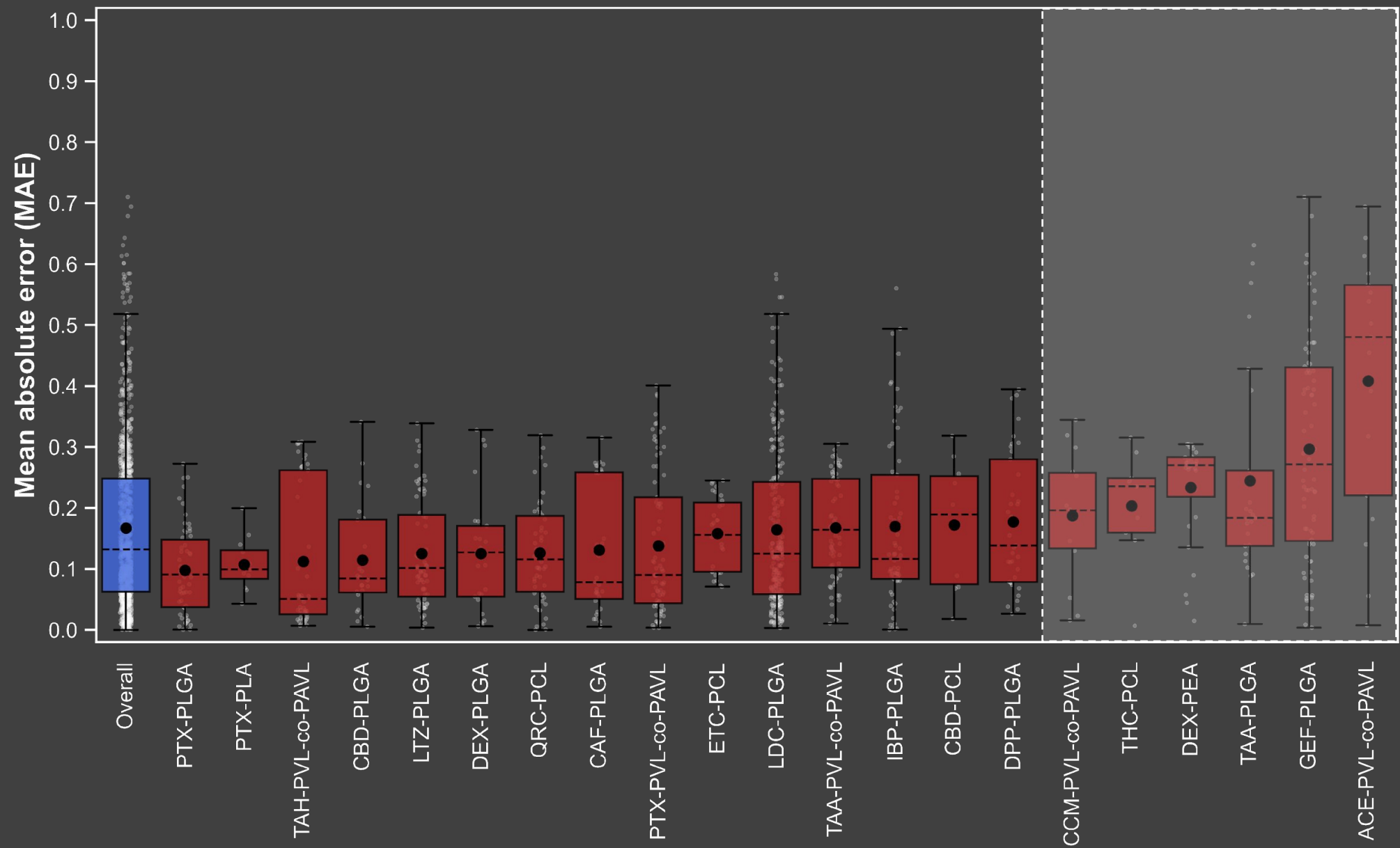
~ 80 drug release

ML Model Screening and Selection

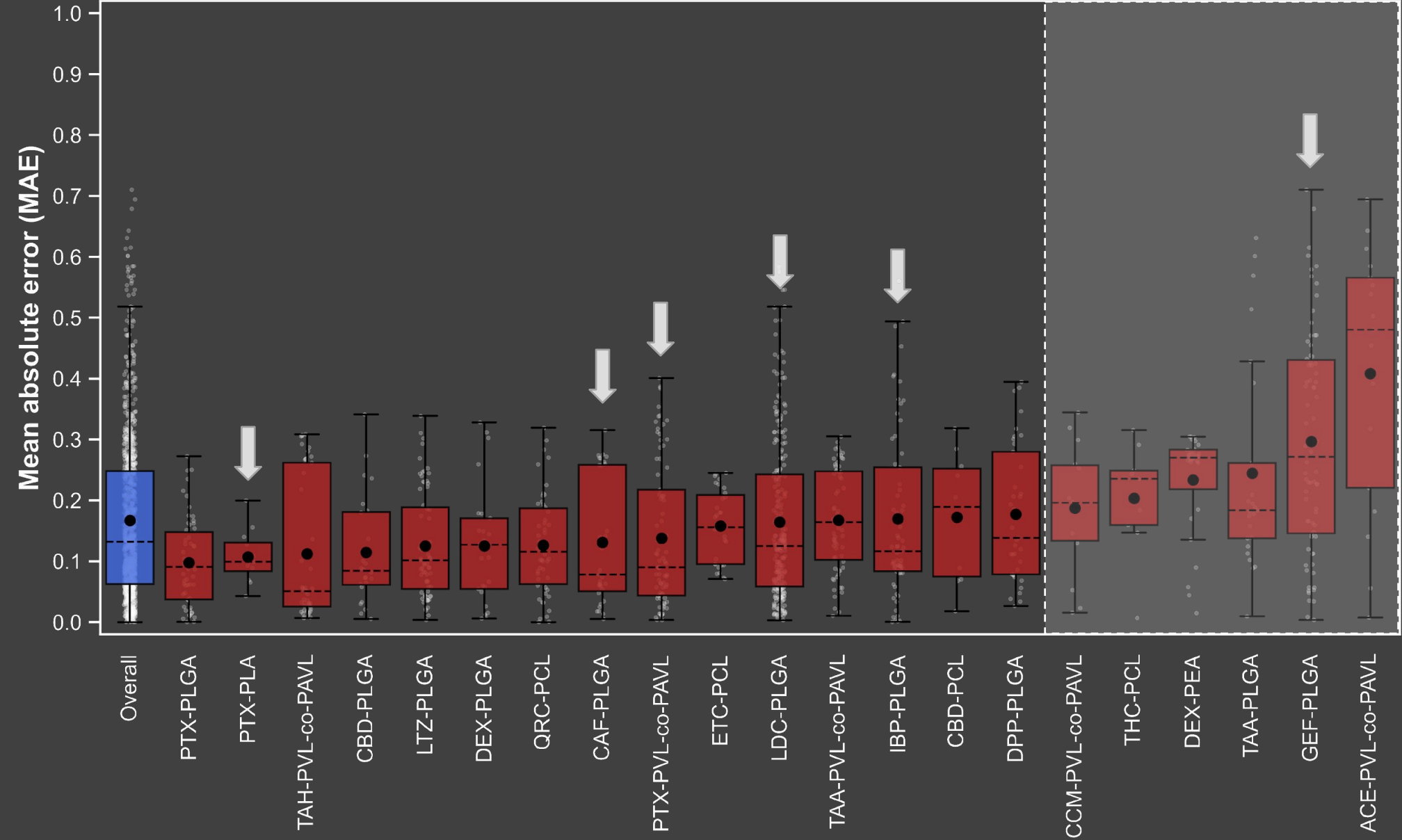


	Random Forest	LightGBM	k-Nearest Neighbors	XGBoost	Decision Tree	Neural Network	Partial Least Squares	Linear Regression
Count	1017	1017	1017	1017	1017	1017	1017	1017
Mean	0.17	0.18	0.17	0.19	0.20	0.23	0.29	0.35
Stdev	0.13	0.14	0.15	0.15	0.16	0.19	0.23	0.27
Median	0.13	0.14	0.14	0.15	0.17	0.18	0.25	0.30

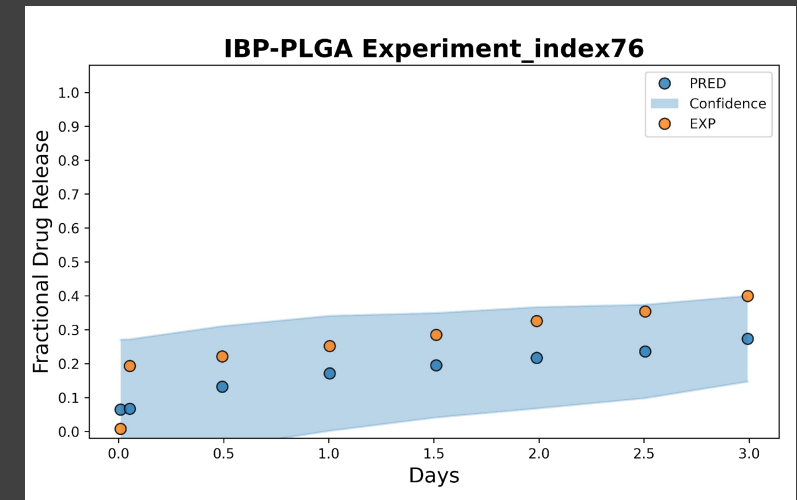
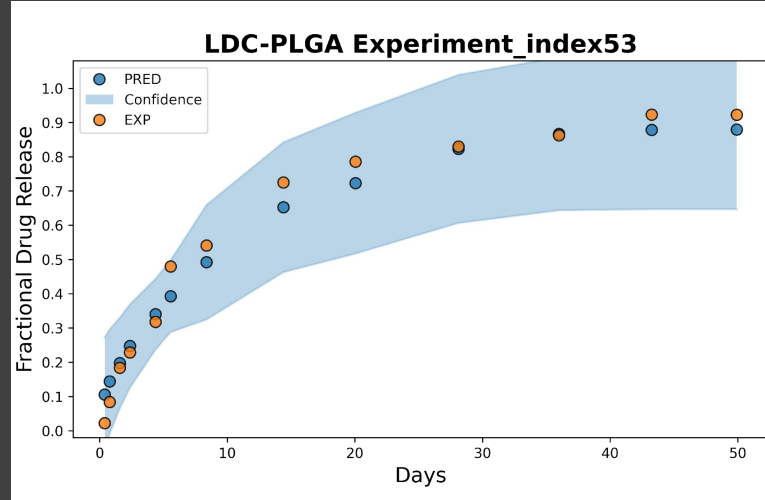
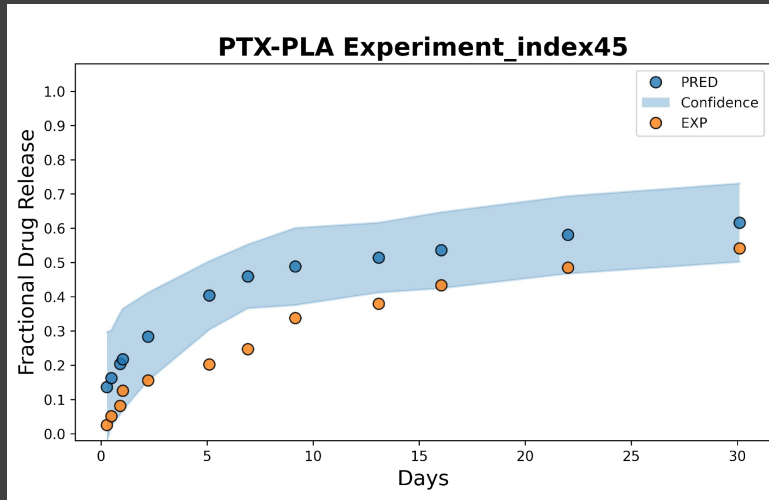
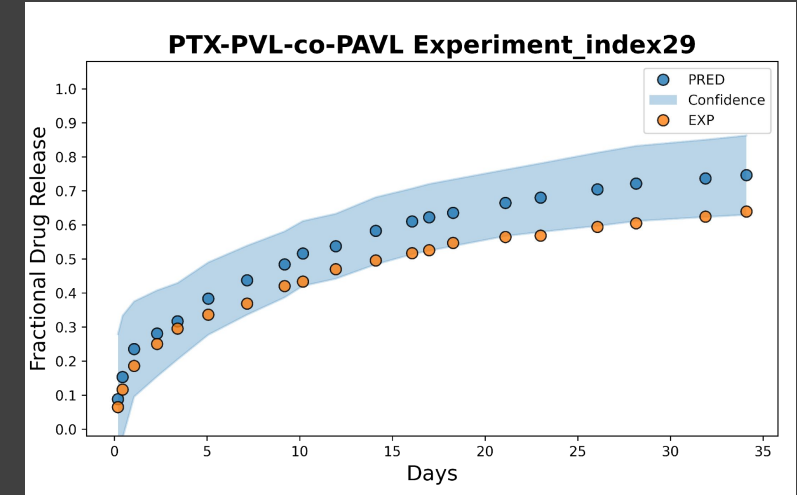
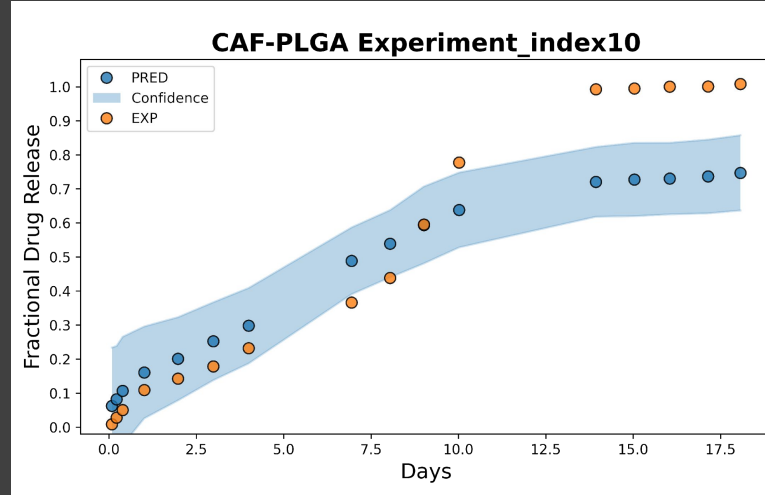
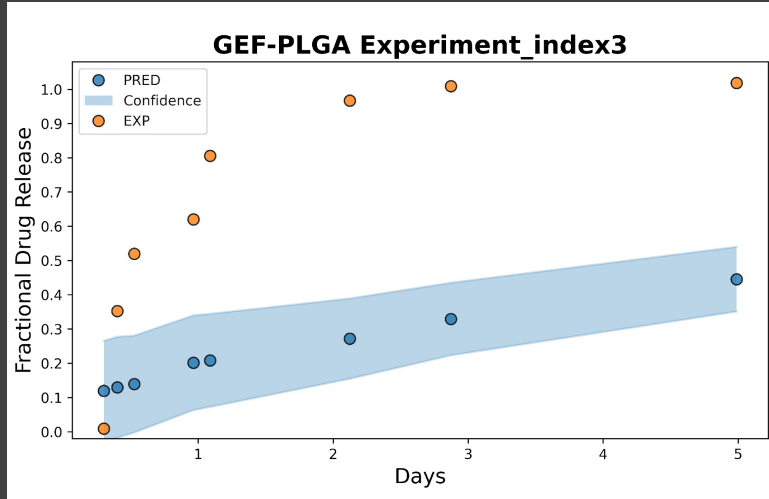
Random Forest Model Performance



Random Forest Model Performance



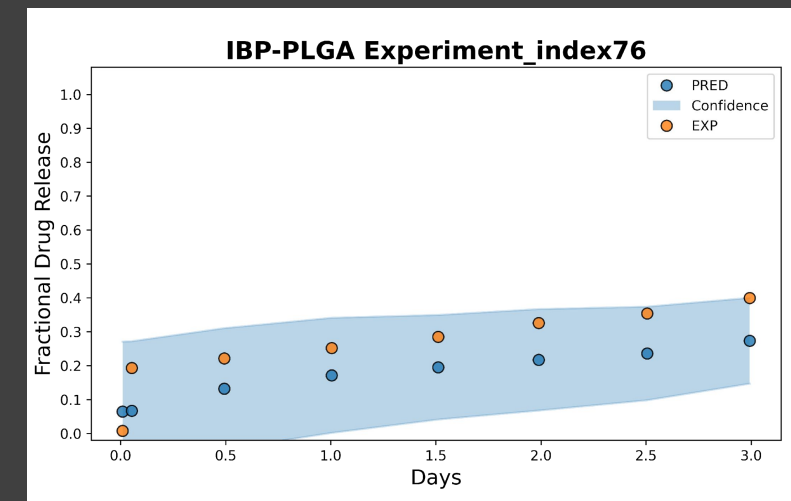
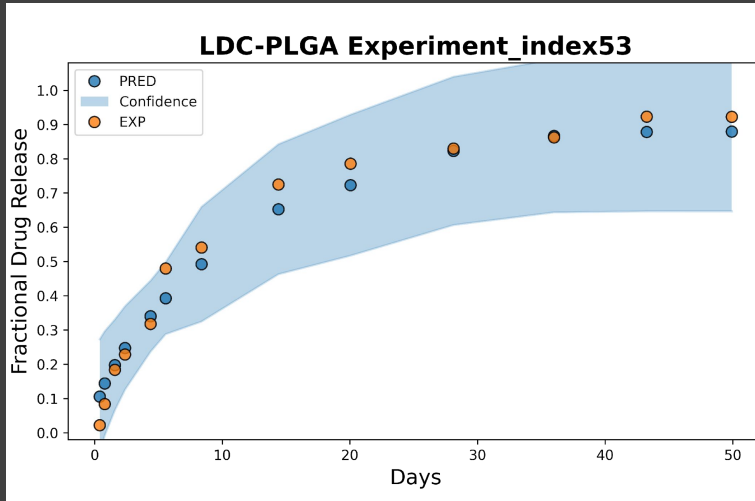
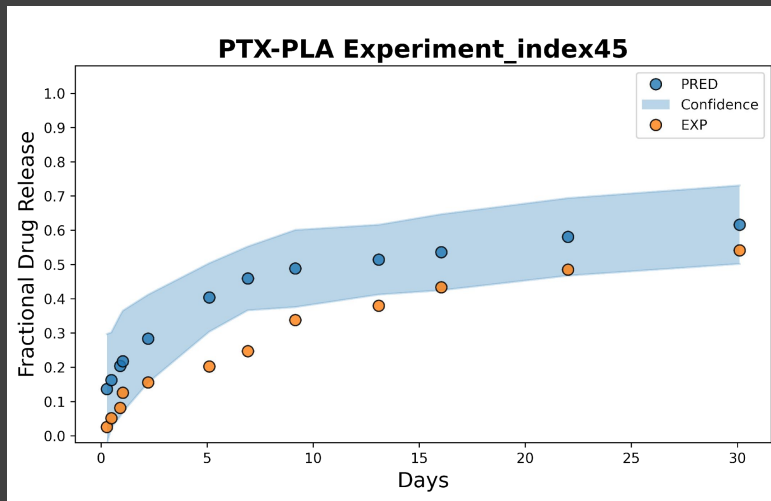
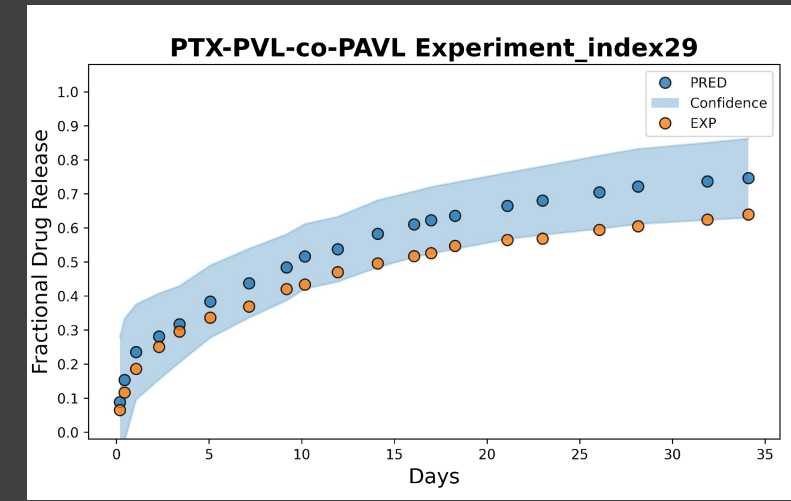
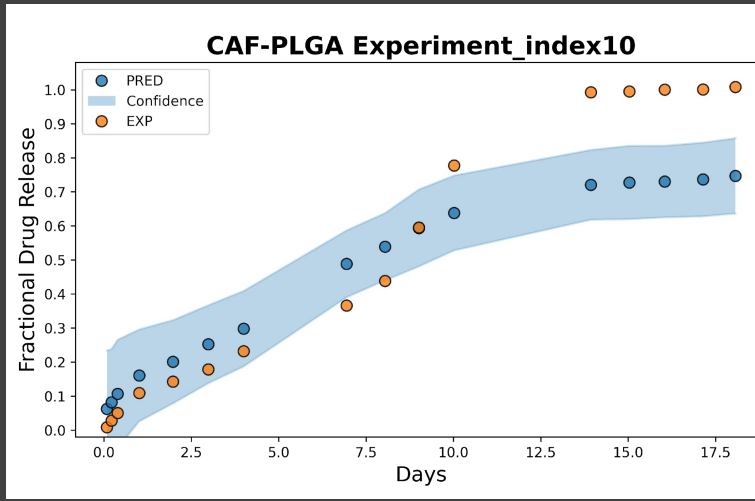
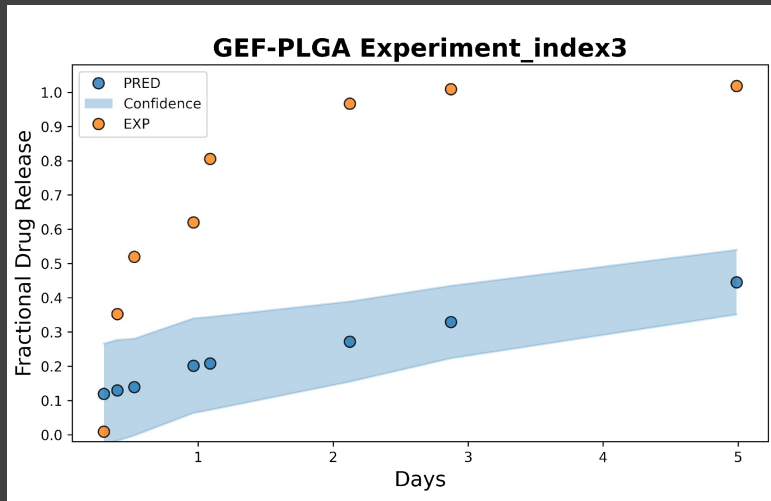
Random Forest Model Predictions



Confidence value calculations:

- Polimis et al, (2017), Confidence Intervals for Random Forests in Python, Journal of Open Source Software, 2(19), 124, doi:10.21105/joss.00124

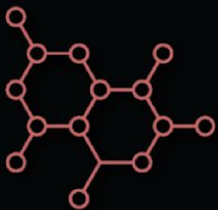
Random Forest Model Predictions



Confidence value calculations:

- Polimis et al, (2017), Confidence Intervals for Random Forests in Python, Journal of Open Source Software, 2(19), 124, doi:10.21105/joss.00124

Analysis of Gefitinib



Gefitinib

pKa: 5.4

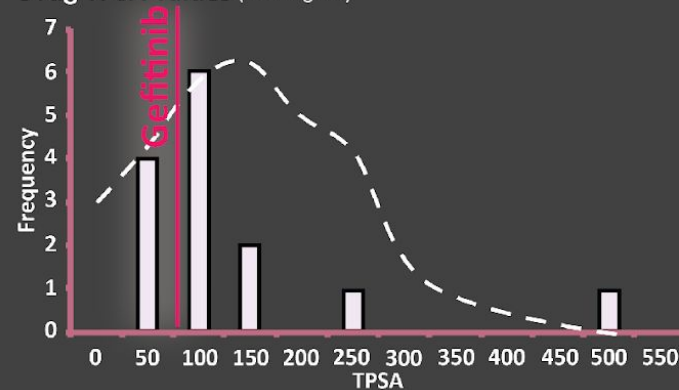
LogP: 3

T_m: 213 °C

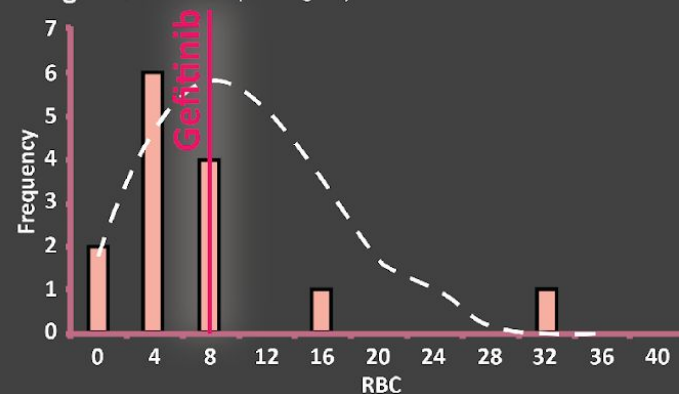
TPSA: 69 Å²

RBC: 8

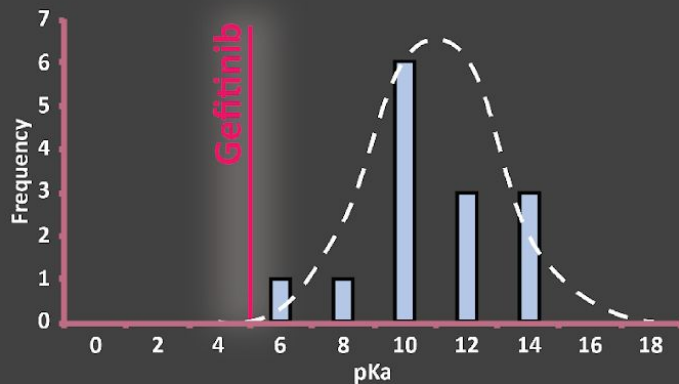
Drug TPSA values (training set)



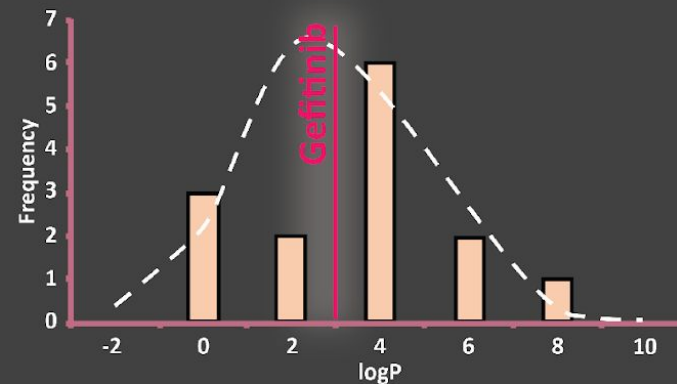
Drug RBC values (training set)



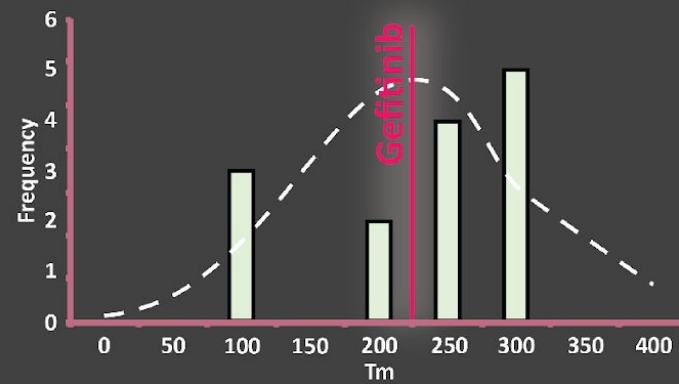
Drug pKa values (training set)



Drug logP values (training set)



Drug T_m values (training set)

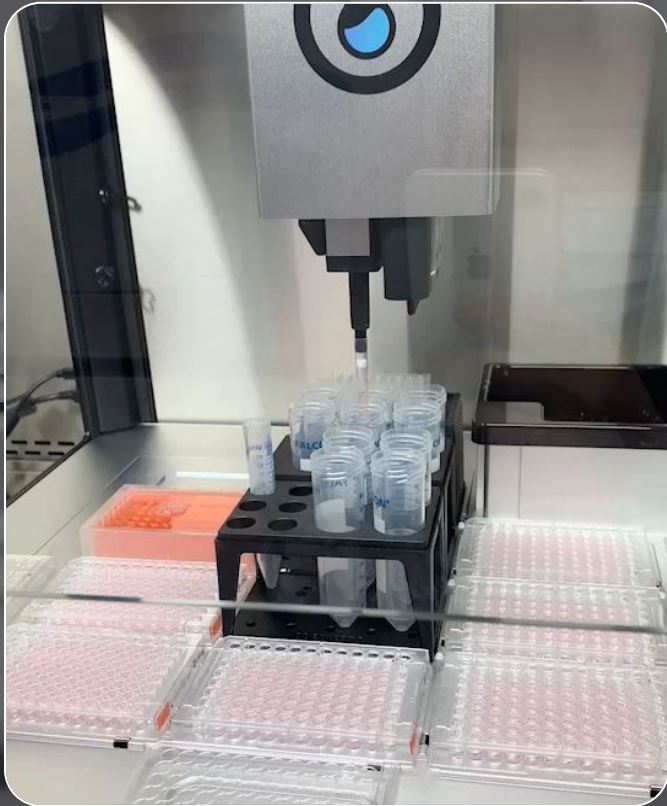


Extracting Data from Previous Studies...

Drug	Polymer Characteristics		Initial Polymer-Drug Ratio (w/w)	Formulation Characteristics		
	Molecular Weight (kDa)	LA:GA		Size (μm)	EE (%)	DL (%)
Lidocaine	N/A	50:50	22:1	14.4	N/A	4
Doxycycline	60	50:50	1.7:1	200 ± 1.52	24.7 ± 0.53	N/A
Etanidazole	40-75	50:50	N/A	422.3 ± 174.5	54.83 ± 0.30	5
Celecoxib	N/A	85:15	N/A	1.11 ± 0.08	51.48 ± 0.42	14.93 ± 0.12
Doxorubicin HCl	10	50:50	7.5:1	3.6 ± 0.4	74.9 ± 3.7	N/A
Simvastatin	50-75	85:15	11.5:1	3.963	96.74 ± 1.46	N/A
Budesonide	75	50:50	10:1	16.67 ± 1.24	74.28 ± 0.83	N/A
Ciprofloxacin	31	50:50	22:1	5.7	8.91 ± 2.90	0.40 ± 0.13
Dexamethasone acetate	19	75:25	N/A	1	N/A	2.5
Leuprolide acetate	N/A	50:50	4:1	N/A	97	20

Experimental Automation

High-throughput
HPLC Sample Preparation



In Vitro Drug
Combination Studies

Acknowledgments



Christine Allen

Pauric Bannigan

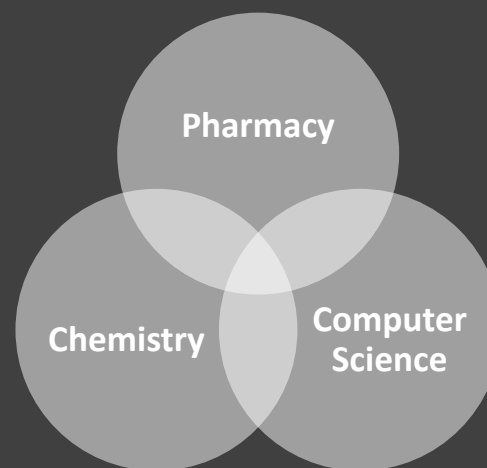


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Riley Hickman

Matteo Aldeghi

Florian Hase



UNIVERSITY OF TORONTO

