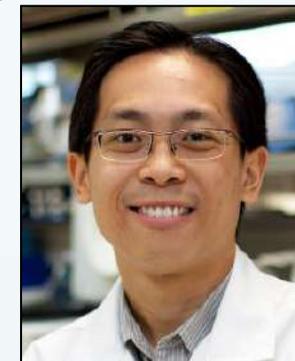
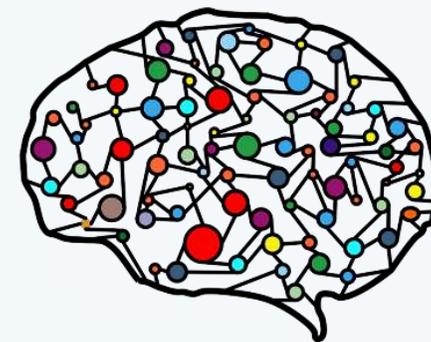
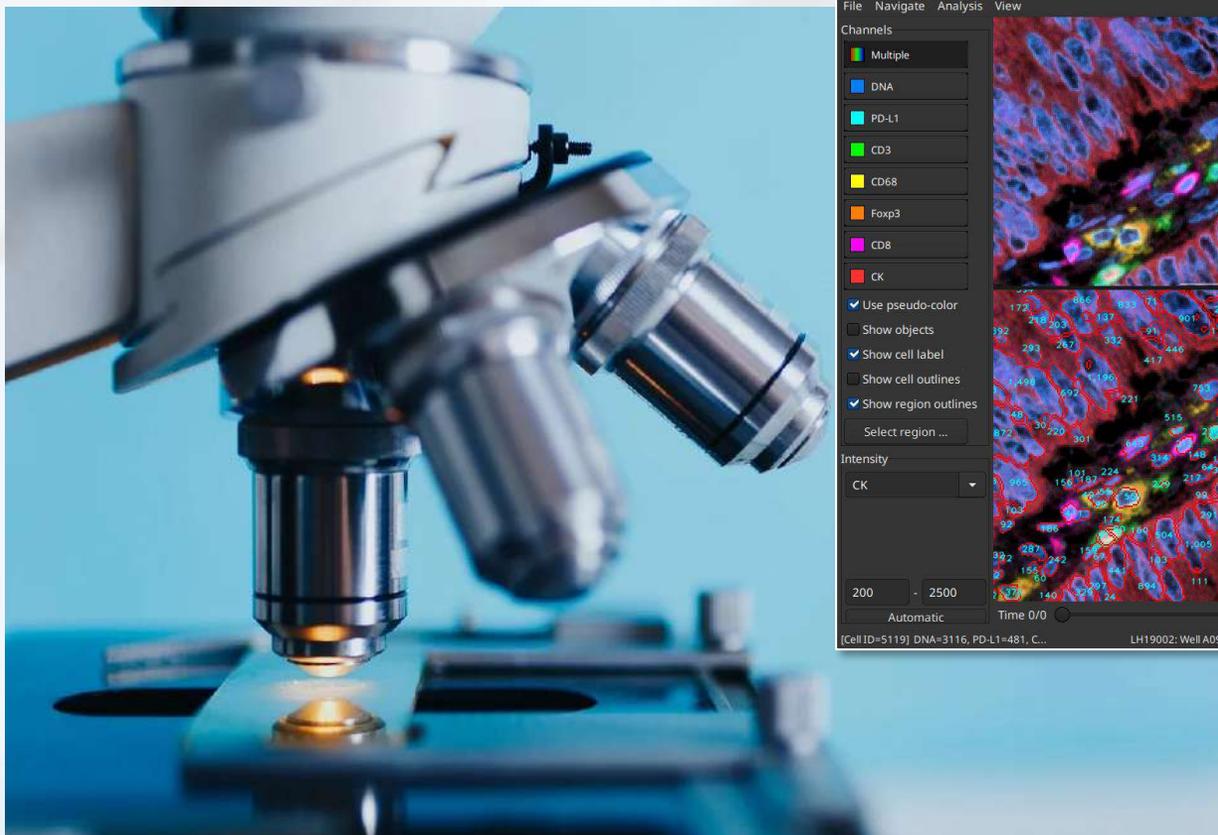




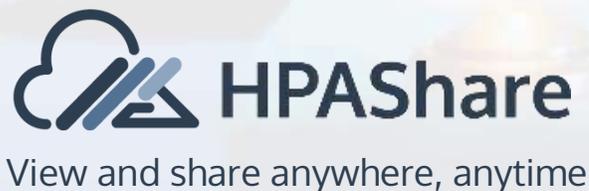
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Complex Cellular Phenotype Analysis

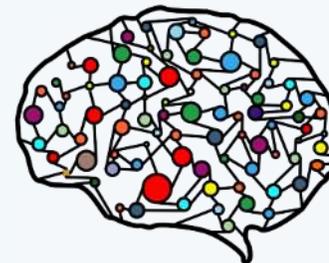


Loo Lit Hsin
Senior Principal Investigator
BII, A*STAR
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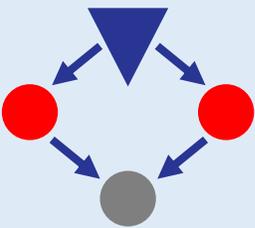


Complex Cellular Phenotype Analysis Group, BII



 **Phenotypic profiling and computational modeling**

- Chemical/drug safety or efficacy assessments
- High-throughput Image-based Phenotypic Profiling
- Machine learning, data analysis, and assay automation

 **ToxMAD**
Toxicity Mode-of-Action
Discovery Platform

Assess chemical safety based on mechanistic reasoning

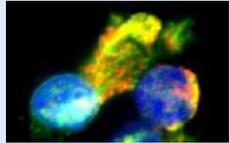
Bioimage Databases

 HPAShare
HPAScore **ImmunoAtlas**

View and share anywhere, anytime

Big bioimage data management, visualization, standardization, and analytics

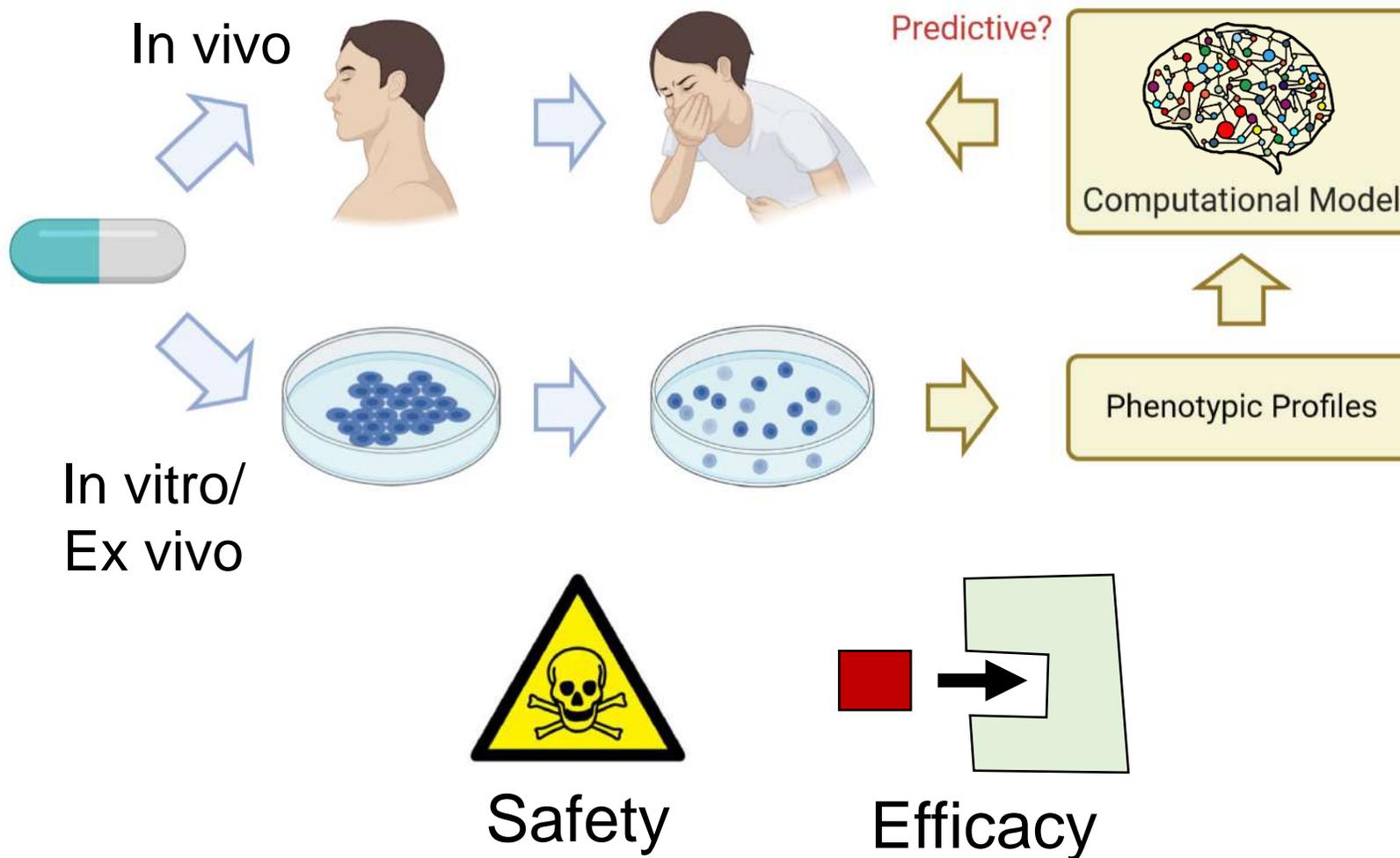
Digital Medicine for Cancer



Data-driven cancer treatment and care

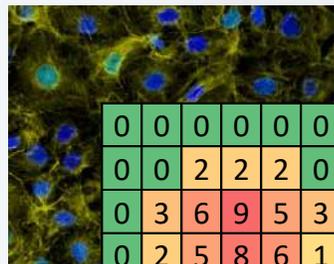


Building Predictive In vitro or Ex vivo Models for Chemical/Drug Safety or Efficacy Assessments

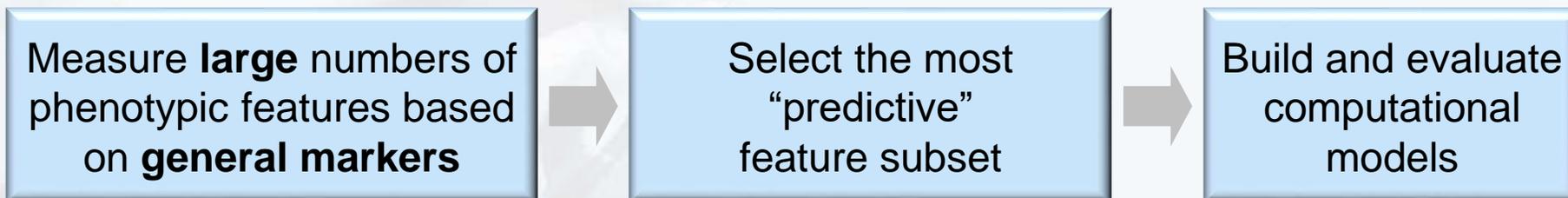




High-throughput Image-based Phenotypic Profiling (HIPP)



0	0	0	0	0	0	0	0
0	0	2	2	2	0	0	0
0	3	6	9	5	3	0	0
0	2	5	8	6	1	0	0
0	1	2	1	1	0	0	0
0	0	0	0	0	0	0	0



Conceptually similar to “RNA expression profiling”

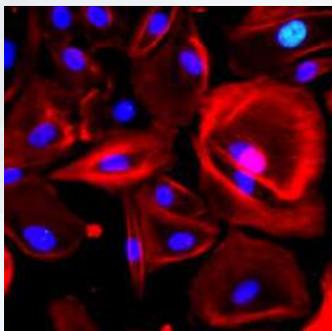
	Phenotypic Profiling	High-Content Analysis (HCA)
Number of measured phenotypic features	Very Large (~100s-1000s)	Small (<10)
How are the features designed and selected?	Automatically based on machine-learning algorithms	Manually based on known or expected mechanisms
What stains/markers are used?	General cell structures or biological processes	Specific structures or biological processes
Modes of action need to be defined a priori?	No	Yes
Can discover novel MoAs?	Yes	No



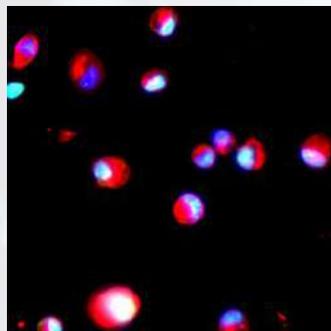
Proximal tubule cells exhibit distinct phenotypes when exposed to different chemicals

Non-PTC-Toxic
reference chemicals

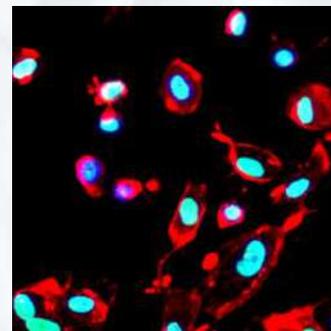
PTC-Toxic
reference chemicals



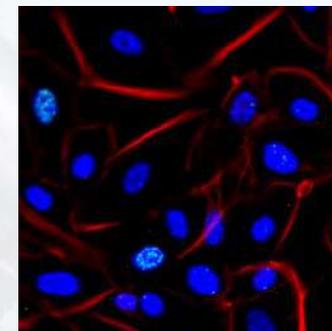
Dexamethasone



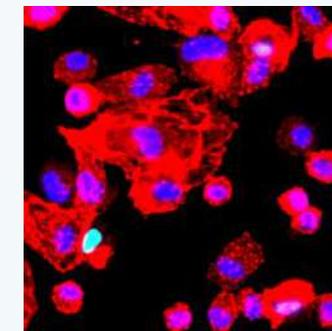
Arsenic Oxide



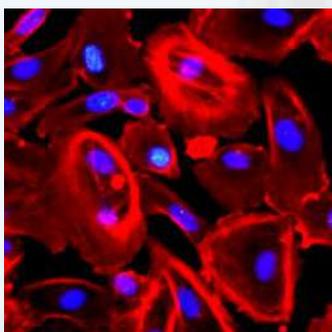
Cephalothin



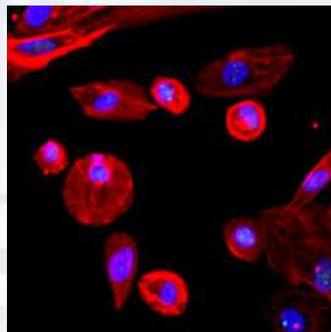
Paraquat



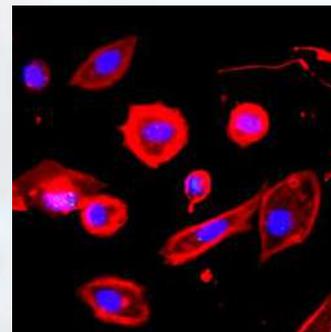
Potassium Dichromate



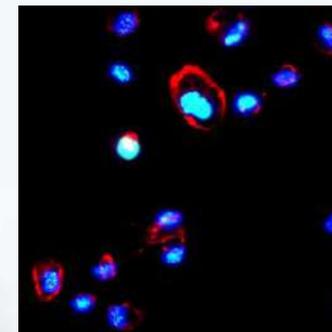
Water



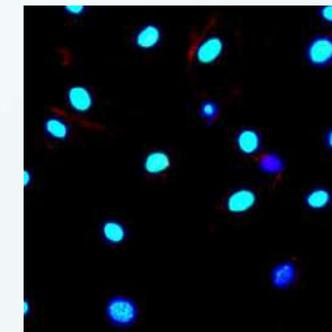
Tetracycline



Bismuth Oxide



Cadmium Chloride



Gold Chloride

Human PTCs were treated with 44 reference chemicals
and stained with DNA, Actin, RelA/gH2AX

In collaboration with Dani Zink, SIFBI



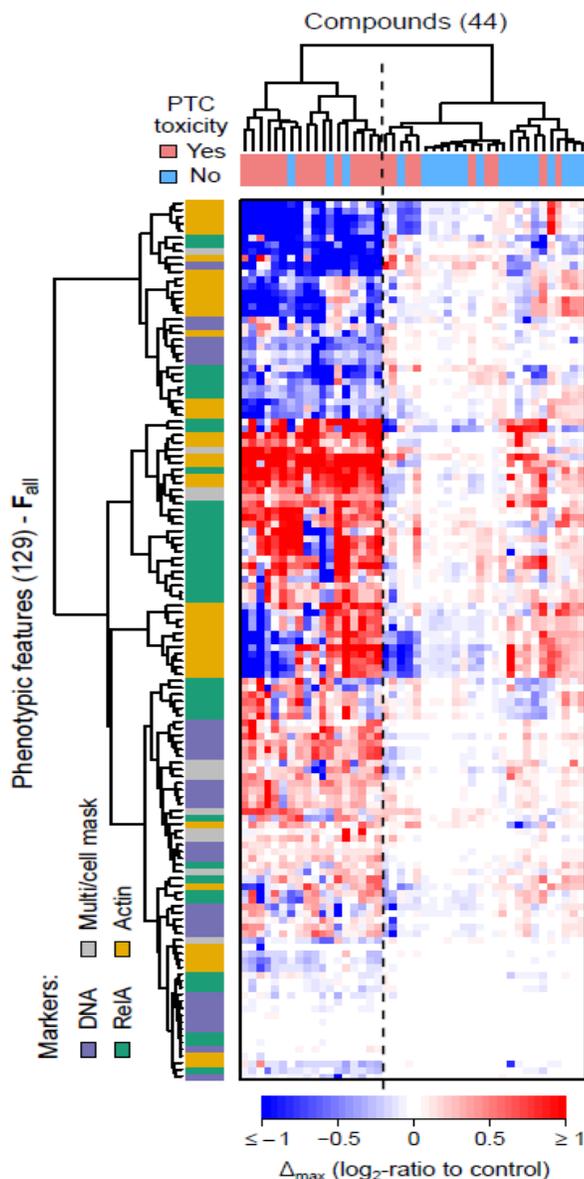
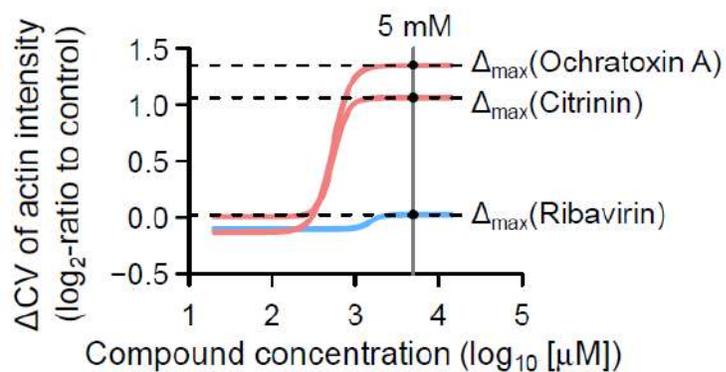
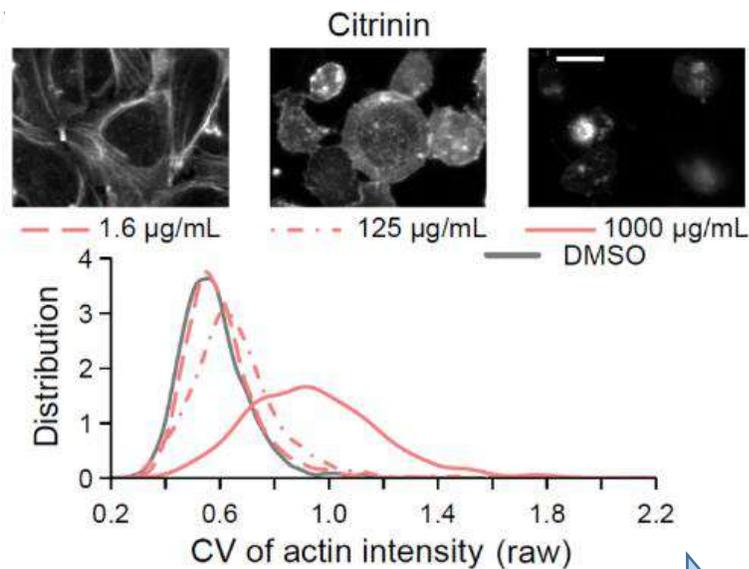
[Su et al., Arch Tox, 2016]



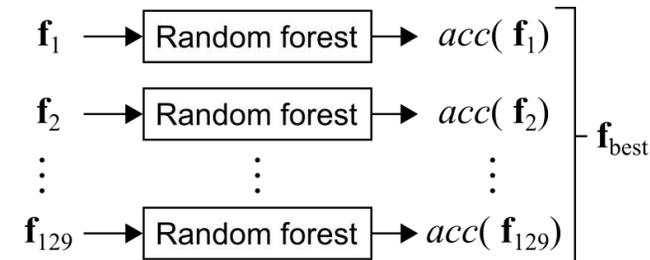
We performed phenotypic profiling ... 129 features



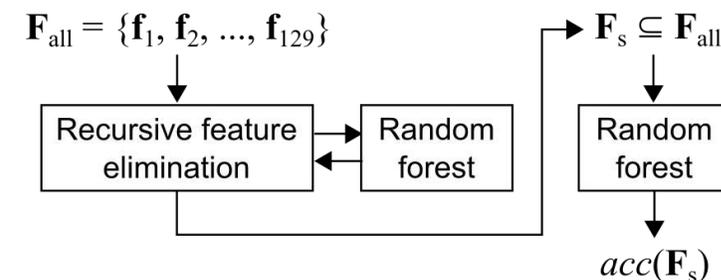
Ran Su



Single-feature classifiers



Multi-feature classifiers



5 features based on DNA, actin, and γ H2AX in HK-2 (89% accuracy)





CREATING GROWTH, ENHANCING LIVES



The first predictive and high-throughput in vitro model for nephrotoxicity



Ran Su is now an Associate Professor at Tianjin University, China



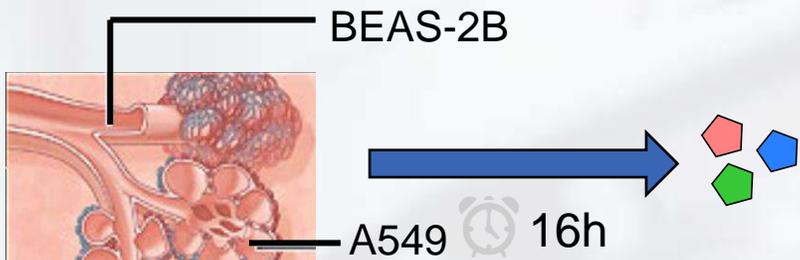
In 2016, Drs. Loo and Zink have become the first from Asia to win the Lush Prize (Science Award), UK



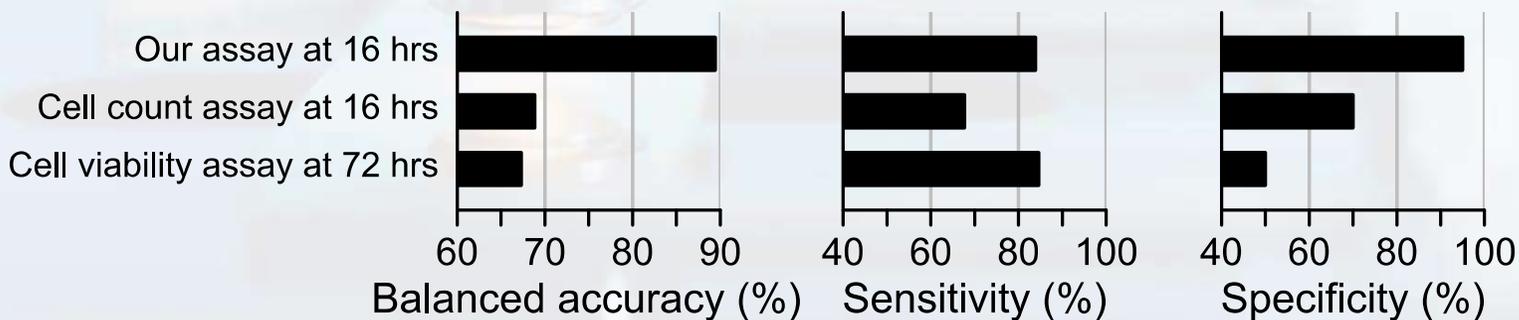
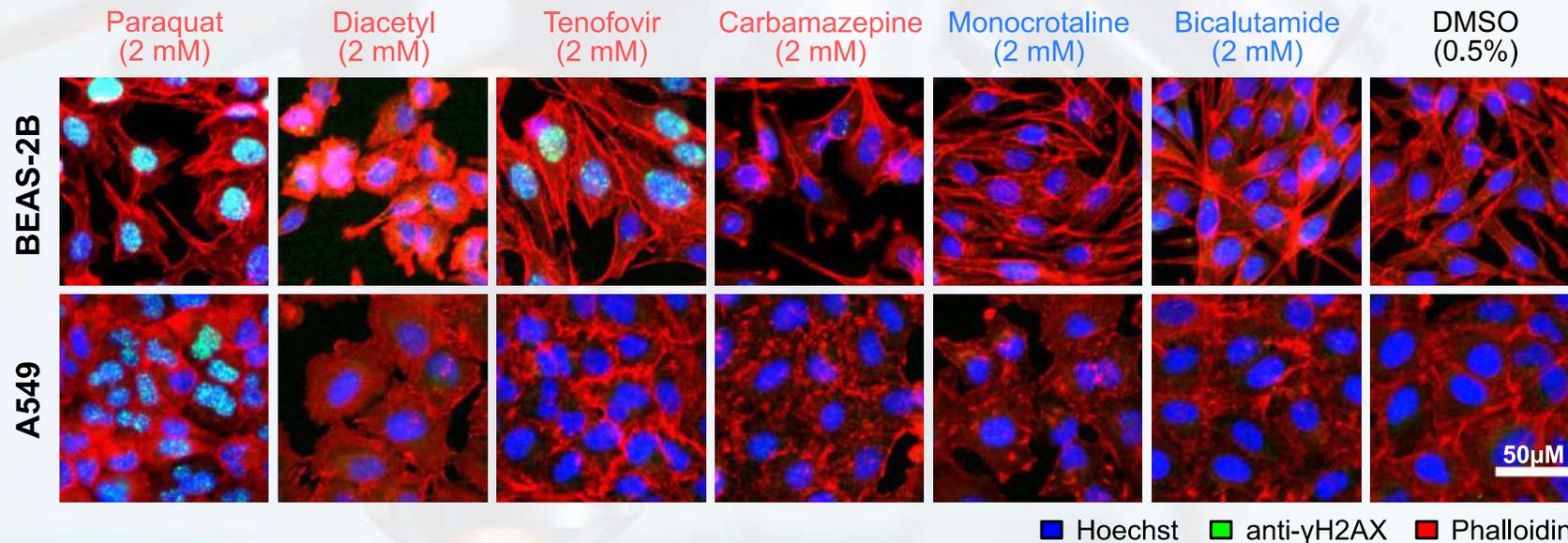


a

Building Lung Toxicity Model Using Phenotypic Profiling



HIPPTox lung assays are more accurate and specific than standard cell count or viability assays



Joey Lee



[Lee et al., Arch Tox, 2018]



Predicting direct hepatocyte toxicity in humans by combining high-throughput imaging of HepaRG cells and machine learning-based phenotypic profiling

Faezah Hussain¹ · Sreetama Basu² · Javen Jun Hao Heng¹ · Lit-Hsin Loo^{2,3} · Daniele Zink¹

Received: 12 November 2019 / Accepted: 5 May 2020 / Published online: 12 June 2020
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Abstract

Accurate prediction of drug- and chemical-induced hepatotoxicity remains to be a problem for pharmaceutical companies as well as other industries and regulators. The goal of the current study was to develop an in vitro/in silico method for the rapid and accurate prediction of drug- and chemical-induced hepatocyte injury in humans. HepaRG cells were employed for high-throughput imaging in combination with phenotypic profiling. A reference set of 69 drugs and chemicals was screened at a range of 7 concentrations, and the cellular response values were used for training a supervised classifier and for determining assay performance by using tenfold cross-validation. The results showed that the best performing phenotypic features were related to nuclear translocation of RELA (RELA proto-oncogene, NF- κ B subunit; also known as NF-kappa B p65), DNA organization, and the F-actin cytoskeleton. Using a subset of 30 phenotypic features, direct hepatocyte toxicity in humans could be predicted with a test sensitivity, specificity and balanced accuracy of 73%, 92%, and 83%, respectively. The method was applied to another set of 26 drugs and chemicals with unclear annotation and their hepatocyte toxicity in humans was predicted. The results also revealed that the identified discriminative phenotypic changes were related to cell death and cellular senescence. Whereas cell death-related endpoints are widely applied in in vitro toxicology, cellular senescence-related endpoints are not, although cellular senescence can be induced by various drugs and other small molecule compounds and plays an important role in liver injury and disease. These findings show how phenotypic profiling can reveal unexpected chemical-induced mechanisms in toxicology.



Example of how our models were used to assess fungicides



National Institute for Public Health and the Environment
Ministry of Health, Welfare and Sport

Chemical Research in Toxicology

pubs.acs.org/crt

Article

A Case Study with Triazole Fungicides to Explore Practical Application of Next-Generation Hazard Assessment Methods for Human Health

Leo T. M. van der Ven,* Emiel Rorije, R. Corinne Sprong, Daniele Zink, Remco Derr, Giel Hendriks, Lit-Hsin Loo, and Mirjam Luijten

Cite This: *Chem. Res. Toxicol.* 2020, 33, 834–848

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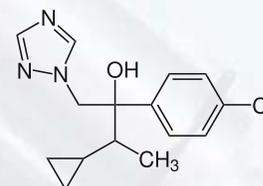
ACCESS |

Metrics & More

Article Recommendations

Supporting Information

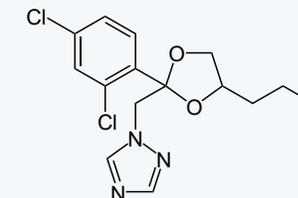
Three agricultural azole fungicides with unknown or unclear human toxicities



Cyproconazole



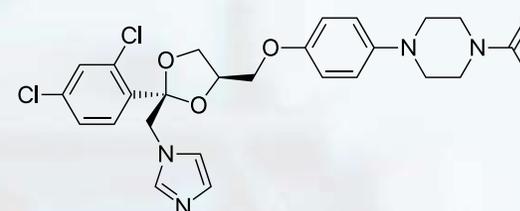
Flusilazole



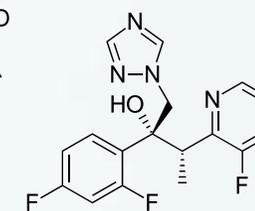
Propiconazole

What are their relative bioactivities?

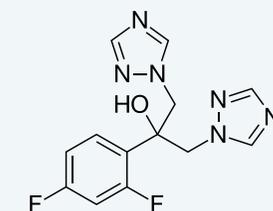
- To validate our model, we also add in 3 negative controls: azole drugs known to have low kidney effects



Ketoconazole



Voriconazole



Fluconazole



[van der Ven et al., Chem Res Tox, 2020]



Our ranking of the chemicals agrees with predictions based on other toxicological endpoints made by RIVM



National Institute for Public Health
and the Environment
Ministry of Health, Welfare and Sport

Table 4. Potency Comparison^a

Toxicological domain	Prediction Models (n)	Targeted endpoints in prediction models	Flu	Pro	Cyp	Ref
Cell functions	4	Dynamic cell state, mitochondrial function, metabolism, transporter functions	+	±	±	26-28
Endocrine perturbation, Mode of action screening	5	Endocrine profiling, nuclear receptors, G-protein-coupled receptors (aminergic/other)	±	±	±	26, 29
Endocrine perturbation, specific	8	Estrogen signalling pathway, estrogen receptor interaction, androgen receptor interaction, aromatase inhibition, steroidogenesis, thyroid hormone synthesis	++	+	+	30-37
Metabolic disorder	7	Glucose metabolism, adipocyte function, feeding behaviour	+	+	±	38, 39
Developmental toxicity	8	Developmental toxicity in rats, rabbits, zebrafish, <i>C.elegans</i> ; vascular development	++	++	+	40-46
(Developmental) Neurotoxicity	6	Ion channel, transporters, enzymes, neuronal network activity, neurobehavior	+	+	±	26, 47-49
Hepatotoxicity	2	Hypertrophy, liver injury, proliferative lesions, oxidative stress	+++	++	+	50, 51
Nephrotoxicity	1	Renal proximal tubular cell toxicity	+++	+++	+	17
Genotoxicity	5	DNA damage, gene mutations, chromosomal aberrations, p53 activation and oxidative stress	-	-	-	9-11, 14, 15, 19, 20, 52, 53
Carcinogenicity	2	Nuclear receptor activity, cancer hazard prioritization (hallmark genes)	+	++	+	54, 55
LOAEL embryotoxicity			10	35	20	
LOAEL hepatotoxicity			2.4	121	25.3	
LOAEL carcinogenicity			384	108	13.2	
Acceptable Daily Intake (ADI)			0.007	0.07	0.02	



[van der Ven et al., Chem Res Tox, 2020]



James Miller



APCRA

ACCELERATING THE PACE OF
CHEMICAL RISK ASSESSMENT

International case study on the use of in vitro bioactivity in risk-based chemical prioritization

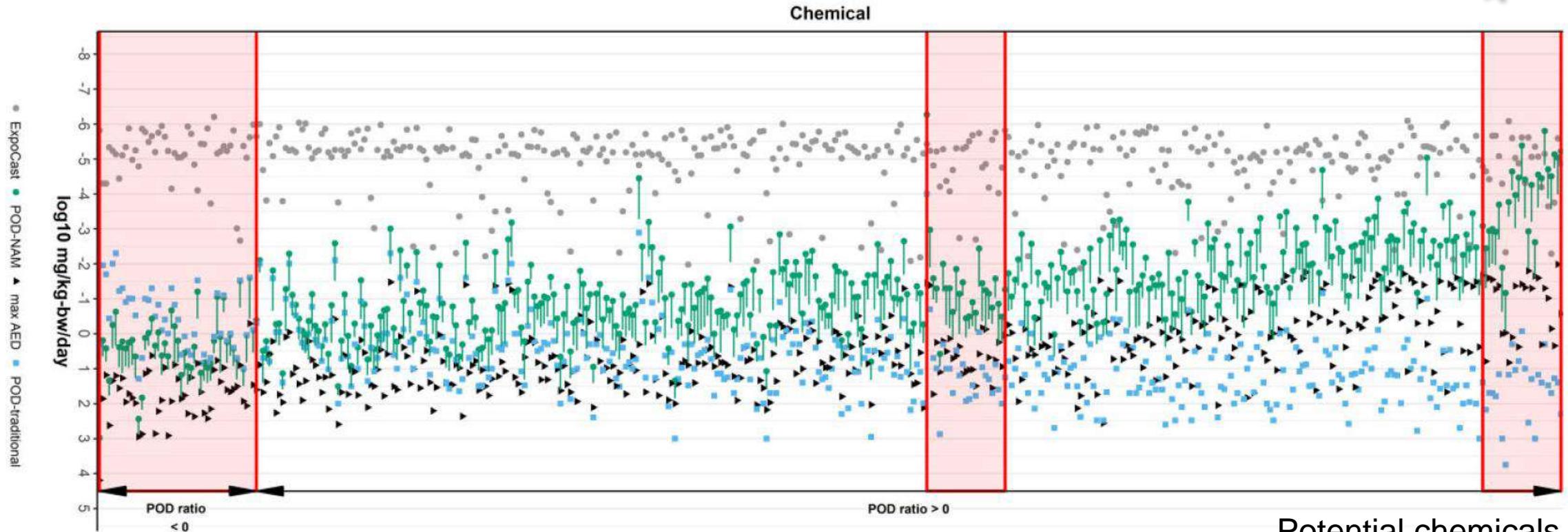


[Paul-Friedman et al., Tox Sci, 2020]

SOT Paper of the Year Award
(Honorable Mention), 2020



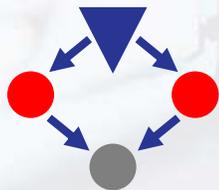
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Potential chemicals
of concern

- Predicted exposure level based on modeling
- PODs based on NAMs (ToxCast + HIPPTox)
- PODs based on traditional animal models





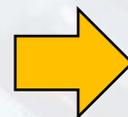
ToxMAD

Toxicity Mode-of-Action
Discovery Platform



Current safety assessments

Phenotypic endpoints



Future safety assessments

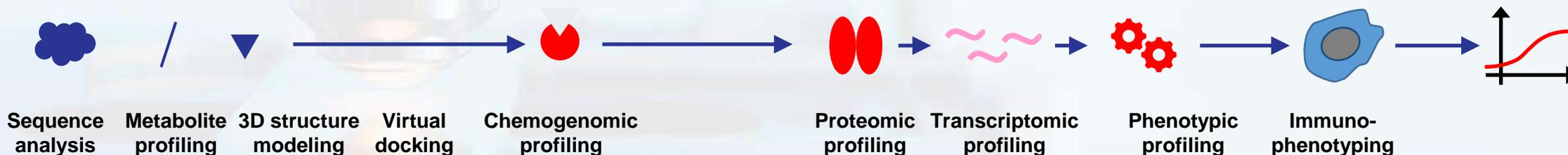
Mechanistic reasoning



ToxMAD is based on various A*STAR-developed *in vitro* and *in silico* technologies, and aim to rapidly and efficiently identify MoAs of chemicals (especially key molecular initiating events and cellular events leading to adverse outcomes)

Molecular initiating events

Key molecular/cellular events



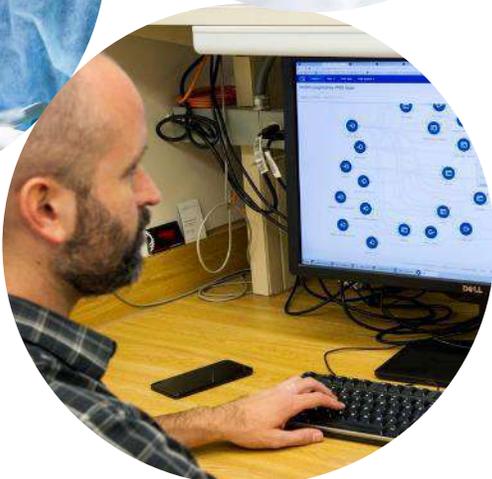


A platform to publish and share large biological images

How do tumors with poor outcomes look like?



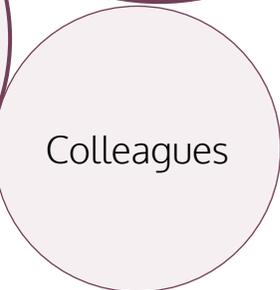
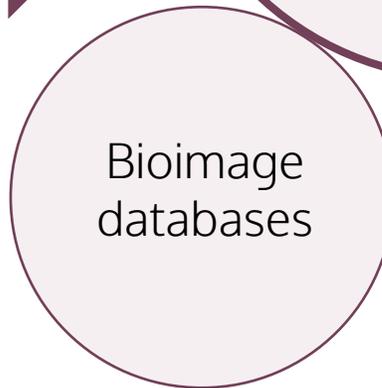
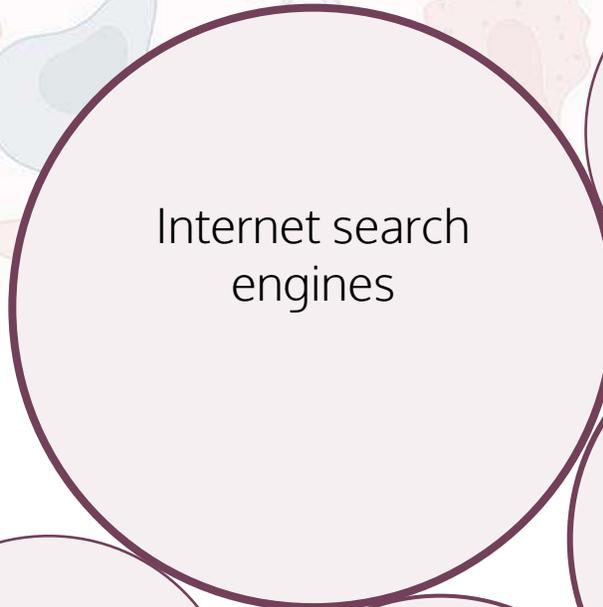
Does this patient have similar phenotypes as the published cases?



How do I train or benchmark my AI algorithms?



Is my staining correct?

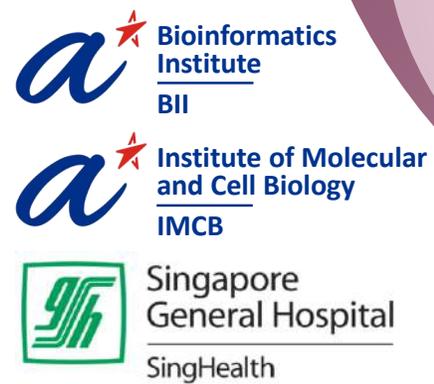
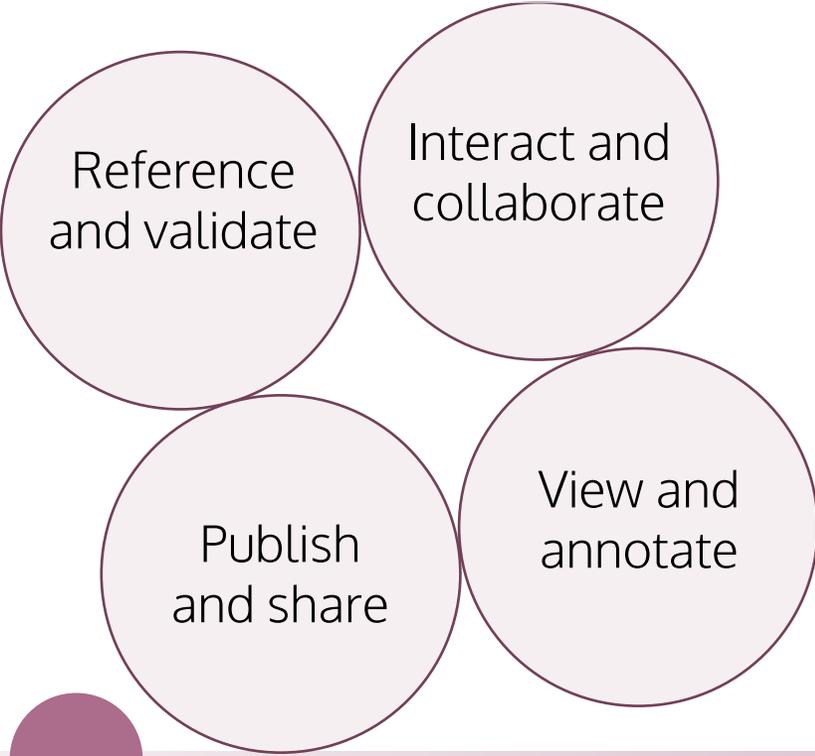




Introducing ImmunoAtlas ...

Standard or reference images of clinically-relevant immunofluorescence or immunotherapy markers

(Users include researchers, pathologists, standard workgroup, and companies)



ImmunoAtlas

Bioimage publishing and sharing portal (2021)

HistoPath Analytics (HPA) Platform

Online tissue image management and analysis platform (2019-2021)

cellXpress 2.0

Image processing engine in C++ (2011-2021)

☰ Samples collected for the case:

