NCATS Improving Health Through Smarter Science

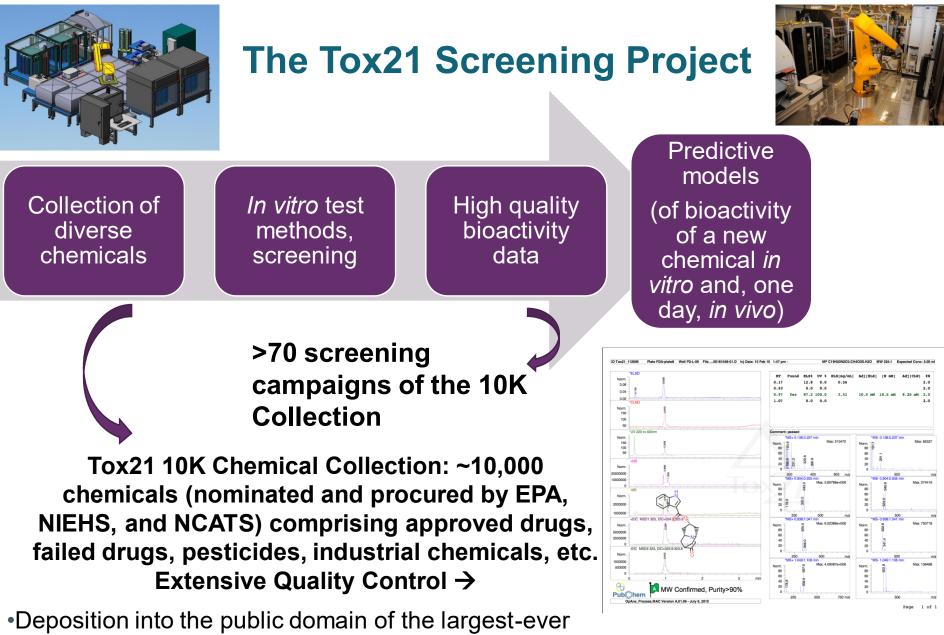
High-throughput screening to advance in vitro toxicology

Anton Simeonov, Ph.D.

Scientific Director, National Center for Advancing Translational Sciences (NCATS), National Institutes of Health (NIH)

Challenges in Public Health Protection in the 21st Century: New Methods, Omics and Novel Concepts in Toxicology November 15, 2021





toxicology dataset (100M datapoints), >100 publications.

•Using crowdsourcing to move from data to knowledge.



Tox21 10K Compound Library NTP

ToxCast I and II compounds

EPA

- Antimicrobial Registration Program
- Endocrine Disruptor • Screening Program
- OFCD Molecular • Screening Working Group
- FDA Drug Induced Liver • **Injury Project**

- NTP-studied compounds
- NTP nominations and related compounds
- NICEATM/ICCVAM reference compounds for regulatory tests
- External collaborators (e.g., Silent Spring Institute, U.S. Army Public Health Command)
- Formulated mixtures

NCATS

- Approved Drugs
- Investigational Drugs

- 88 internal standards. •
- Three library • replicates.
- Each sample arrayed in 15 doses.

Failed Drugs •

The Tox21 10K Compound Library: Collaborative Chemistry Advancing Toxicology

Ann M. Richard,* Ruili Huang, Suramya Waidyanatha, Paul Shinn, Bradley J. Collins, Inthirany Thillainadarajah, Christopher M. Grulke, Antony J. Williams, Ryan R. Lougee, Richard S. Judson, Keith A. Houck, Mahmoud Shobair, Chihae Yang, James F. Rathman, Adam Yasgar, Suzanne C. Fitzpatrick, Anton Simeonov, Russell S. Thomas, Kevin M. Crofton, Richard S. Paules, John R. Bucher, Christopher P. Austin, Robert J. Kavlock, and Raymond R. Tice



Cite This: Chem. Res. Toxicol. 2021, 34, 189-216

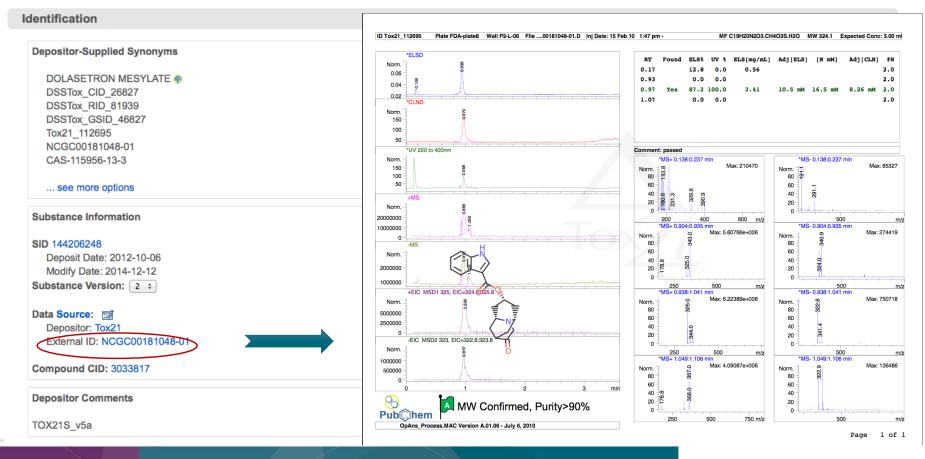




Chem Res Toxicol 2021 34(2):189-216

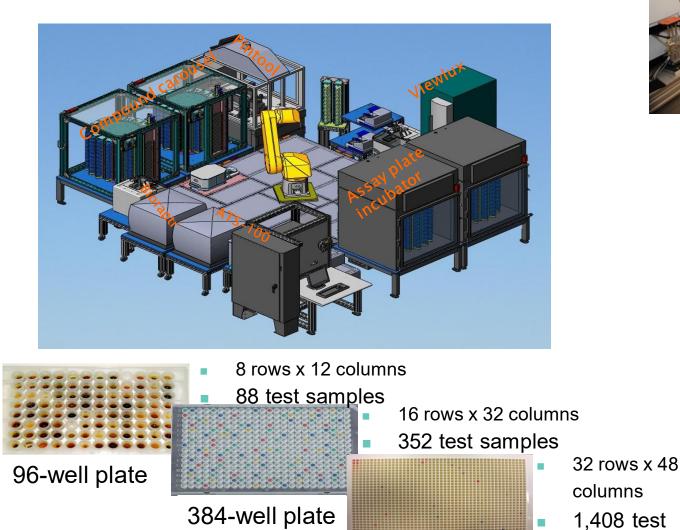
Entire-Library QC Project

- Multi-year undertaking using a range of LC-/GC-MS and NMR methods.
- >10,000 analytical chromatograms in PDF format available through PubChem: <u>http://www.ncbi.nlm.nih.gov/pcsubstance</u>

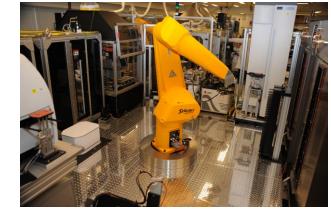


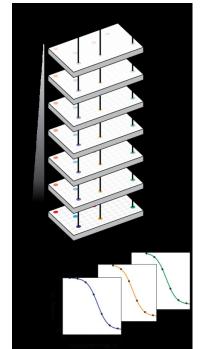


Tox21 Robot Platform



4 x 96-well plates





Dose-response-based screening Proc Natl Acad Sci 103:11473



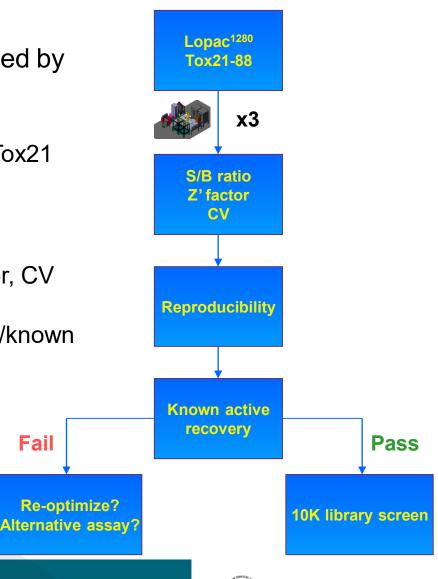
samples

1536-well plate

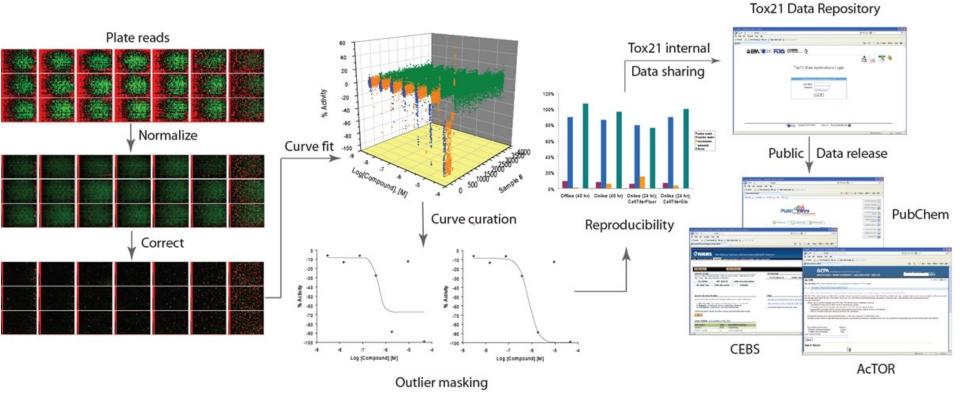
16 x 96-well plates

Assay Nomination and Validation Process

- Screening assay proposed and discussed by Assays and Pathways WG.
- Online validation on Tox21 Robot
 - Tox21 validation plate: Lopac¹²⁸⁰ + 88 Tox21 replicates
 - Triplicate runs
- Acceptance criteria
 - Performance metrics: S/B ratio, Z' factor, CV
 - Reproducibility
 - Ability to recover reference compounds/known actives
- Pass
 - Proceed to 10K library screening
- Fail
 - Go back to optimization?
 - Select alternative assay?



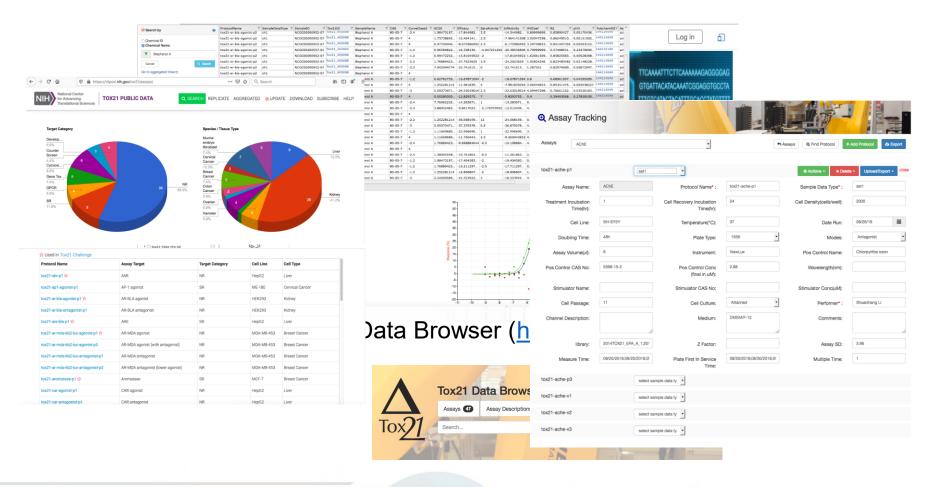
Informatics Analysis Process



NIH National Center for Advancing Translational Sciences

Tox21 Applications Gateway

http://tripod.nih.gov/tox/





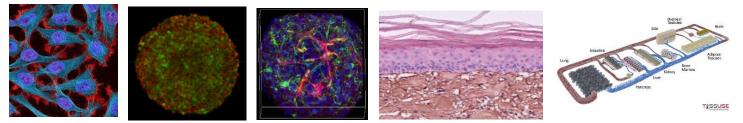
Predictive modeling with Tox21 data: collaborations with the scientific community

Chemical Research in To <u>xicology</u>		Cher Resea To <u>xico</u>	rch in	
pubs.acs.org/crt https://pubs.acs.org/crt?ref=pdf	Article	pubs.acs.org/crt		Article
Trade-off Predictivity and Explainability for Powered Predictive Toxicology: An in-Depth			h Learning with Prope ed Liver Injury	ty Augmentation for Predicting
Tox21 Data Sets	> Arch Toxicol. 2017 Dec;91(12):3885-3895.	. doi: 10.1007/s00204-017-	1995-9. Epub 2017 May 27.	n, Ruili Huang, and Junzhou Huang*
Leihong Wu, Ruili Huang, Igor V. Tetko, Zhonghua Xia, Joshua X				Read Online
Cite This: Chem. Res. Taxical. 2021, 34, 541–549	Why are most phospho blockers?	olipidosis indu	cers also hERG	Recommendations Supporting Information
ABSTRACT: Selecting a model in predictive toxicology often involves a trade-off between prediction performance and explainability: should we sacrifice the model performance and explainability:	Svetoslav Slavov ¹ , Iva Stoyanov Menghang Xia ² , Richard Beger	1	² , Ruili Huang ² ,	biLl Production Performance
stut toxi 65 12.1 syst	Affiliations + expand PMID: 28551711 DOI: 10.1007/			A Section 508-conformant HTML version of this article is available at https://doi.org/10.1289/EHP5580.
rep res Scale Multitask Deep Learning QSAR Model	년 Full text links " Cit			for Androgen Receptor Activity menico Alberga, ⁶ Vinicius M. Alves, ^{7,8} Patrik L. Andersson, ⁸ ¹² Emilio Benfenati, ¹⁴ Barun Bhhatarai, ¹⁵ Scott Boyer, ¹⁶
per in Alexey V. Zakharov,* [©] Tongan Zhao, Dac-Trung Nguyen, Tyler Pe Ruili Huang, Noel Southall, and Anton Simeonov	Abstract	FDA U.S. FOOD & DRUG		¹⁸ , ¹⁸ Alfonso T. García-Sosa, ¹⁹ Paola Gramatica, ¹⁵ vrvath, ²¹ Xin Hu, ²² Ruili Huang, ²² Nina Jeliazkova, ²³ ¹ Giuseppe F. Mangiatordi, ⁶⁴⁺ Uko Maran, ¹⁹
National Center for Advancing Translational Sciences (NCATS), National Institute Rockville, Maryland 20850, United States		National Center for oxicological Research		yen, ²² Orazio Nicolotti, ⁶ Nikolai G. Nikolov, ¹³ l Pogodin, ²⁶ Vladimir Poroikov, ²⁶ Xianliang Qiao, ¹⁷
Supporting Information	Recent reports have noted that a gene (hERG) ion channel also inc	chief Scientist	human Ether-à-go-go related a hypothesis explaining why	1 Rupakheti, ^{24,28} Sugunadevi Sakkiah, ²⁰ aj, ²⁰ Imran Shah, ¹ Sulev Sild, ¹⁹ Lixia Sun, ²⁹ shini, ¹² Weida Tong, ²⁰ Daniela Trisciuzzi, ⁶
ABSTRACT: Advances in the development of high-throughput screening and automated chemistry have rapidly accelerated the production of chemical and biological data, much of them freely accessible through	ment DLD inducers are also hEDC	Publication Award for Data Methods/ analysis/Study Design	indertaken with data sets pmpounds assayed for PLD	ek, ²¹ Zhongyu Wang, ¹⁷ Eva B. Wedebye, ¹³ eng, ⁹ and Richard S. Judson ¹
International obligitation and motogeneric such as ChEIMBL and PubChem. Here, we explore how to use this comprehensive mapping of chemical biology space to support the development of large-scale quantitative structure-activity relationship (QSAR) models. We propose a new deep learning consensus architecture (DLCA) that combines consensus and multitask deep learning approaches together to generate large-scale QSAR models. This method improves knowledge transfer across different target/assays while also integrating contributions from models based on different descriptors. The proposed approach was validated and compared with proteochemometrics, multitask deep learning, and Random Forest methods paired with various	3D-1	3D-SDAR Modeling Algorithm for Phospholipidosis Team 2018	riangle Park, North Carolina, USA s, Inc., Morrisville, North Carolina, USA I Interagency Center for the Evaluation of Alternat ungle Park, North Carolina, USA	nt, U.S. Environmental Protection Agency (U.S. EPA), Research Triangle ive Toxicological Methods (NICEATM), National Institute of Environmental alhrung, Landnutzang und Umwelt, Department für Biowissenschaftliche lend University of Goiás, Goiánia, Brazil I, Chapel Hill, North Carolina, USA



Increasing the predictivity of *in vitro* assays: a continuum of 3D models of healthy and diseased tissues

2D Spheroids Organoids Printed Tissues Organ-on-a-chip



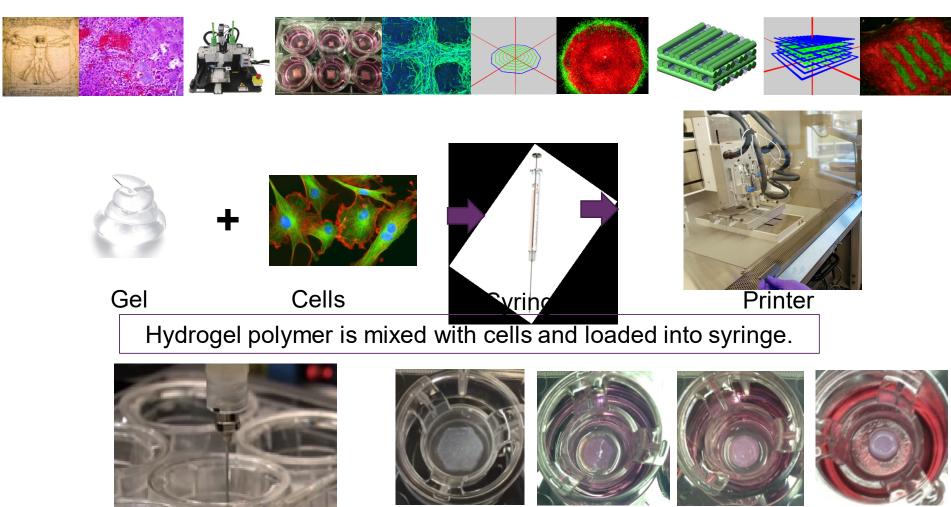
Physiological complexity







3D Tissue Bioprinting



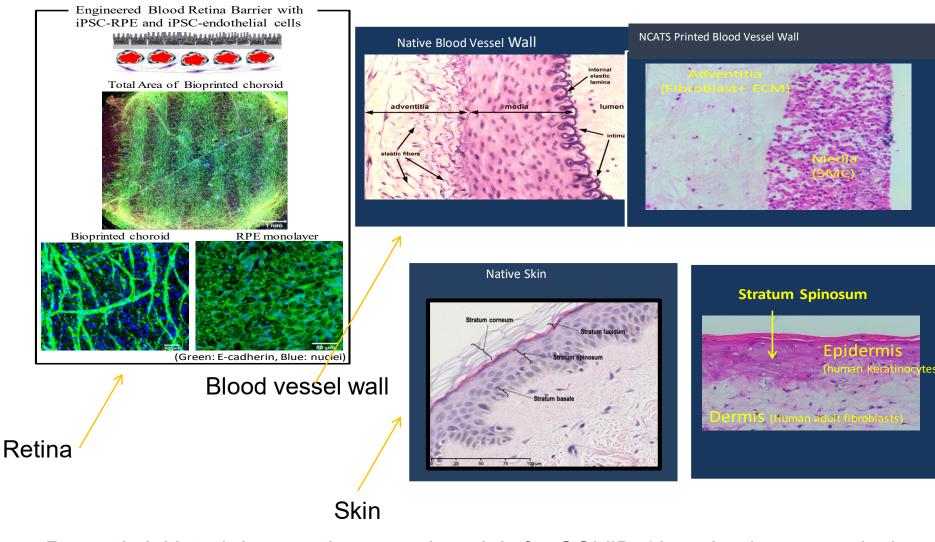
Printed construct 1 day

1 week

2 weeks

The printer "3D prints" the cell/gel mixture in a layer by layer approach. The printed construct is incubated to allow the cells to form a tissue, and to enable proper cell differentiation.

Examples of 3D Bioprinting Projects

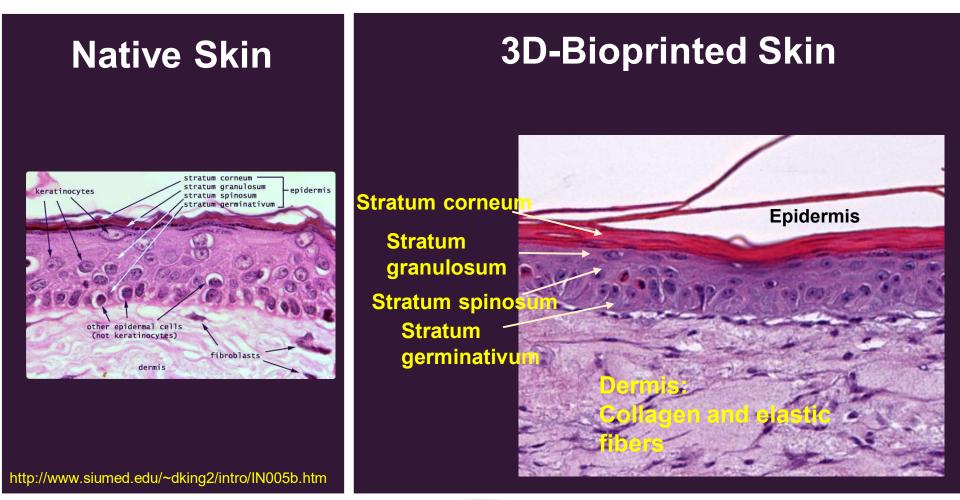


Recently initiated: lung and neuronal models for COVID-19 and pain, respectively.



National Center for Advancing Translational Sciences

Skin biofabrication

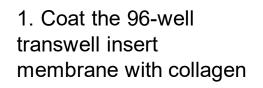




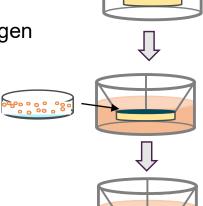
National Center for Advancing Translational Sciences

Generation of bioprinted skin tissues to test for irritants and sensitizers

Reconstructed human epidermis (RhE)

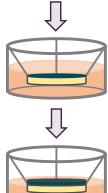


2. Add keratinocytes



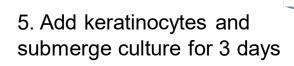
3. Submerge culture for 3 days

4. Air-liquid interface culture for 8 days



2. Bioprint fibroblast bioink to a 3-layer U shape on bottom side of 96-well transwell insert membrane

4. Submerge bioprinted tissue in medium for 7 days

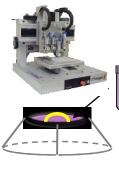


6. Air-liquid interface culture for 8 days

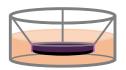
Z Wei, et al., Frontiers in Bioengineering and Biotechnology (2020) 🤇

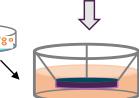
Full thickness skin tissue (FTS)

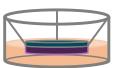
1. Suspend fibroblasts in bioprinting gel



3. Add bioprinting gel to cover the U shape



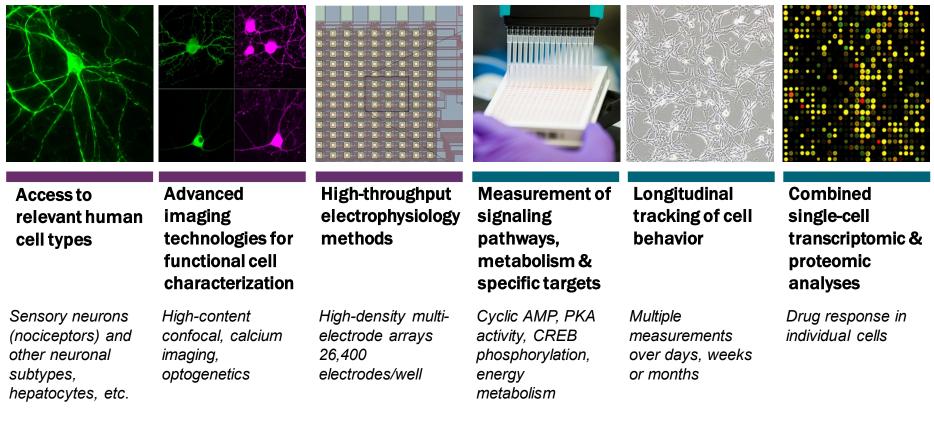






Enabling advanced 3D models through stem cell technologies

NCATS Stem Cell Translation Laboratory:



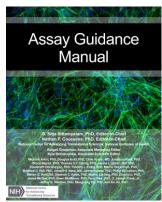
Example: a cheap 4-molecule cocktail that dramatically enhances survival of hPSCs and facilitates single-cell cloning, cell thawing and passaging, and CRISPR applications. Chen, *et al*, *Nature Methods* (2021)



Where do I go for more information about assay development and screening?



Sharing internal know-how: Assay Guidance Manual (47 chapters/ 1,338 printed pages)



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Drofaco

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Instrumentation	2 Chapters
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Glossary of Quantitative Biology Terms	1 Chapter

Website: https://ncats.nih.gov/expertise/preclinical/agm

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Linkedin: www.linkedin.com/groups/7437344

Linked in

Facebook: <u>www.facebook.com/assayguide</u>

https://ncats.nih.gov/agm-video

August 7th Videos

- 1. Austin, CP: Welcome to the Assay Guidance Manual (AGM) Workshop
- 2. Coussens, NP: Strategies for Assay Selection & Robust Biochemical Assays
- 3. Riss, T: Treating Cells as Reagents to Design Reproducible Screening Assays
- 4. Trask, OJ: Assay Development Considerations for High Content Imaging
- 5. Auld, DS: Studies in Mechanisms and Methods in Assay Interferences
- 6. Dahlin, JL: Assay Interference by Chemical Reactivity
- 7. Chung, TDY: Basic Assay Statistics, Data Analysis & Rules of Thumb
- 8. Devanarayan, V: Reproducibility & Differentiability of Potency Results
- 9. Sittampalam, GS: Avoiding Artifacts & Interferences in Assay Operations

March 26-27th Videos

- 1. Austin, CP: Welcome to the Assay Guidance Manual (AGM) Workshop
- 2. Coussens, NP: Robust Assays Define Success in Preclinical Research
- 3. Lal-Nag, M: Target Identification & Validation in Translational Discovery
- 4. Foley, TL: Development & Validation of Cell-Based and Biochemical Assays
- 5. Riss, T: Treating Cells as Reagents to Design Reproducible Screening Assays
- 6. Trask, OJ: Assay Development for HCS & Best Practices for 3D HCS
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- 14. Weidner, JR: Assay Operations: Keeping Assays Robust and Reproducible





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