# Al in Precision Health

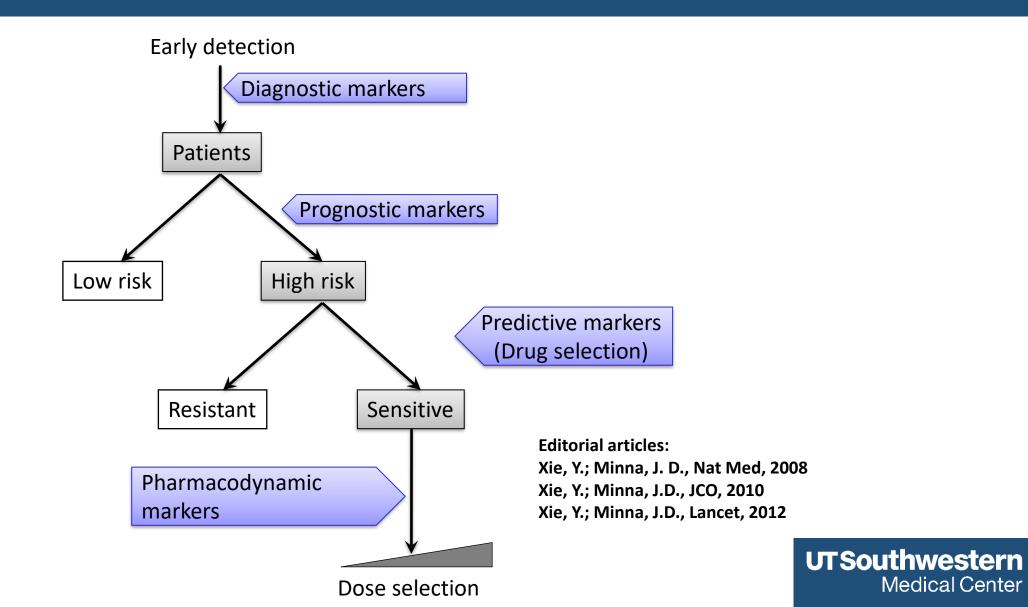
Yang Xie, Ph.D, M.D.

Professor & Associate Dean of Data Science Raymond D. and Patsy R. Nasher Distinguished Chair in Cancer Research Chair, Department of Health Data Science and Biostatistics Peter O'Donnell Jr. School of Public Health UT Southwestern Medical Center at Dallas

> 2025 UT System AI Symposium in Health Care Houston, Texas May 15, 2025

UT Southwestern Medical Center

# AI Can Identify Different Types of Biomarkers for Precision Health



Medical Center

# Genomic based Precision Health

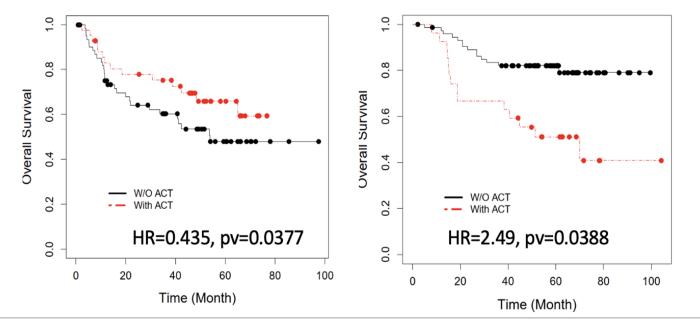
#### Genomic signature for treatment response prediction (Adjuvant Genomics)

PREDICTIVE BIOMARKERS AND PERSONALIZED MEDICINE | MARCH 14 2013

#### A 12-Gene Set Predicts Survival Benefits from Adjuvant Chemotherapy in Non–Small Cell Lung Cancer Patients 🔗

Hao Tang; Guanghua Xiao; Carmen Behrens; Joan Schiller; Jeffrey Allen; Chi-Wan Chow; Milind Suraokar; Alejandro Corvalan; Jianhua Mao; Michael A. White; Ignacio I. Wistuba; John D. Minna; Yang Xie 🐱

#### **Predicted benefit group**



Predictive performance in MD Anderson dataset (n=207) (Similar validation in three independent cohorts) PRECISION MEDICINE AND IMAGING | JANUARY 03 2019

#### Validation of the 12-gene Predictive Signature for Adjuvant Chemotherapy Response in Lung Cancer 🔗

Yang Xie ≥ ; Wei Lu; Shidan Wang; Ximing Tang; Hao Tang; Yunyun Zhou; Cesar Moran; Carmen Behrens; Jack A. Roth; Qinghua Zhou; David H. Johnson; Stephen G. Swisher ⓑ ; John V. Heymach ⓑ ; Vassiliki A. Papadimitrakopoulou; Guanghua Xiao ⓑ ; John D. Minna ⓑ ; Ignacio I. Wistuba ≥

#### Precision Medicine - For each patient we examine biologically relevant signals to derive a personalized nucleic acid fingeror

relevant signals to derive a personalized nucleic acid fingerprint

Artificial Intelligence - Using Machine Learning approaches, our scientific team examines each patient's personalized fingerprint and derives predictions regarding both the prognosis and likely benefit of chemotherapy for that patient

SAVING TIME SAVES LIVES

ADJUVANT GENOMICS APPROACH TO NSCLC DIAGNOSIS

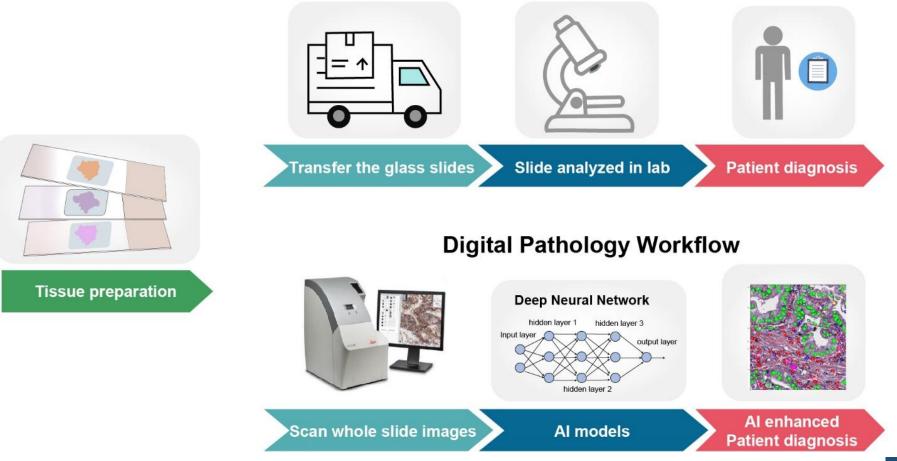
A clinical assay has been produced and is under clinical validations

#### UT Southwestern Medical Center

#### Predicted non-benefit group

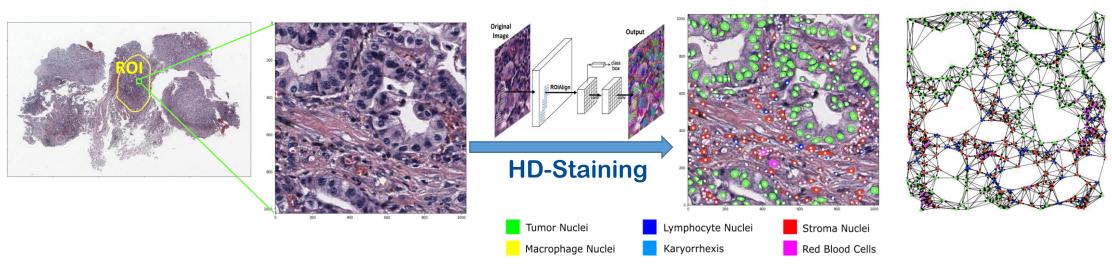
# Digital Pathology

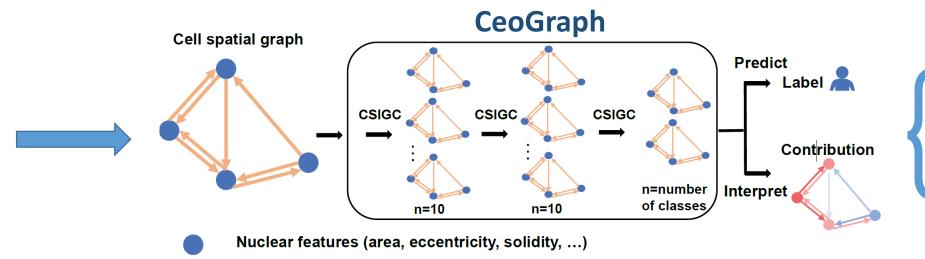
#### **Traditional Pathology Workflow**



UT Southwestern Medical Center

## Al Models for Digital Pathology



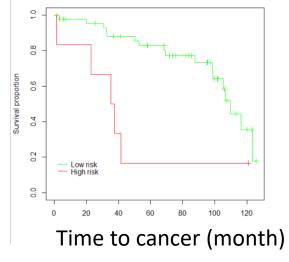


→ Cell-cell interactions (interaction type, closeness, parallelism)

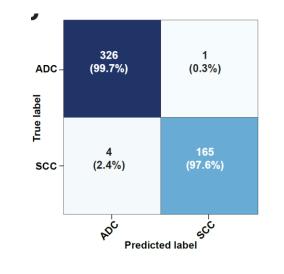
- Faster diagnosis
- More Accurate
- Cost-effective

# AI Models for Digital Pathology

#### 1. Early detection and risk stratification 2. Disease diagnosis and subtyping



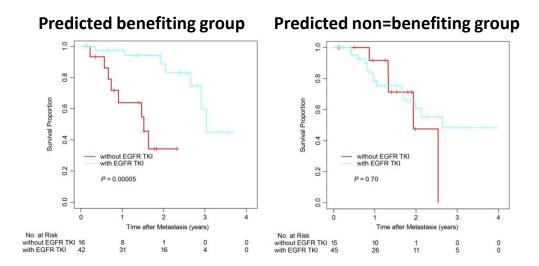
Predicted high-risk group
Predicted low-risk group



#### Disease areas:

- Lung cancer
- Breast cancer
- Head & neck cancer
- Oral potentially malignant disorders
- Rhabdomyosarcoma
- Liver cancer

#### 3. Treatment response prediction



#### Key publications:

- Wang at al, Cancer Research, 2020
- Zhang et al, Am J Pathol. 2022
- Rong et al, Am J Pathology., 2023
- Wang et al, J. Clin. Invest., 2023
- Rong et al, Modern Pathology, 2023
- Wang et al, Nature Communications, 2023

#### pMTnet : predicts the binding of T cell receptors (TCR) with TCR targets using AI approaches

#### In silico

- pMTnet and associated technologies for TCR design (IP protected)\*
- Proprietary database of TCR-antigen binding relationships

#### \*PCT nationalized/filed in the US in March 2023

ARTICLES https://doi.org/10.1038/s42256-021-00383-2 mature machine intelligence

# Deep learning-based prediction of the T cell receptor-antigen binding specificity

Tianshi Lu<sup>1,6</sup>, Ze Zhang<sup>1,6</sup>, James Zhu<sup>1</sup>, Yunguan Wang<sup>1</sup>, Peixin Jiang<sup>2</sup>, Xue Xiao<sup>1</sup>, Chantale Bernatchez<sup>3</sup>, John V. Heymach<sup>2</sup>, Don L. Gibbons<sup>©</sup><sup>2</sup>, Jun Wang<sup>4</sup>, Lin Xu<sup>1</sup>, Alexandre Reuben<sup>©</sup><sup>2</sup> ⊠ and Tao Wang<sup>4</sup>, Dis ⊠

**Empirical validation** 

- Extensive library of TCR targets and corresponding TCRs available to enhance pMTnet
- Functional validation of enhanced TCRs
- Established workflow to isolate ultra-rare TCRs

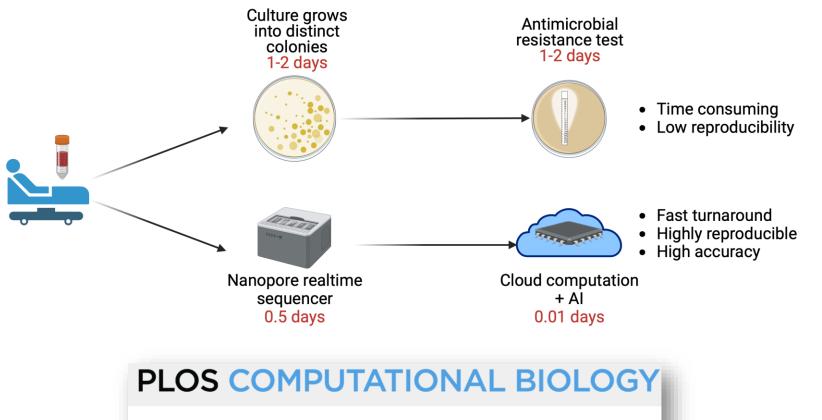
Real-time feedback to improve in silico assets

TCRs predicted and validated iteratively in-house

Licensable TCRs

### Al-assisted Antimicrobial Resistance Tests

#### PARP (Pan-Antibiotic Resistance Prediction) improves antimicrobial resistance tests



🔓 OPEN ACCESS 尨 PEER-REVIEWED

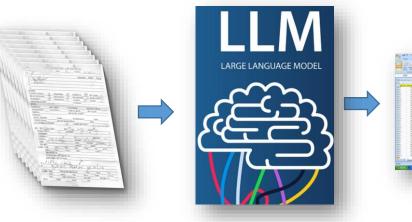
RESEARCH ARTICLE

### VAMPr: <u>VA</u>riant <u>Mapping</u> and <u>Prediction</u> of antibiotic <u>resistance</u> via explainable features and machine learning

Jiwoong Kim 😹, David E. Greenberg 🕷 🖾, Reed Pifer, Shuang Jiang, Guanghua Xiao, Samuel A. Shelburne, Andrew Koh, Yang Xie, Xiaowei Zhan 🖾

# A critical assessment of using ChatGPT for extracting structured data from clinical notes

Prompt engineering design



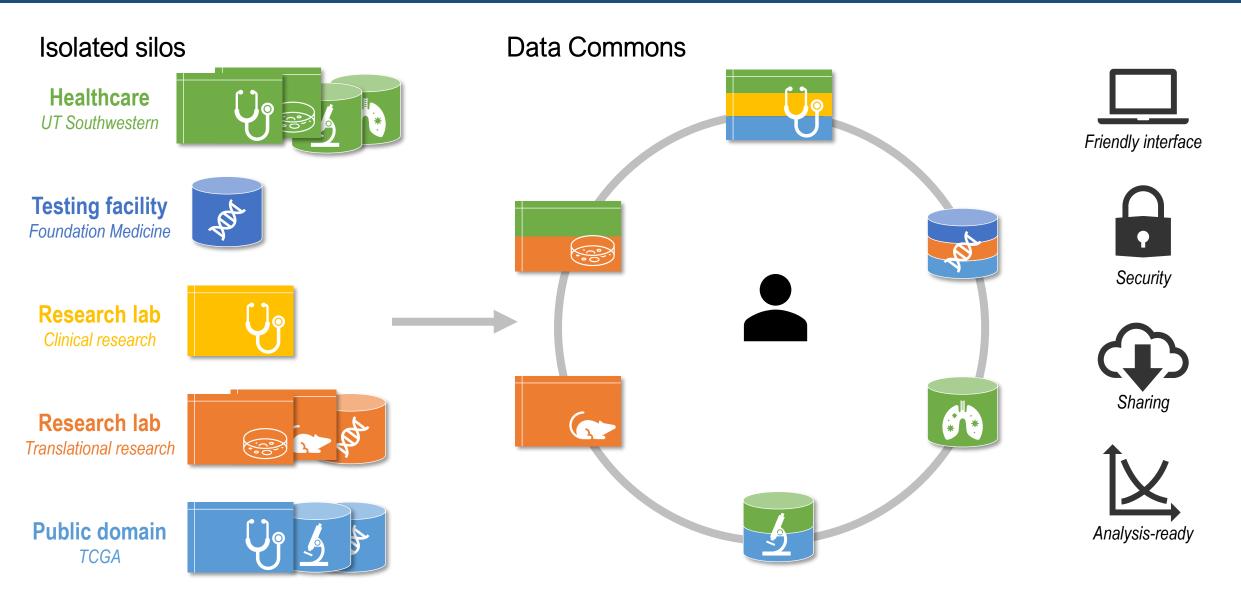
2	1000											
1	-	Aug-bank.	-		1.21							
		0.4			1.16	2.02					8.3	
	-				-	and here is	No. oracle for					
				Course of the local division of the local di					Station 1		-	
R.		A.										
Ļ		A 44 ANA	10.00	No. Distance in the		A 100 A	ALC: NAME OF TAXABLE	1.1	State of Concession, Name	10.1	10.0	
			2									
		ALC: NO.		14	1141	214	10.0	144	- 10 K A		- 22	
		1.0.00				1014	0.07	104.4	1011	4414	100	
		100.000			222	10.75	10.7	10.07	1000	111	14	
	10.00	40.04.00m				-	847.7	00.77		1017		_
	21.00	10.00	- 10	12		max.e	0114		212	414	125	
			100									_
	1.0.00	20.00.007		10.0	141	1027 A	14.7	100.0	100.0	20.4 8	164	
		14.45.00m		12	847	8.8 8.8 X		10.0	10.0 ×	100.0	14 14	
					222		212					
		10.00		12	2020	107.	101	0.1		07.8	12	
			- 21-			100.0	10.0	100		10.1		_
		10.00					10101	10.13		101.0		
		No. or all street				222	104.0	10.0	40.0	10.17	22	_
			1.0									
		10.00 mm		11	111	100	2012	141		14.8	12	_
	100	A	- 2-		41.4	10.00	14.4	14	0.0	10.0	4.0	
			- 61									
	- 1 m							100				and the second
											a Denned Looks	

**Overall performance of ChatGPT on data extraction from pathology reports** 

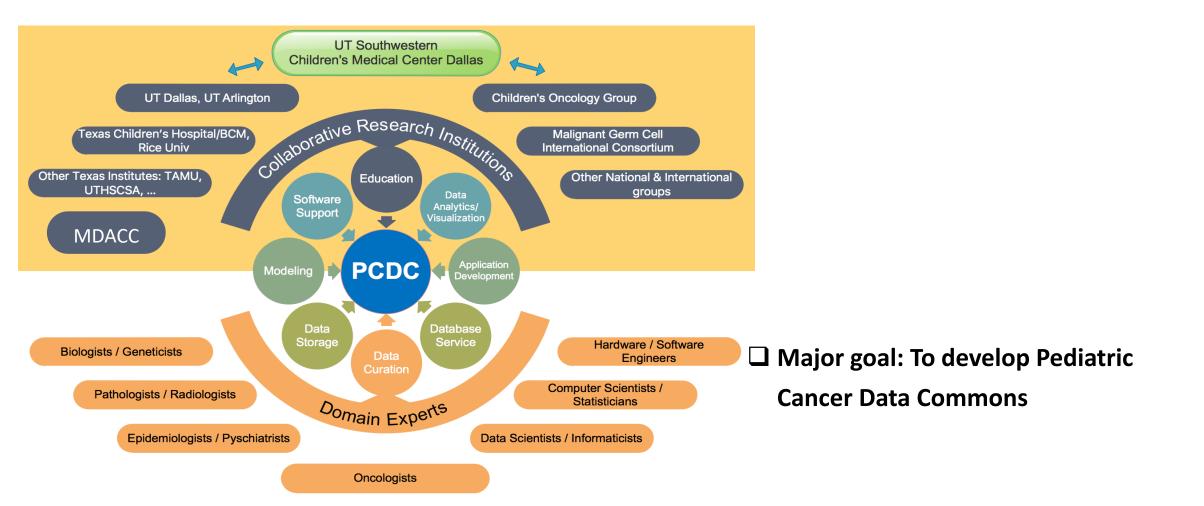
Attribute	Accuracy	F1	Kappa	Recall	Precision	Coverage
primary tumor features (pT)	0.87	0.87	0.76	0.87	0.89	0.97
regional lymph node involvement (pN)	0.91	0.91	0.84	0.91	0.92	0.94
overall tumor stage	0.76	0.76	0.61	0.76	0.77	0.94
histological diagnosis	0.99	0.99	0.98	0.99	0.99	0.96
Average	0.89	0.88	0.80	0.89	0.89	0.95

Huang et al, npj Digital Medicine, 2024

# Data Integration Platform



# PCDC: Pediatric Cancer Data Core (CPRIT RP180805)



UT Southwestern Medical Center

### Funding

NCI, NIGMS, NIAID, CPRIT, UTSW Endowment

#### **Quantitative Biomedical Research Center (QBRC)**



https://qbrc.swmed.edu/