

Generative AI (GenAI) for Toxicology and Drug Safety



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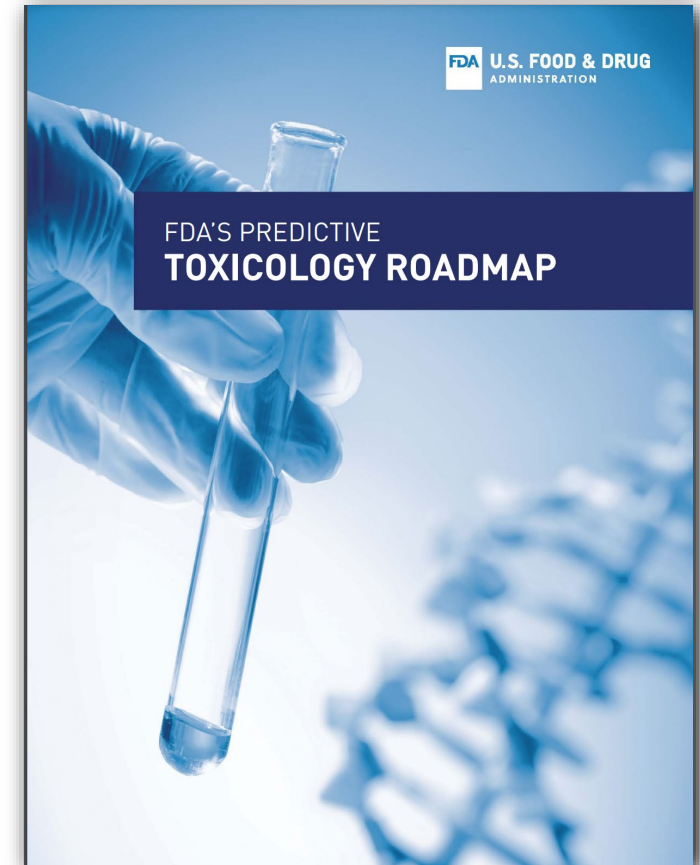
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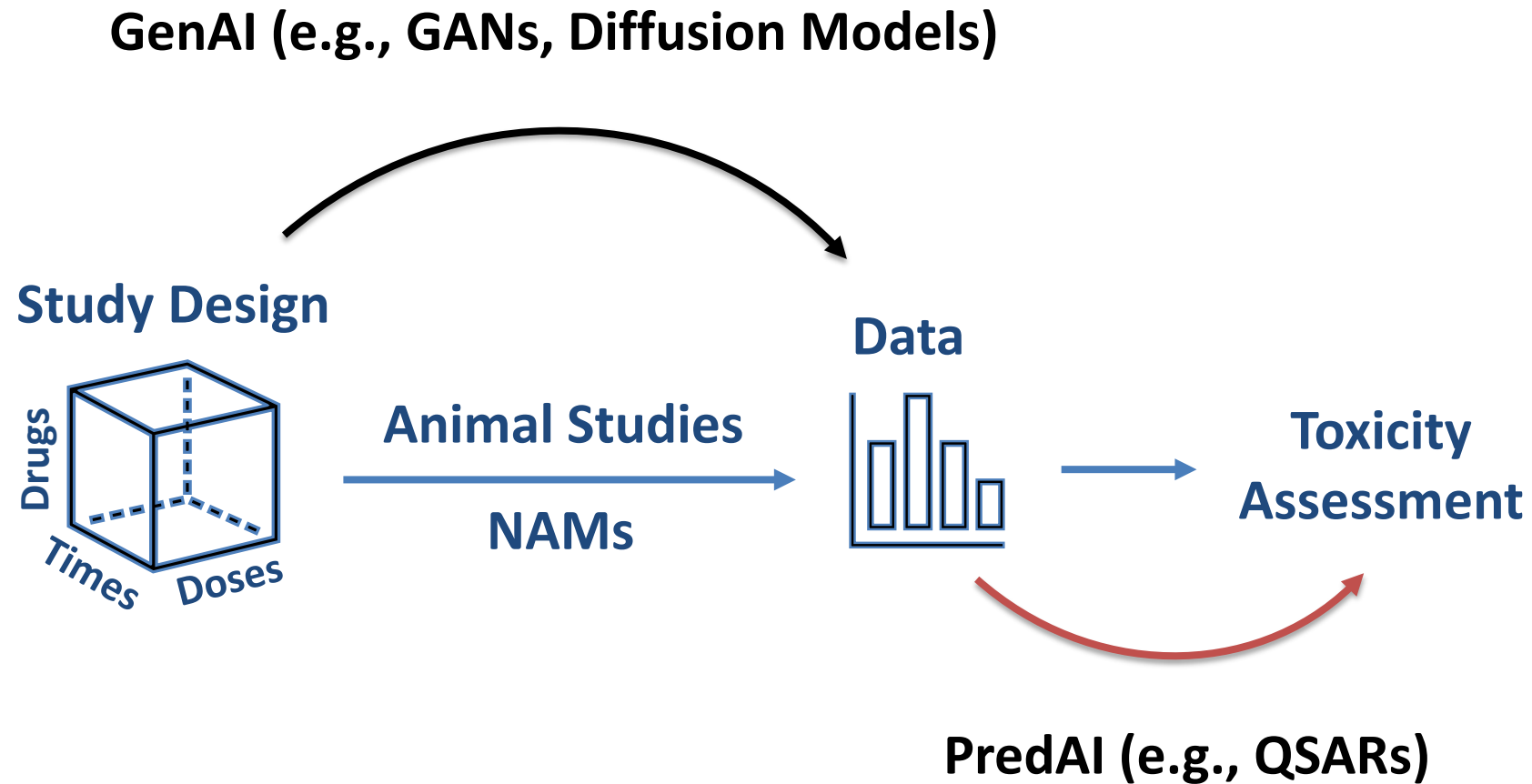
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The Paradigm Shift in Toxicology

- Global community has been promoting the 3Rs principle of reducing, refining and replacing animal uses by evaluating diverse NAMs in toxicological research and even for regulatory consideration, e.g., MPS, *in vitro*, *in silico* (AI/ML).
- The FDA Modernization Act 2.0 has been signed into the Law by the President; it urges the adoption of alternative methods to advance 3Rs science of animal use.
- FDA Predictive Toxicology Roadmap promotes the role of NAMs and alternative methods to support regulatory science and application.
- FDA ISTD program (Innovative Science and Technology Approaches for New Drugs) qualifies the new tools to support drug discovery and development.



Generative AI vs Predictive AI



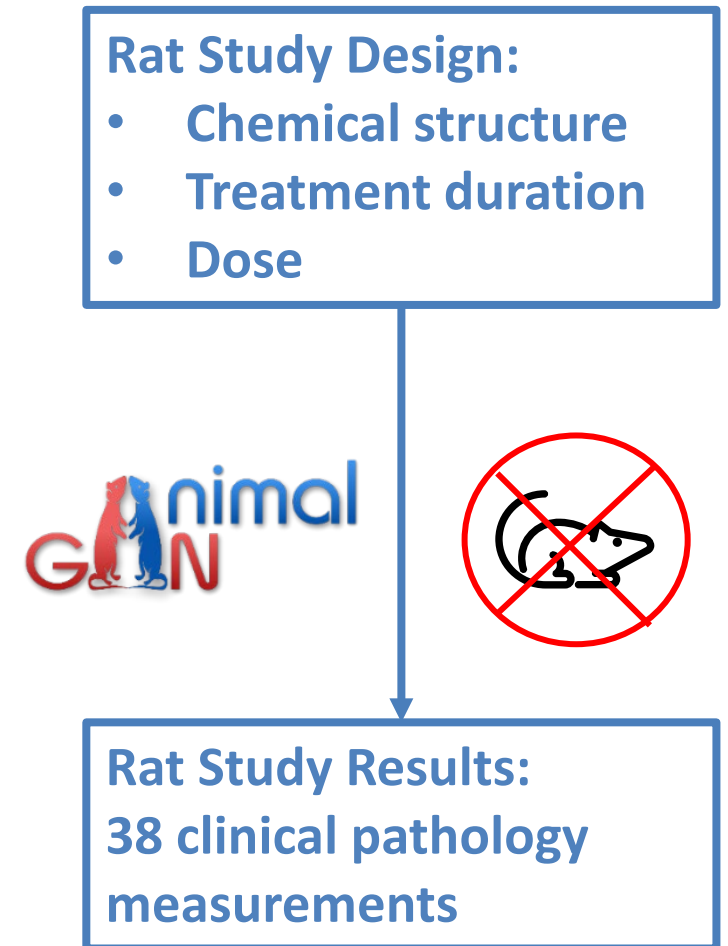


Virtual Animal models to Generate Animal Study Results with Generative AI

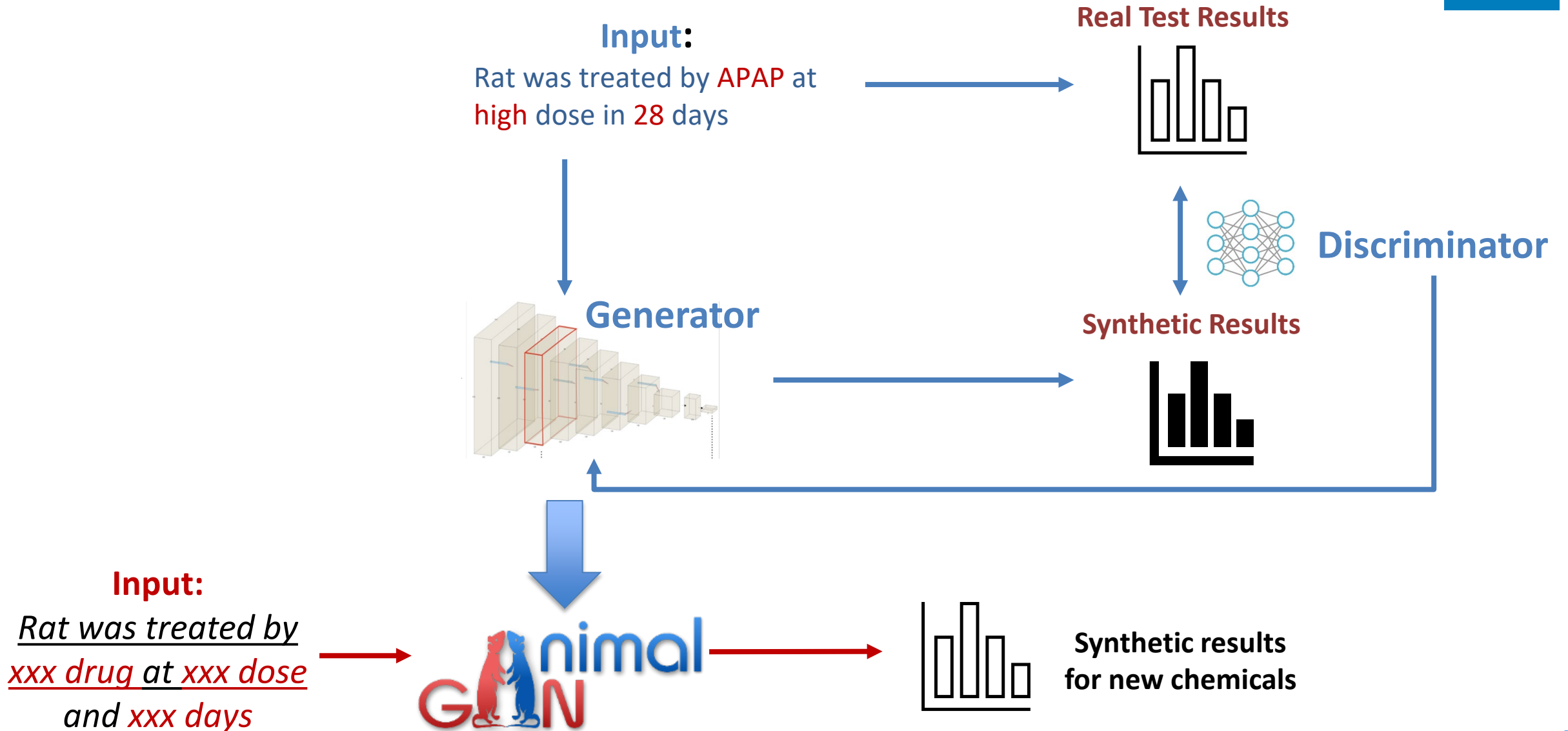


- **Why?** Animal studies assess safety of consumer products, but they are expensive and can pose ethical concerns. Can AI learn from past animal study data to generate animal study results of new untested compounds without using animals?
- **How?** **AnimalGAN** was developed using a Generative Adversarial Networks (GANs) framework (a DeepFake algorithm) to learn from the legacy animal data to produce new animal data *without using animals*.
- **Impact:** AnimalGAN can aid in assessing animal toxicity, potentially reducing or replacing animal testing in specific contexts.

Chen X., Liu Z., and Tong W. **A Generative Adversarial Network Model Alternative to Animal Studies for Clinical Pathology Assessment.**
Nature Communications. 2023, 14, 7141.



A GAN Framework for Animal Study

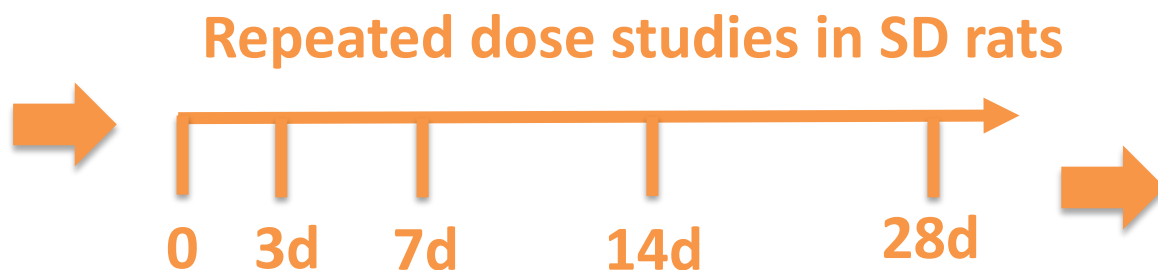


Proof-Of-Concept Studies with TG-GATEs Data

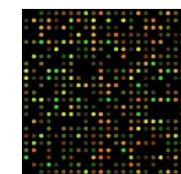


Study Design

- 138 compounds
- 3 doses (L/M/H)
- 5 rats per group



38 clinical
pathology
measures

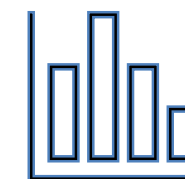


Gene
expression
profiles

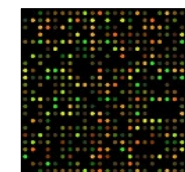


A new compound

- Chemical descriptors
- Dose (L, M, or H)
- Time (3d, 7d, 14d, or 28d)



38 clinical
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measures



Gene
expression
profiles

17 Hematologic measures

Short Name	Full name
RBC(x10 ₄ /ul)	Erythrocytes
Hb(g/dL)	Hemaglobin
Ht(%)	Hematocrit (%)
MCV(fL)	Mean corpuscular volume
MCH(pg)	Mean corpuscular hemaglobin
MCHC(%)	Mean corpuscular hemaglobin concentration
Ret(%)	Reticulocytes
Plat(x10 ₄ /uL)	Platelets
WBC(x10 ₂ /uL)	Leukocytes
Neu(%)	Neutrophils
Eos(%)	Eosinophils
Bas(%)	Basophils
Mono(%)	Monocytes
Lym(%)	Lymphocytes
PT(s)	Prothrombin Time
APTT(s)	Activated Partial Thromboplastin Time
Fbg(mg/dL)	Fibrinogen

21 Clinical Chemistry

Hepatotoxicity

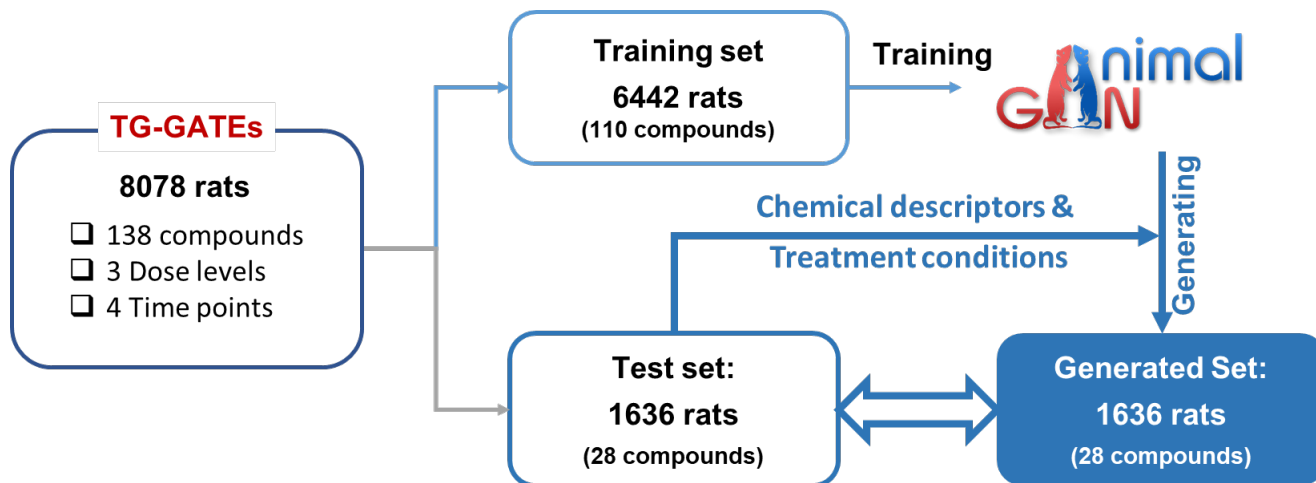
Nephrotoxicity

Short Name	Full name
ALP(IU/L)	Alkaline phosphatase
TBIL(mg/dL)	total bilirubin
DBIL(mg/dL)	direct bilirubin
AST(IU/L)	aspartate aminotransferase
ALT(IU/L)	Alanine aminotransferase
LDH(IU/L)	Lactate Dehydrogenase
GTP(IU/L)	Gamma-glutamyltranspeptidase
BUN(mg/dL)	Blood Urea Nitrogen
CRE(mg/dL)	Creatinine (mg/dL)
Na(meq/L)	sodium
K(meq/L)	potassium
Cl(meq/L)	chlorine
Ca(mg/dL)	calcium
IP(mg/dL)	inorganic phosphorus
TC(mg/dL)	Cholesterol
TG(mg/dL)	Triglycerides
PL(mg/dL)	Phospholipid
GLC(mg/dL)	glucose
TP(g/dL)	Total protein
RALB(g/dL)	Albumin
A/G	Albumin/globulin ratio

AnimalGAN Study Design and Performance



Predict 38 measurements of clinical pathology (e.g., ALT, ALP)



98% concordance in absolute value by comparing to the experimental results for 28 different compounds

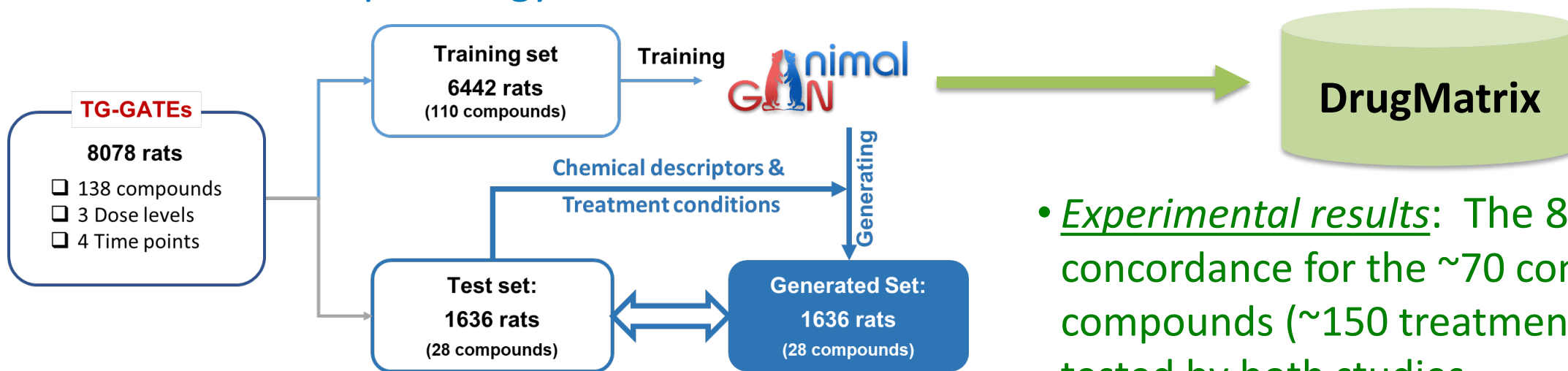
“Stress” Test of AnimalGAN:

- 1. Applicability Domain** – The 28 drugs were structurally unsimilar to the 110 drugs used for training.
- 2. Cross-Drug Classes Application** – The 28 drugs belong to the therapeutic classes that were not represented in the training set.
- 3. Cross-Year Application** – The 28 drugs were approved in the late years compared to the training set drugs.

AnimalGAN: Validation on DrugMatrix



Predict 25 clinical pathology measures



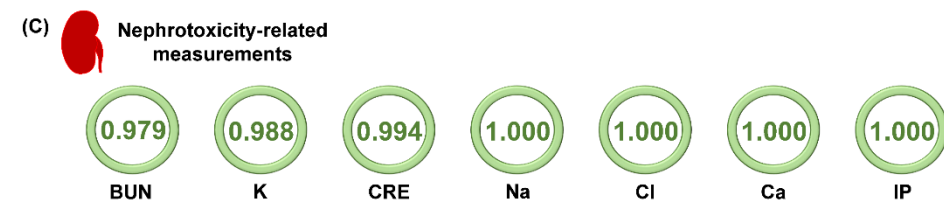
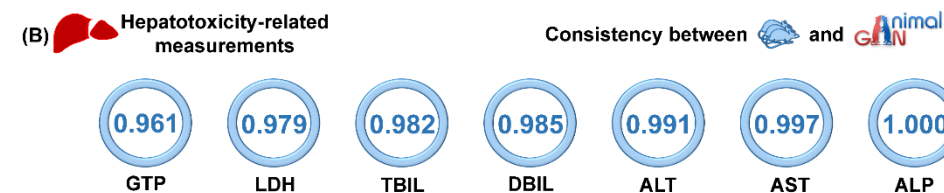
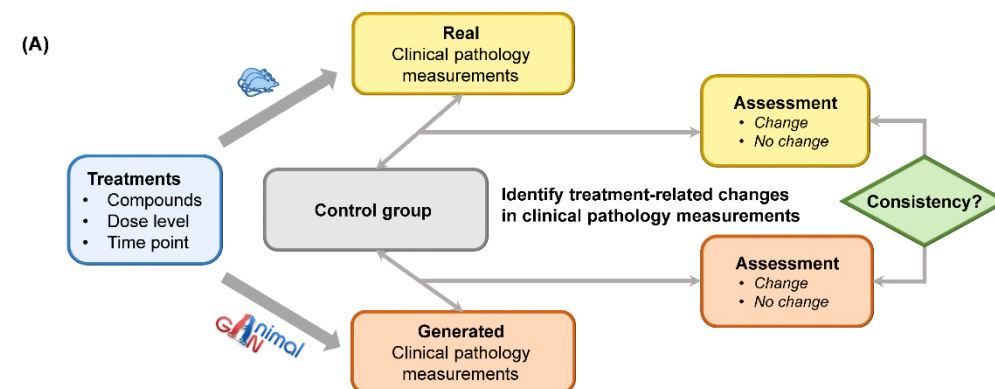
- Experimental results: The 82% concordance for the ~70 common compounds (~150 treatments) tested by both studies.

- AnimalGAN predicted results: the 83% concordance for 355 compounds (>700 treatments) that were tested in DrugMatrix but not by TG-GATEs.

	TG-GATEs	DrugMatrix
Rat strain, sex and age	SD, male, and 6 weeks	SD, male, and 6-8 weeks
Treatment duration	<u>3</u> /7/14/28 days	0.25/1/ <u>3</u> /4/5/7/14 days
Dose	<ul style="list-style-type: none">• MTD is based on 7 days• 3 treatment doses	<ul style="list-style-type: none">• MTD is based on 5 days• Most are 2 doses

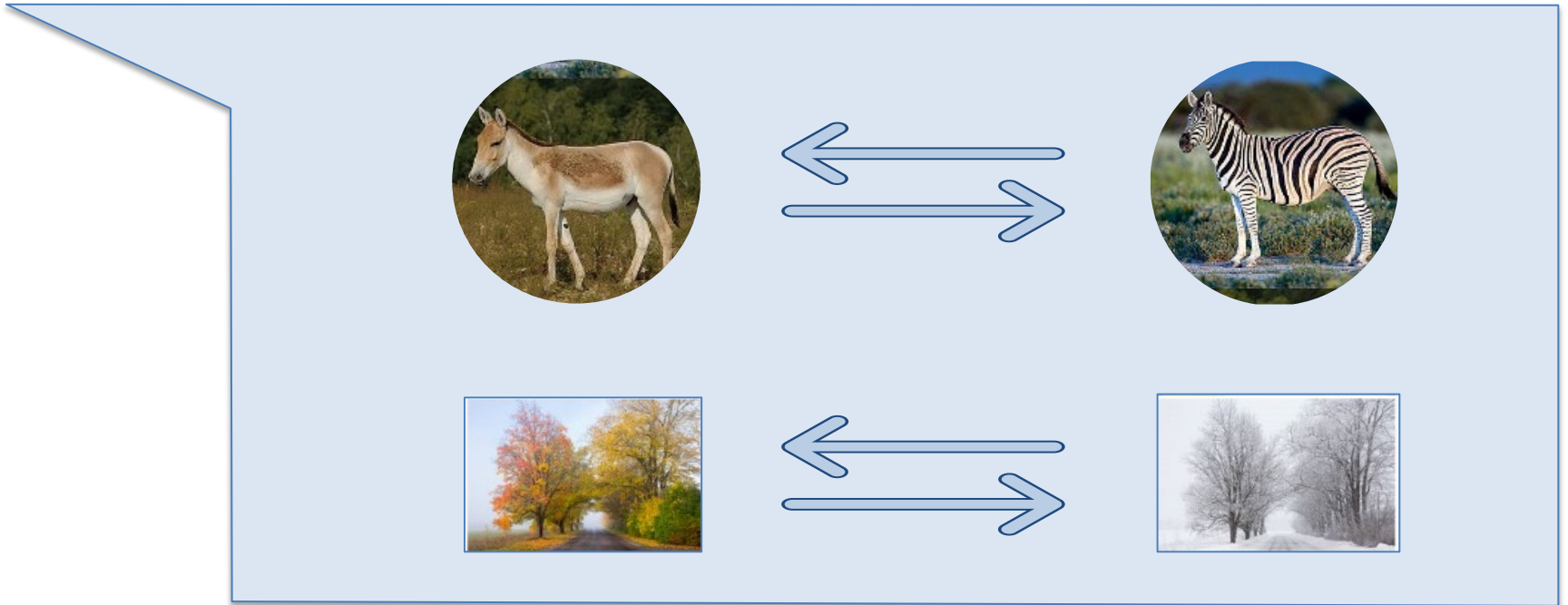
Consistency between real and synthetic results for hepatotoxicity assessment:

- Assessed hepatotoxicity based on FDA Guidance
 - ALT > 3 of normal condition indicates **intermediate hepatotoxicity**: **100%** in agreement
 - ALT = 1-2 of normal condition indicates **minor hepatotoxicity**: **83%** in agreement
- Assessed hepatotoxicity patterns
 - ALT/ALP > 5 indicates **hepatocellular** injury: **100%** in agreement
 - ALT/ALP < 2 indicates **cholestatic** injury: **99%** in agreement
 - ALT/ALP = 2-5 indicates **mixed** Injury: **81%** in agreement



GAN Variations

- C-GAN
- CycleGAN
- BiGAN
- SemiGAN
- DC-GAN
- StackGAN
- StyleGAN
- DualGAN
- ...





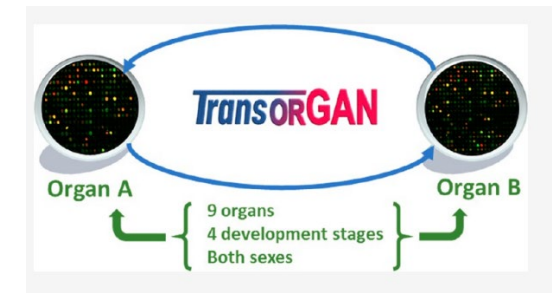
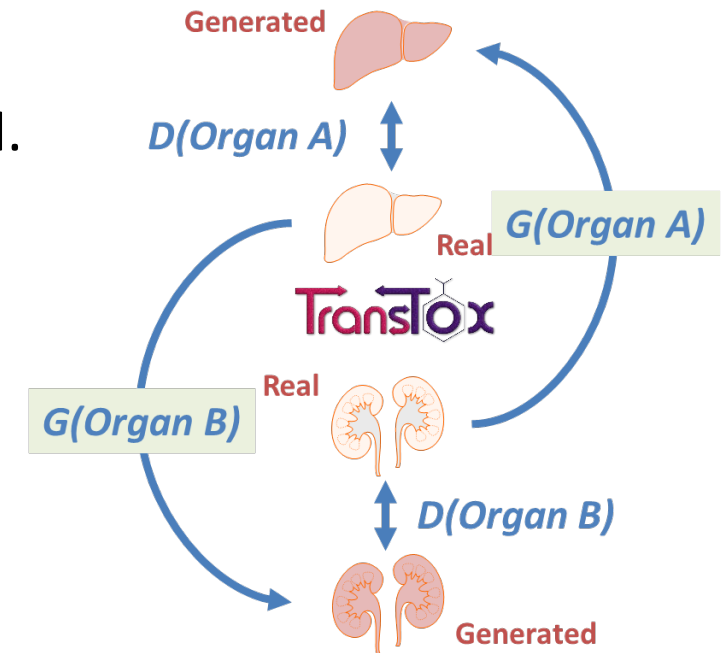
Translational Research with Generative AI



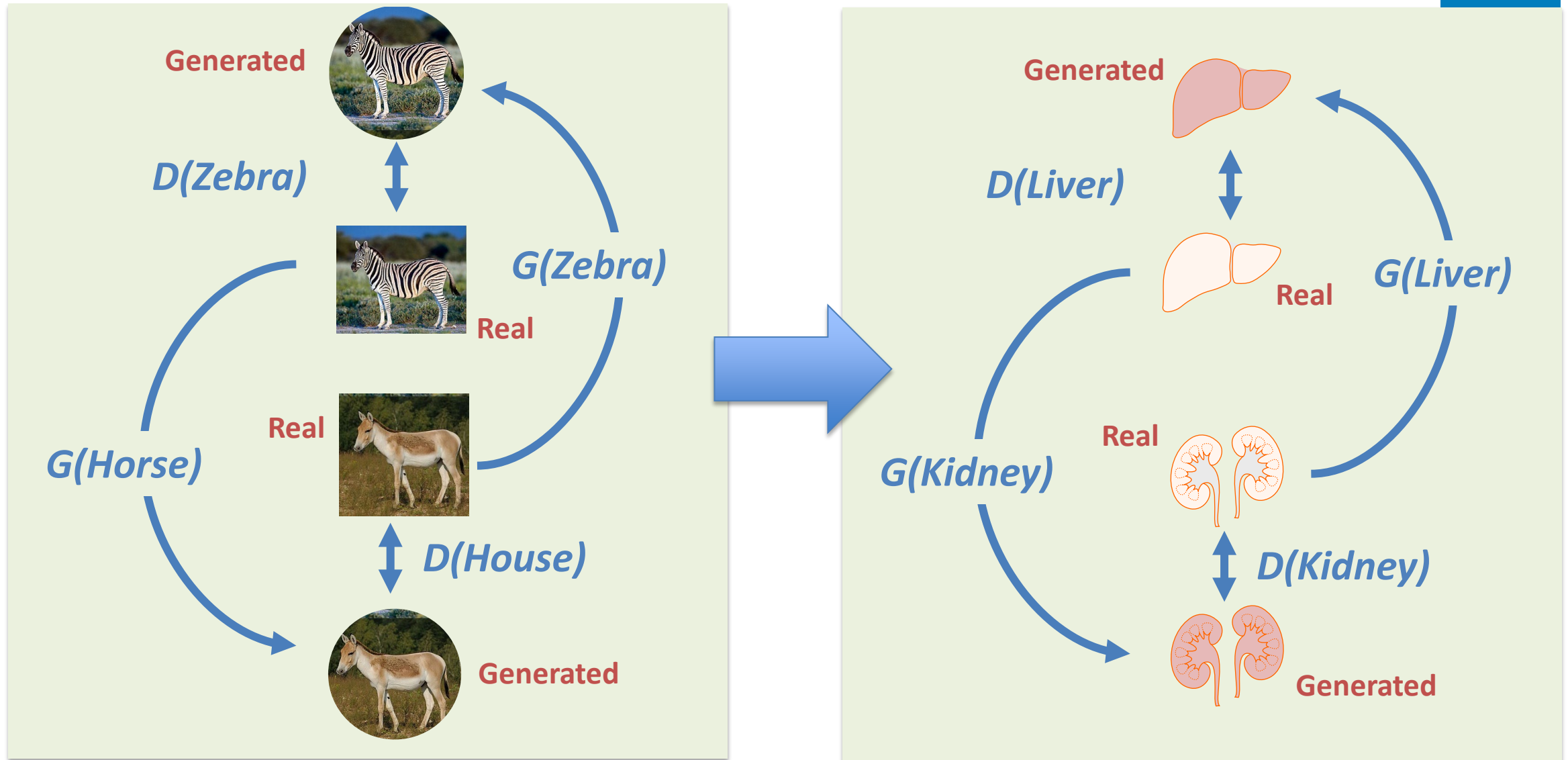
Objective: To translate experimental findings across domains where the data are scarce or difficult to obtain with generative AI.

Projects:

1. Translating rat genomic profiles across 10 organs at 4 development stages and with both sexes (published)
2. Bridging transcriptomics data between liver and kidney under the drug treatment (submitted)
3. Converting microarray data to RNA-seq to leverage the past investment (in preparation)
4. In vitro to in vivo extrapolation (IVIVE) to advance 3Rs (on-going)



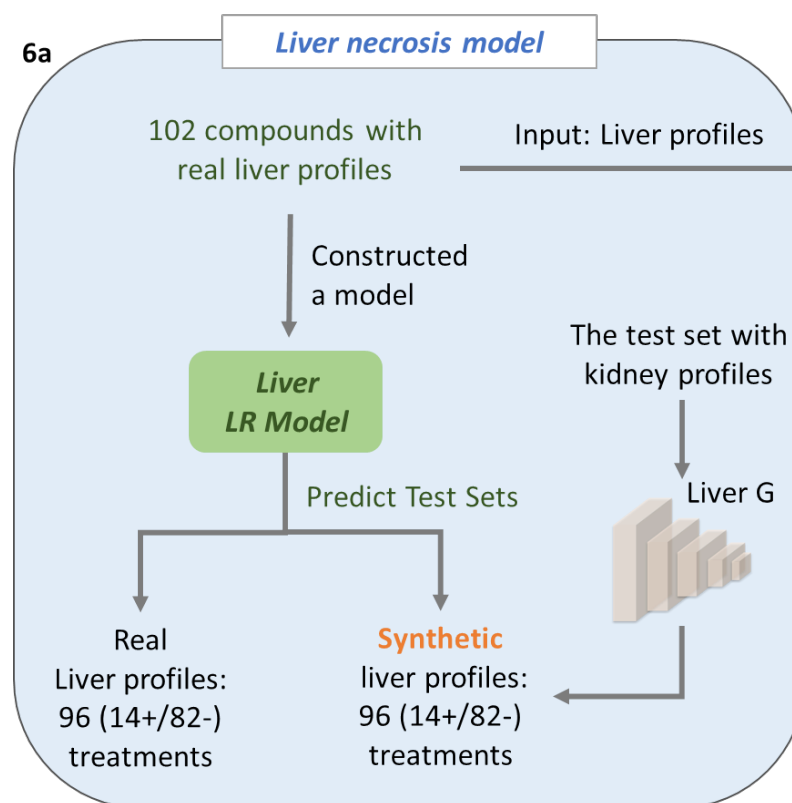
From CycleGAN to TransTox



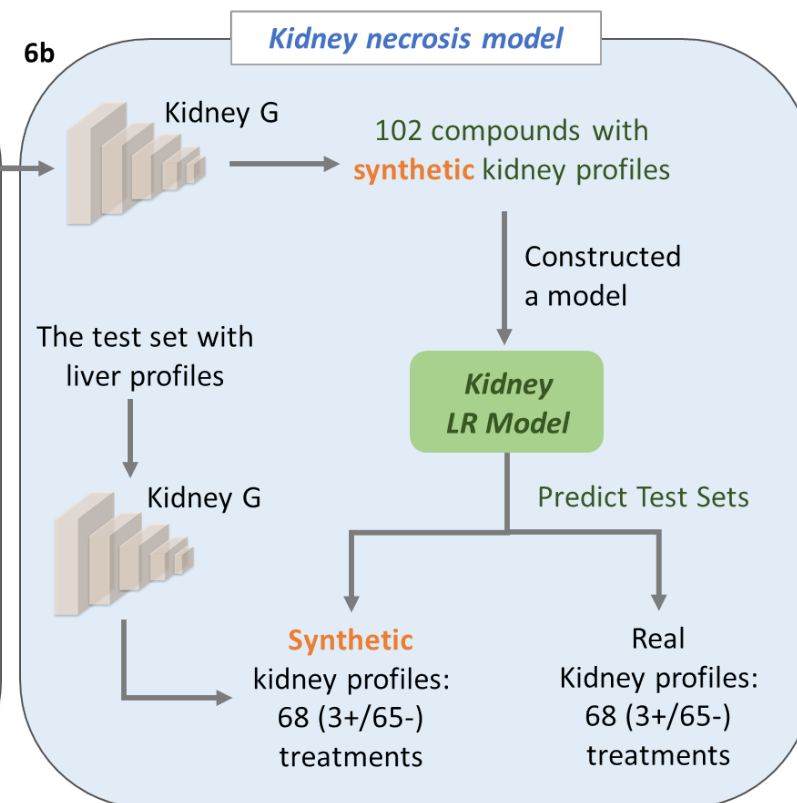
- Validated with data from the same lab as well as from a different lab
- Compared the synthetic profiles against data from a real experimental setting in elucidating toxicity mechanisms
- Predictive model necrosis prediction

Ting Li, Xi Chen, Weida Tong, **Bridging Organ Transcriptomics for Advancing Multiple Organ Toxicity Assessment with a Generative AI Approach**, NPJ Digital Medicine (in revision)

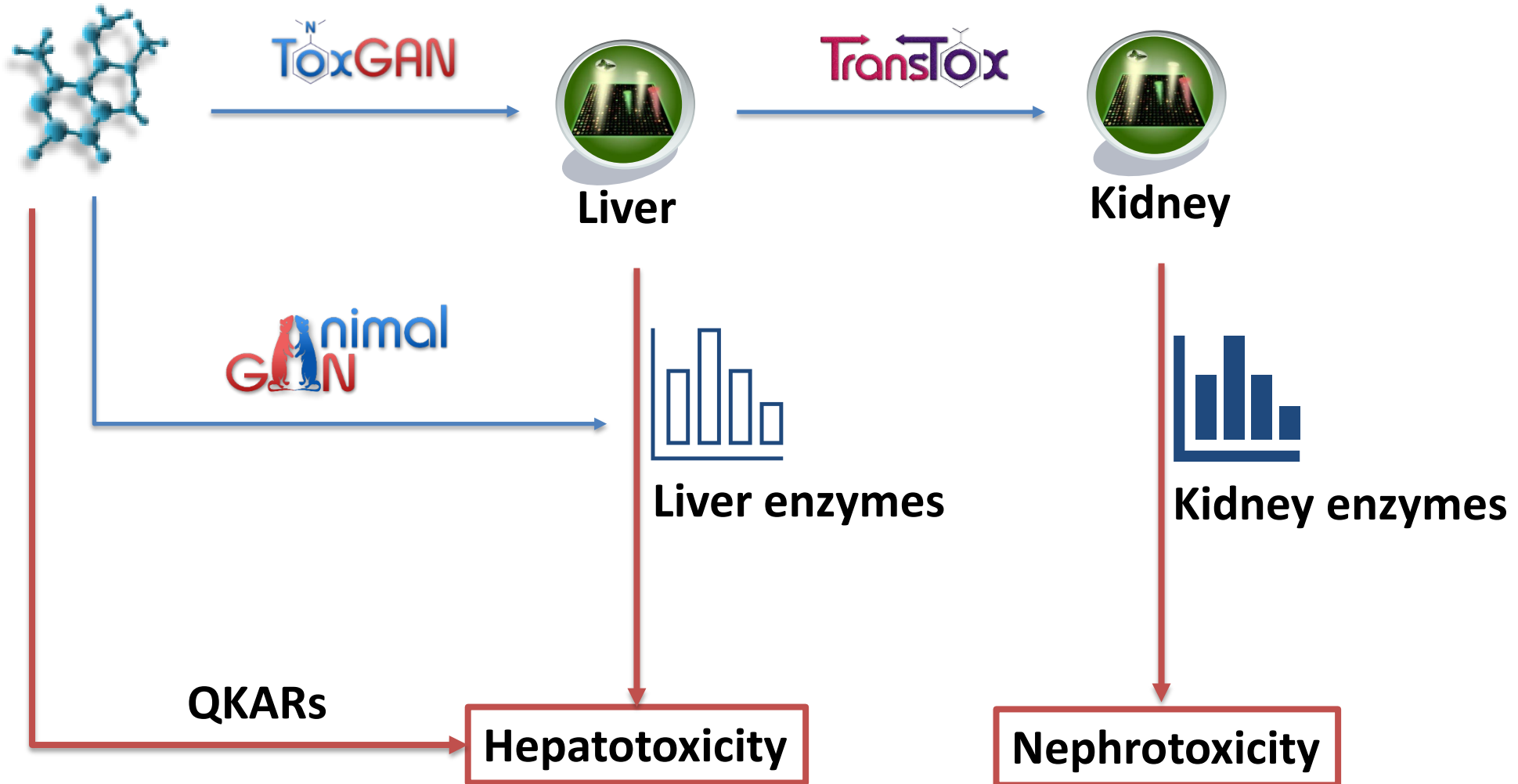
Can the synthetic data serve as “digital twin” for diagnosis



Can the synthetic data be used to develop a reliable predictive model



GenAI Application in Drug Safety Assessment: A Scenario



FDA/NCTR AI Program for Toxicology (AI4TOX)



LLMs for FDA documents to improve regulatory efficiency, enhance information retrieval, and maintain institutional memory at FDA



Predictive models for safety endpoints critical to drug safety review, particularly in IND Application review at CDER



Virtual animals to generate animal study results with generative AI to advance 3Rs for animal use and digital twin



Generative AI models to translate experiment findings across different domains such as across organs, in vitro-to-in vivo (IVIVE), and between genomic technologies



AI-driven digital pathology for preclinical histopathology images

<https://www.fda.gov/about-fda/nctr-research-focus-areas/artificial-intelligence>



***YOUR ONLY
LIMITATION IS YOUR
IMAGINATION !***



Acknowledgment

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- Skylar Connor (Adaptive AI)
- Yanyan Qu (CardioTox and QSARs)
- Shivangi Shrimali (MPS and women's health)
- Alex Chen (AnimalGAN)

CDER Collaborators on AnimalGAN

- Shraddha Thakkar (OCS and SafetAI)
- Kevin Snyder (OND)
- Paul Brown (OND)

FDA Leadership

- Principle Deputy Commissioner and Chief Scientist
- CDER Leadership: Lilliam Rosario (OCS), ShaAvhree Buckman Garner (OTS), and Bob Ball (OSE)

External Collaborators:

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