

University of Connecticut
Economic Development Forum 2025

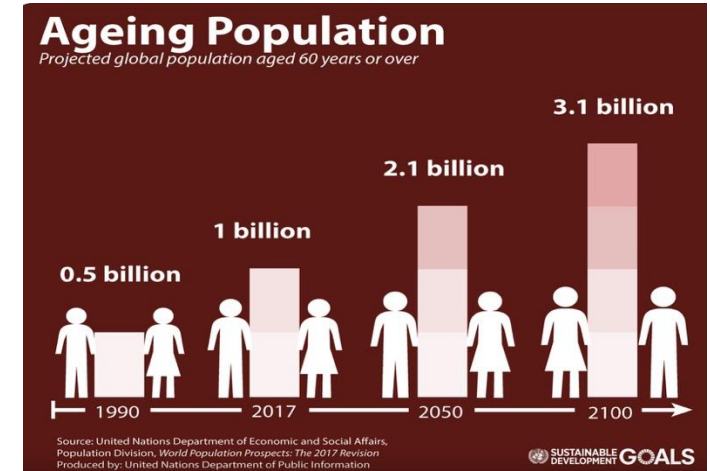


Agentic Generative AI for Precision Medicine

Martin Renqiang Min
NEC Laboratories America
Nov 13, 2025

Modern Healthcare Needs Transformation

A growing elderly population, driven by increases in the expected human lifespan, will pressure the existing medical infrastructure.



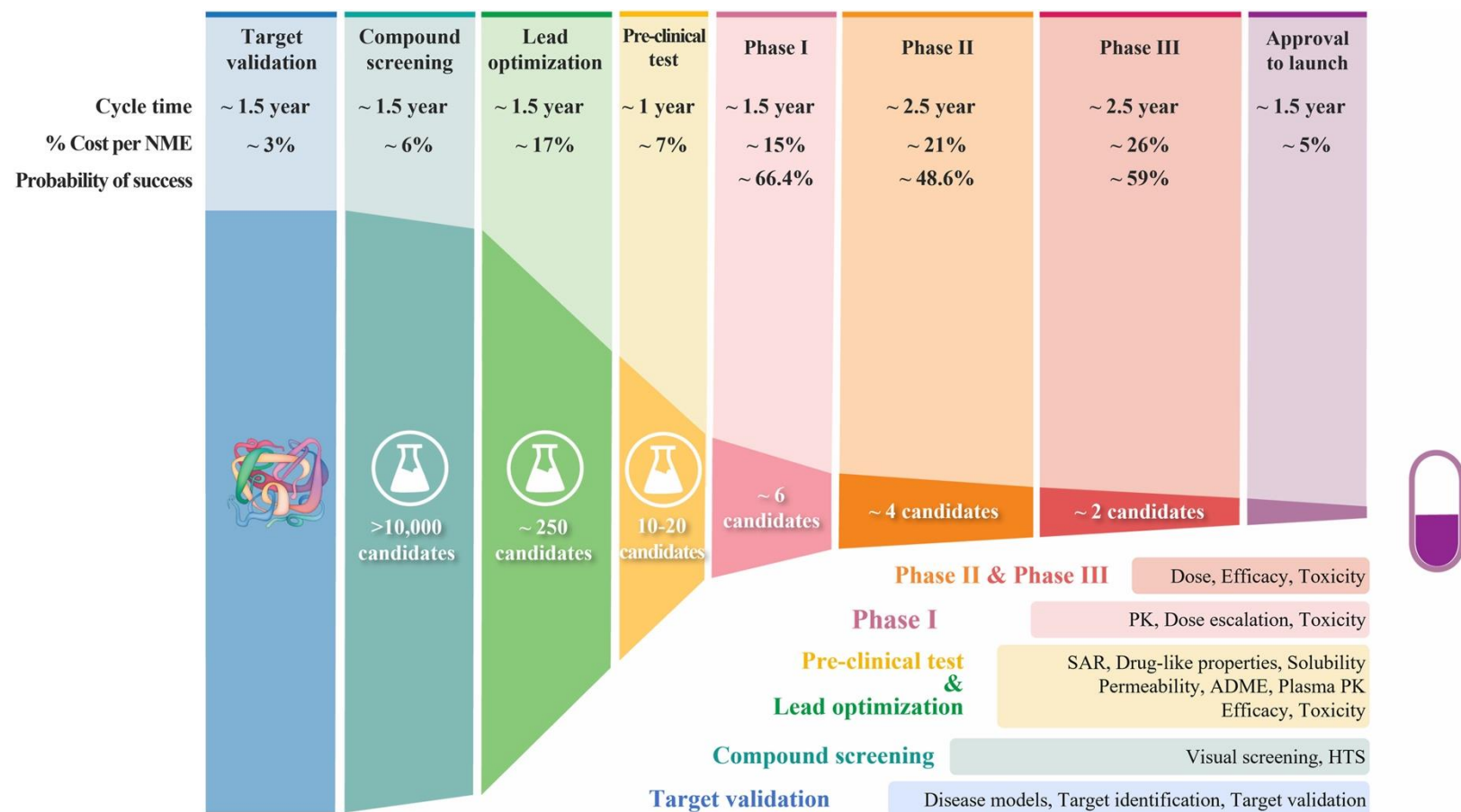
Widespread and low-cost world travel facilitates the quickly and easily spread of disease between countries, increasing the potential of frequent global pandemics.



Medical knowledge explosion will require an ever-increasing number of medical experts to manage and deliver life saving treatments.



Traditional Drug Design Business is Not Sustainable



Increasing pharmaceutical R&D cost, but still low success rate (<10%)

Average time/cost for designing one new drug = >10 years + \$~2B

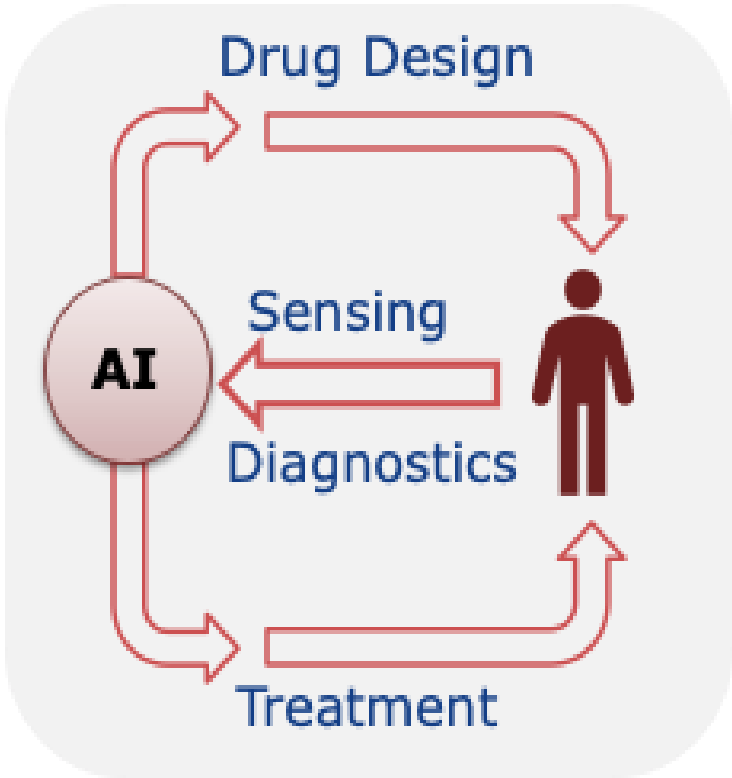
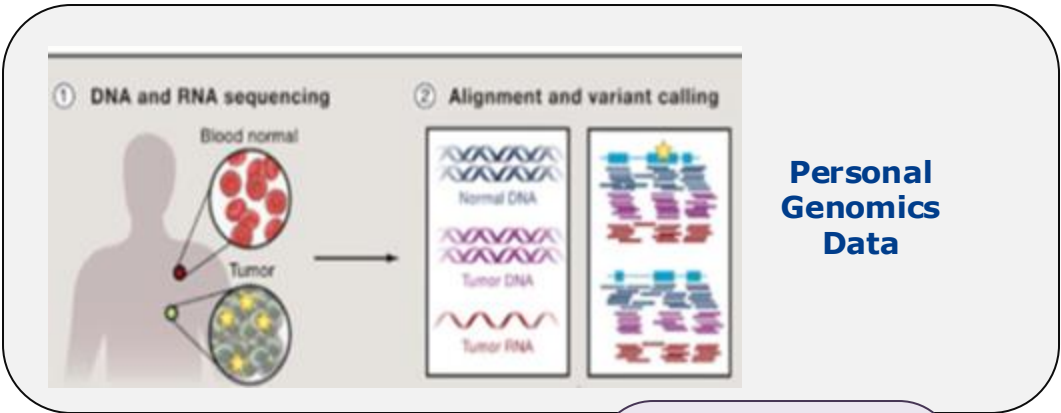
Large-Scale Biomedical Data Enables Precision Medicine and Healthcare



TCGA's Pan-Cancer Atlas



TCGA Outcomes & Impact



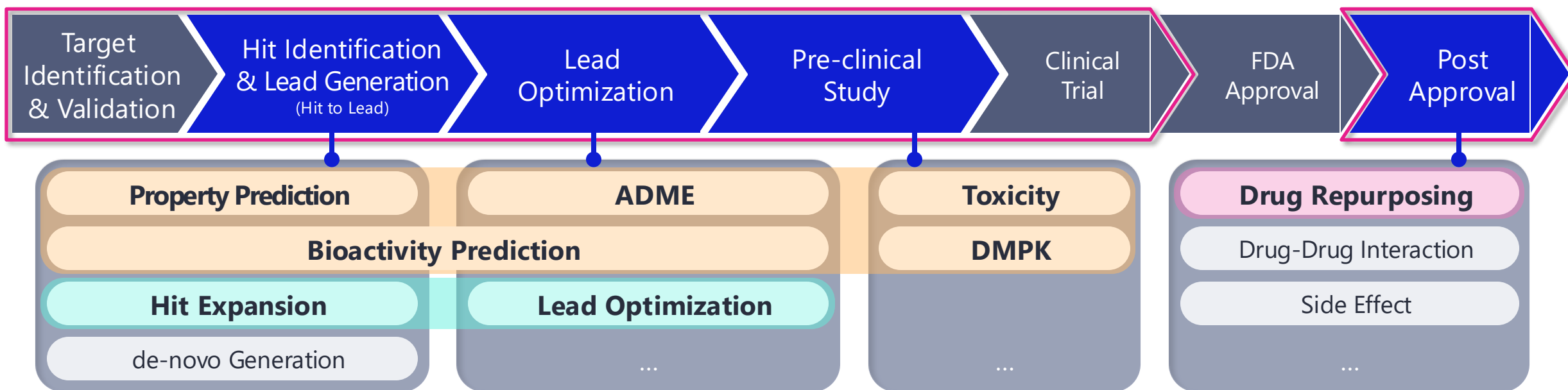
AI Accelerated Drug Development Process

FDA NEWS RELEASE

Advances in **AI and high-throughput experiments** empower **accelerated drug development**

FDA Announces Plan to Phase Out Animal Testing Requirement for Monoclonal Antibodies and Other Drugs

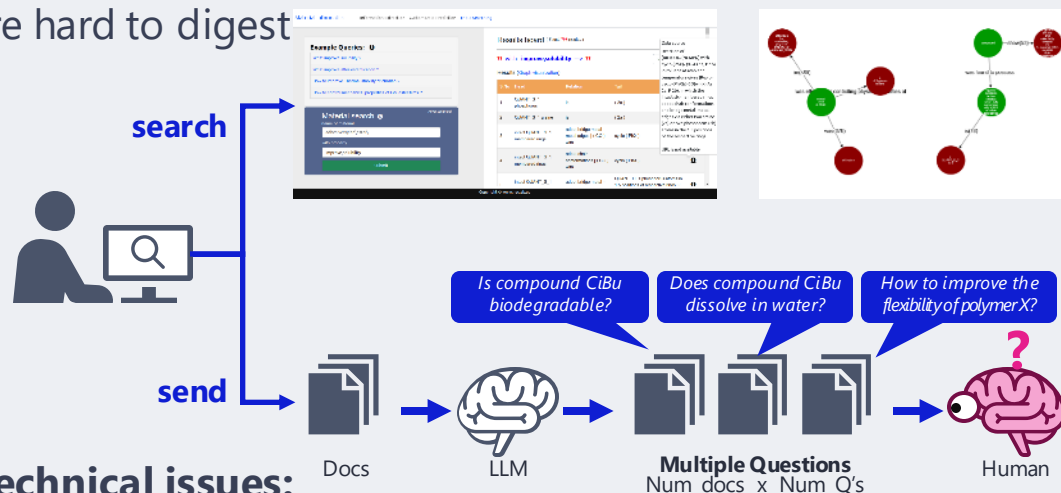
For Immediate Release: April 10, 2025



Human Collaborative Multi-Agents for Drug Screening

Enable interactive privacy-preserving insight discovery from multimodal data to significantly reduce drug investigation, screening, and design time

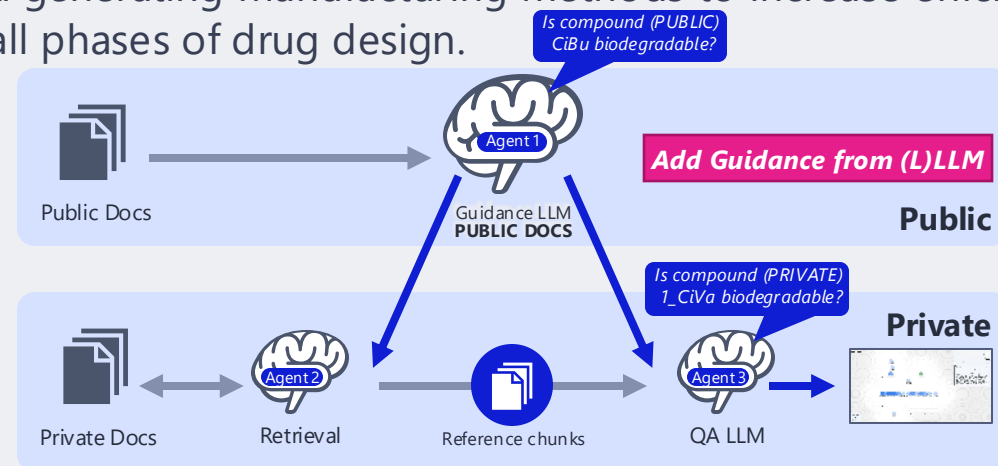
Traditional: Scientists manually search thousands of research papers to investigate/screen drug candidates or send a lot of private data to public LLMs to get lengthy textual answers that are hard to digest



Technical issues:


Reading many literature papers and reports is time-consuming; Sending private company data to public LLMs poses data-leaking and privacy issues.

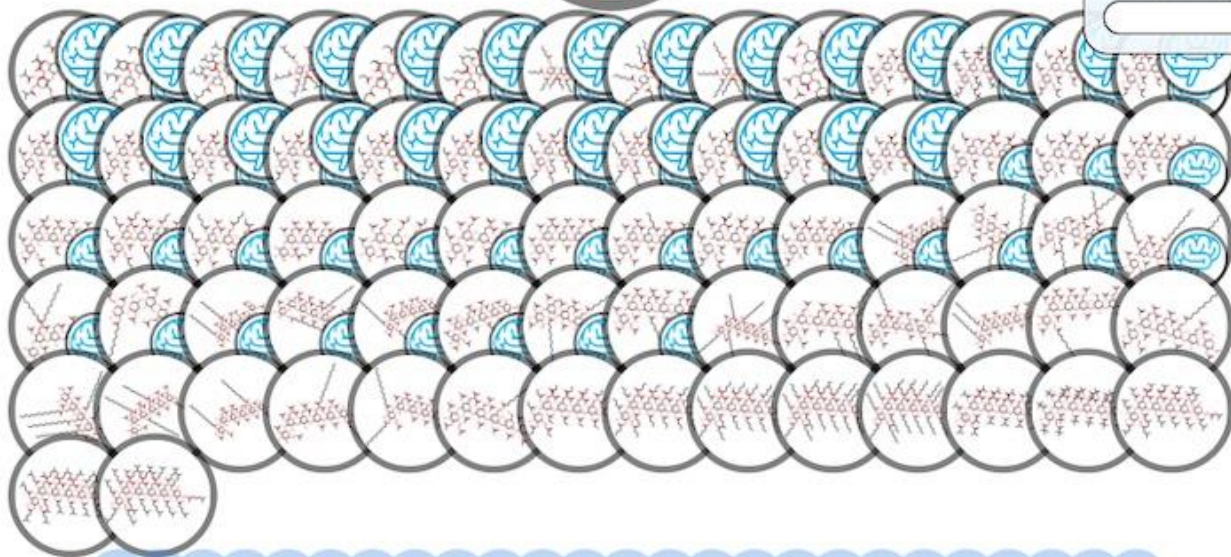
NEC: Human collaborative AI agents for interactively visualizing large-scale existing drugs, filtering candidates, answering questions, and generating manufacturing methods to increase efficiency in all phases of drug design.



Efficient and safe drug insight discovery:

LLM-enhanced interactive data visualization and QA increases the efficiency of insight discovery for drug design; Local agents enable data safety.

Begin Retrieval and QA 



Select Query:

Load

x

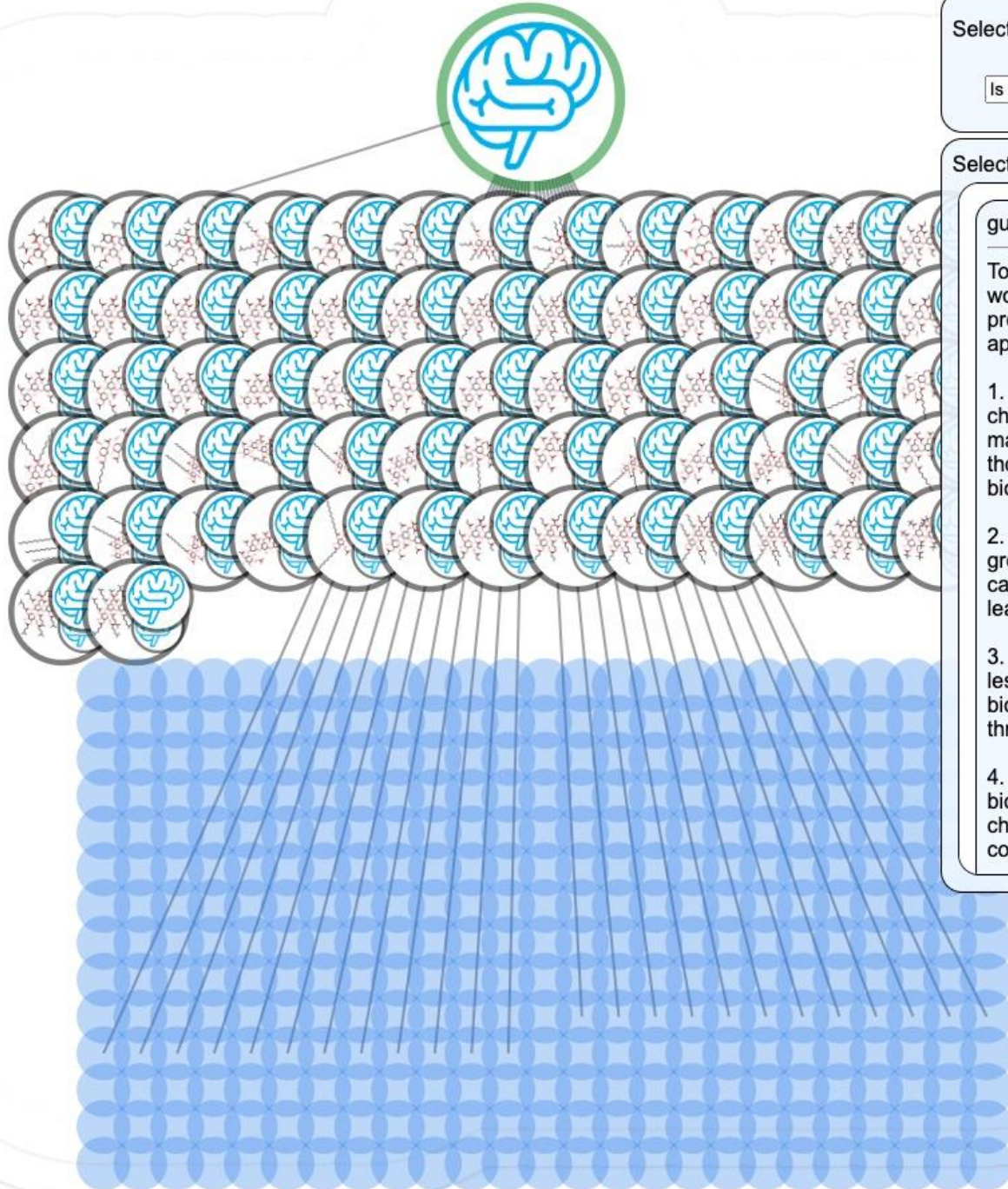
bool

AQA

Selected Objects ☒ Show Full Text

ZoomPanSelect

Lasso



Select Query:

Load

x

bool

AQA

Selected Objects ☒ Show Full Text

guidance_0

x

To determine if the provided polymer sample is biodegradable, I would analyze the information given and cross-reference it with the properties known to influence biodegradability. Here's how I would approach it:

1. **Chemical Structure**: I would examine the main chain and side chain composition to see if they consist of naturally occurring materials (like cellulose) or synthetic materials. Natural polymers or those derived from natural materials are more likely to be biodegradable.
2. **Functional Groups**: I would look into the type of functional groups present in the polymer structure. Ester groups, for example, can be subject to hydrolysis under the right conditions, which can lead to degradation.
3. **Degree of Substitution (DS)**: The higher the DS, typically, the less soluble the polymer might be, which might impact biodegradability. I would compare the DS value provided with known thresholds that might predict biodegradability.
4. **Molecular Weight**: Higher molecular weights can hinder biodegradation; however, there can be exceptions depending on the chemical structure. The molecular weight provided would be considered in the context of known data for similar polymers.

ZoomPanSelect

Lasso

Zoom Fit



Select Query: Is this polymer sample biodegradable?

Load

x

Is this polymer sample biodegradable?

bool

AQA

Selected Objects ☒ Show Full Text

answer_192_PaAcPr

x

[0.9] [link](#)

answer_204_PaAcPr

x

[0.5] [link](#)

answer_220_CdLa

x

[0.8]

The probability that the polymer sample is biodegradable is 0.8. [link](#)

answer_221_CdLa

x

[0.8]

The probability that the polymer sample is biodegradable is 0.8. [link](#)

ZoomPanSelect

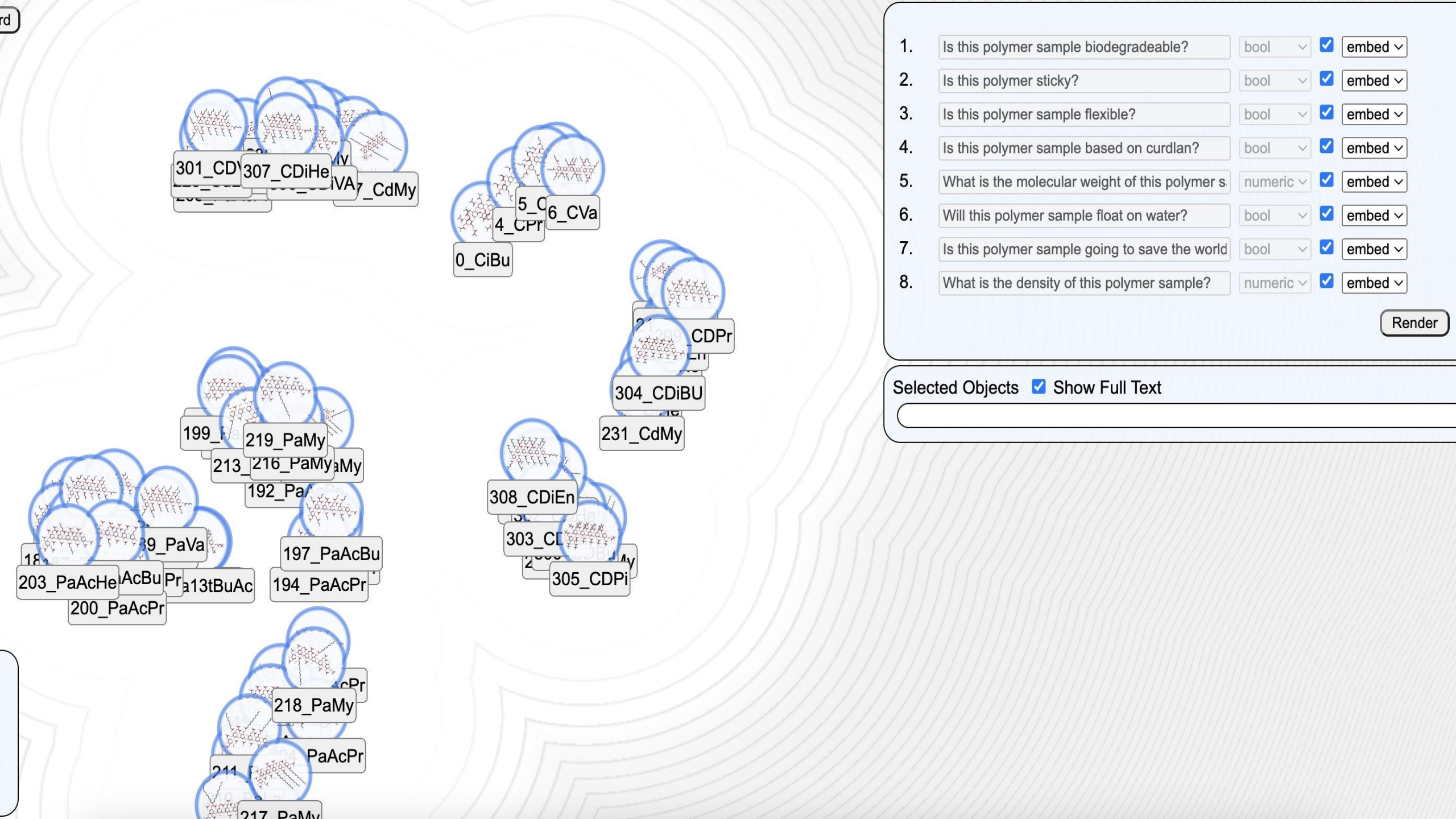
Lasso



- Is this polymer sample biodegradeable? ☒ embed
- Is this polymer sticky? ☒ embed
- Is this polymer sample flexible? ☒ embed
- Is this polymer sample based on curdlan? ☒ embed
- What is the molecular weight of this polymer s. ☒ embed
- Will this polymer sample float on water? ☒ embed
- Is this polymer sample going to save the world? ☒ embed
- What is the density of this polymer sample? ☒ embed

Render

Selected Objects ☒ Show Full Text



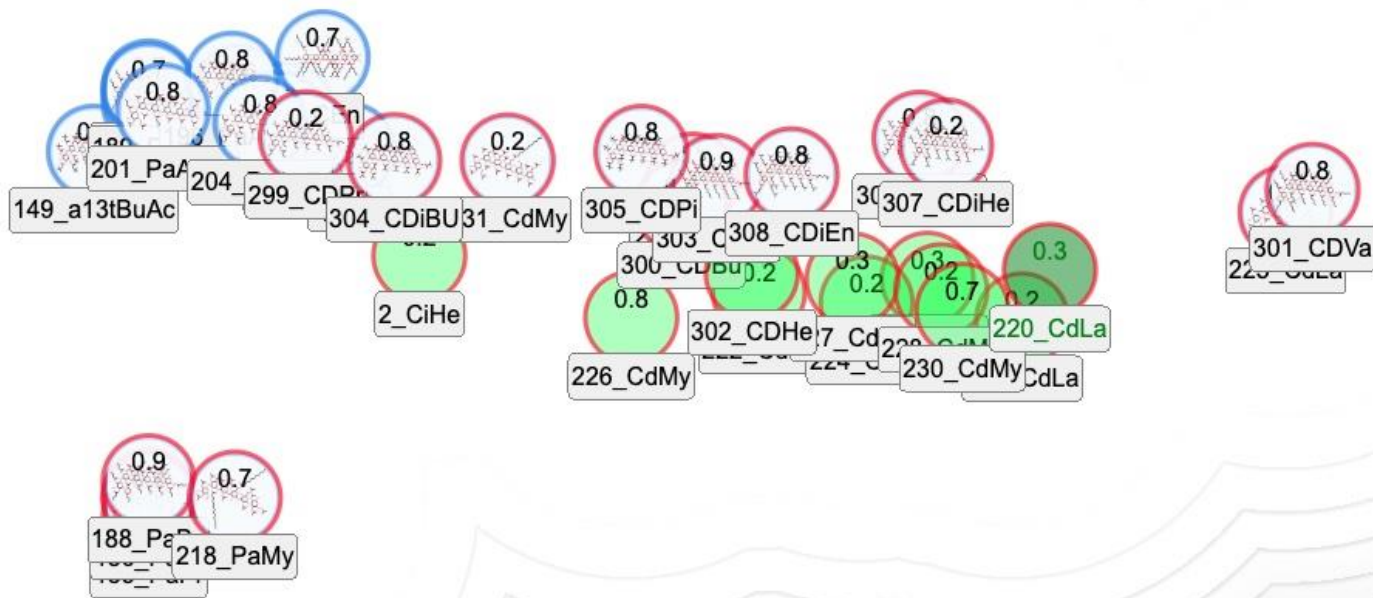
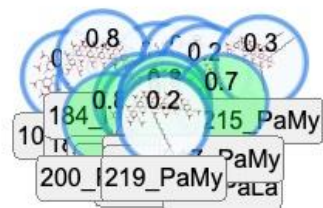
1. Is this polymer sample biodegradeable? bool ☒ embed
2. Is this polymer sticky? bool ☒ embed
3. Is this polymer sample flexible? bool ☒ embed
4. Is this polymer sample based on curdlan? bool ☒ embed
5. What is the molecular weight of this polymer s. numeric ☒ embed
6. Will this polymer sample float on water? bool ☒ embed
7. Is this polymer sample going to save the world? bool ☒ embed
8. What is the density of this polymer sample? numeric ☒ embed

Render

Selected Objects ☒ Show Full Text

Forward

139_a203_PaAcHe



ZoomPanSelect

Lasso

Zoom Fit

What is the molecular weight of this polymer sample in $M_w/10^{**5}$?

- | | | | | |
|----|--|---------|-------------------------------------|--------|
| 1. | Is this polymer sample biodegradeable? | bool | <input checked="" type="checkbox"/> | fill |
| 2. | Is this polymer sample flexible? | bool | <input type="checkbox"/> | text |
| 3. | Is this polymer sample based on curdian? | bool | <input checked="" type="checkbox"/> | stroke |
| 4. | What is the molecular weight of this polymer san | numeric | <input type="checkbox"/> | x |
| 5. | Will this polymer sample float on water? | bool | <input type="checkbox"/> | embed |
| 6. | Is this polymer sample going to save the world? | bool | <input type="checkbox"/> | embed |
| 7. | What is the density of this polymer sample? | numeric | <input type="checkbox"/> | embed |

Render

Selected Objects ☒ Show Full Text

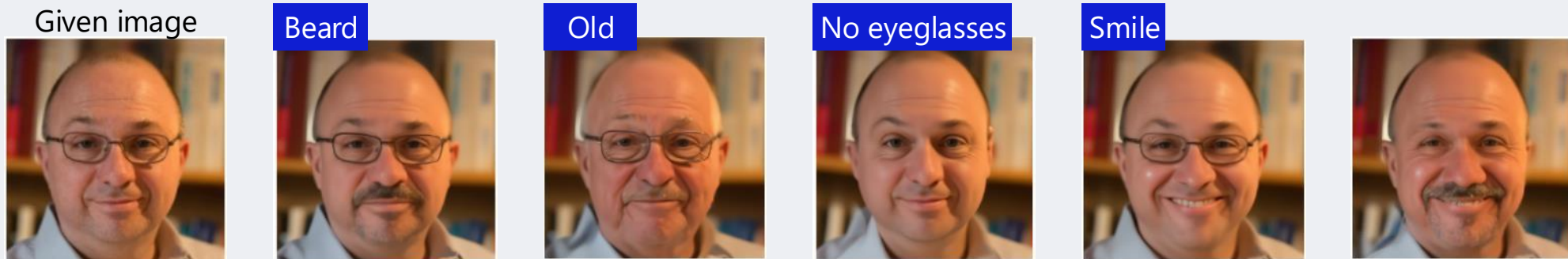
220 CdLa

This chemical name of this polymer sample is CdLa. It was made from the polysaccharide Curdlan. It is referenced in a paper titled "Effects of Long-Chain Acyl Substituents on the Thermoplasticity and Mechanical Properties of Paramylon Mixed Esters". The molecular weight of this sample is 23.55 Mw/10**5. The main chain is b13. Side chain 1 is Ac . Side chain 2 is La . The C number, branch and Degree of Substitution for acyl1 are: 2.0, 0.0 and 1.59 . The C number, branch and Degree of Substitution for acyl2 are: 12.0, 0.0 and 1.0 . The total DS is 2.59. Mw/Mn : 1.28, Tg(DSC)2 : 74.1, film/fiber : film, film/fiber type : hot-press, temperature for melting : 200.0, Tensile Strength(Mpa) : 6.5 +/- 19.9, Elongation at Break(%) : 6.9 +/- 3.2, Young's modulus(Gpa) : 0.6551 +/- 0.2389, Young's modulus(Gpa).1 : 0.434 +/- 0.0343, main : b13, [link](#)

AQA: 1) 0.8 2) 0.3 3) 1.0 4) 23.55 5) 0.5 6) 0.2 7) 0.0

Compositional Generation with Diffusion Probabilistic Models

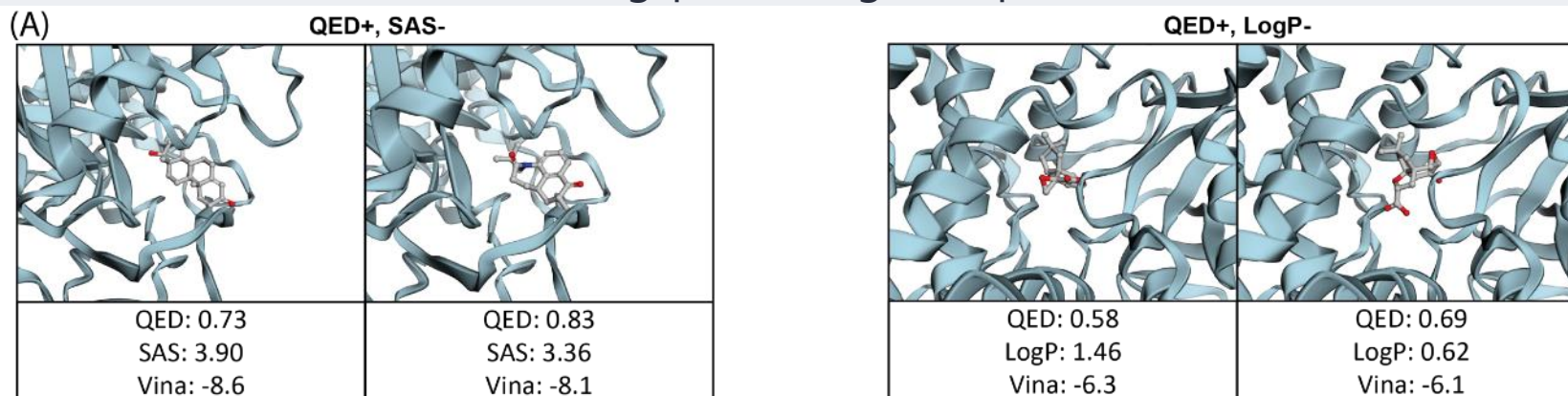
Compositionally manipulate multiple attributes in generation



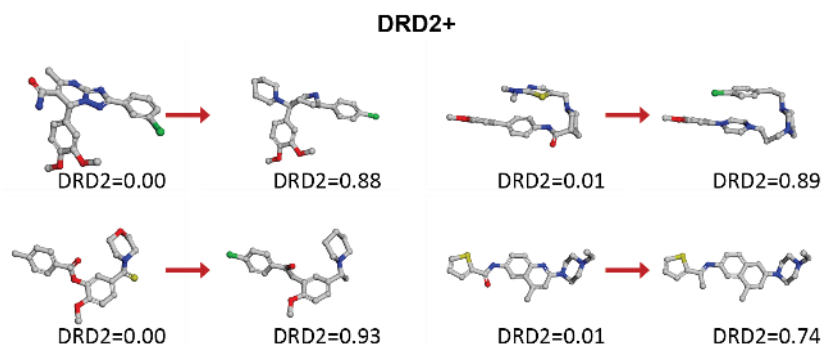
Shi *et al.*, Exploring Compositional Visual Generation with Latent Classifier Guidance. CVPR Workshop 2023.

Efficient Multi-Objective Manipulation of 3D Molecules

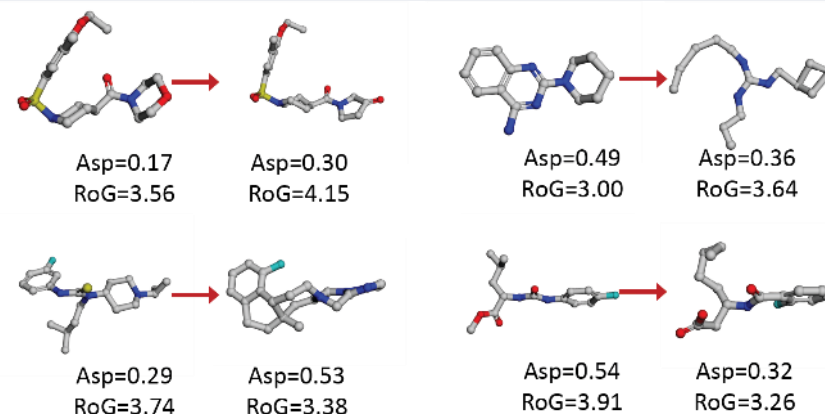
(1) Binding-preserving manipulation



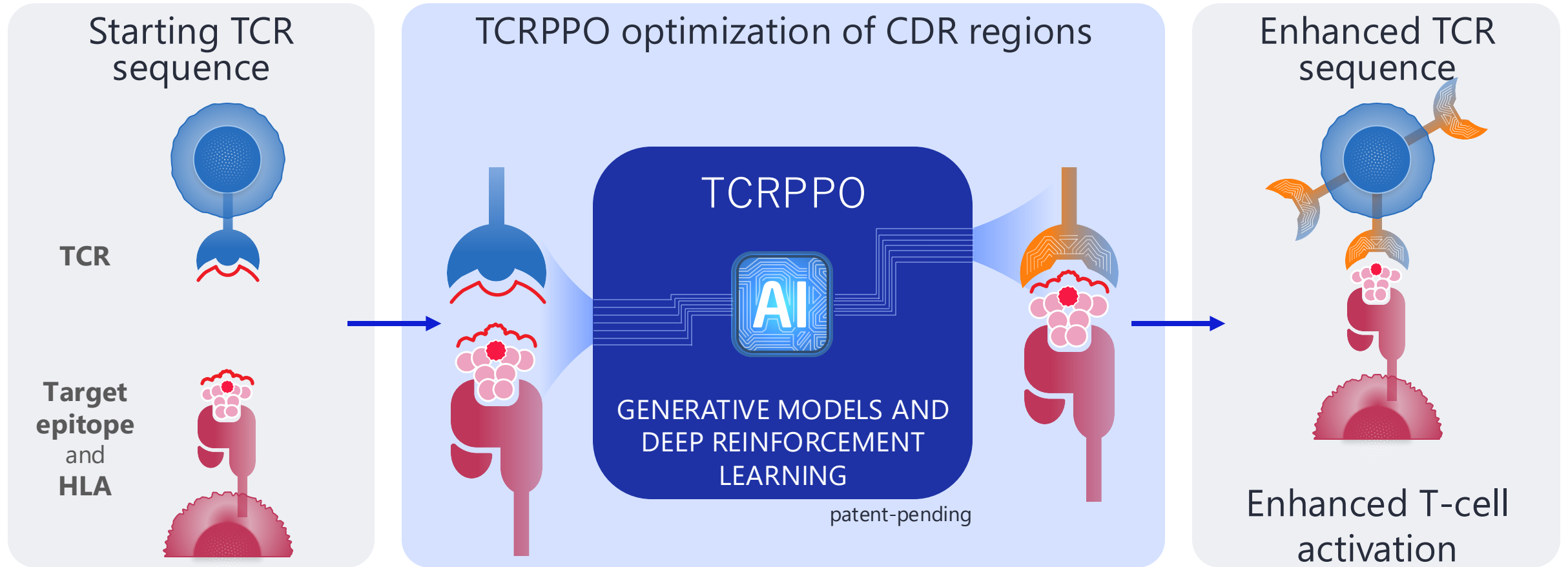
(2) Data efficient binding affinity enhancement (~1000 positive training data)



(3) Manipulation of 3D shapes and properties



Generative Models and RL for TCR Engineering

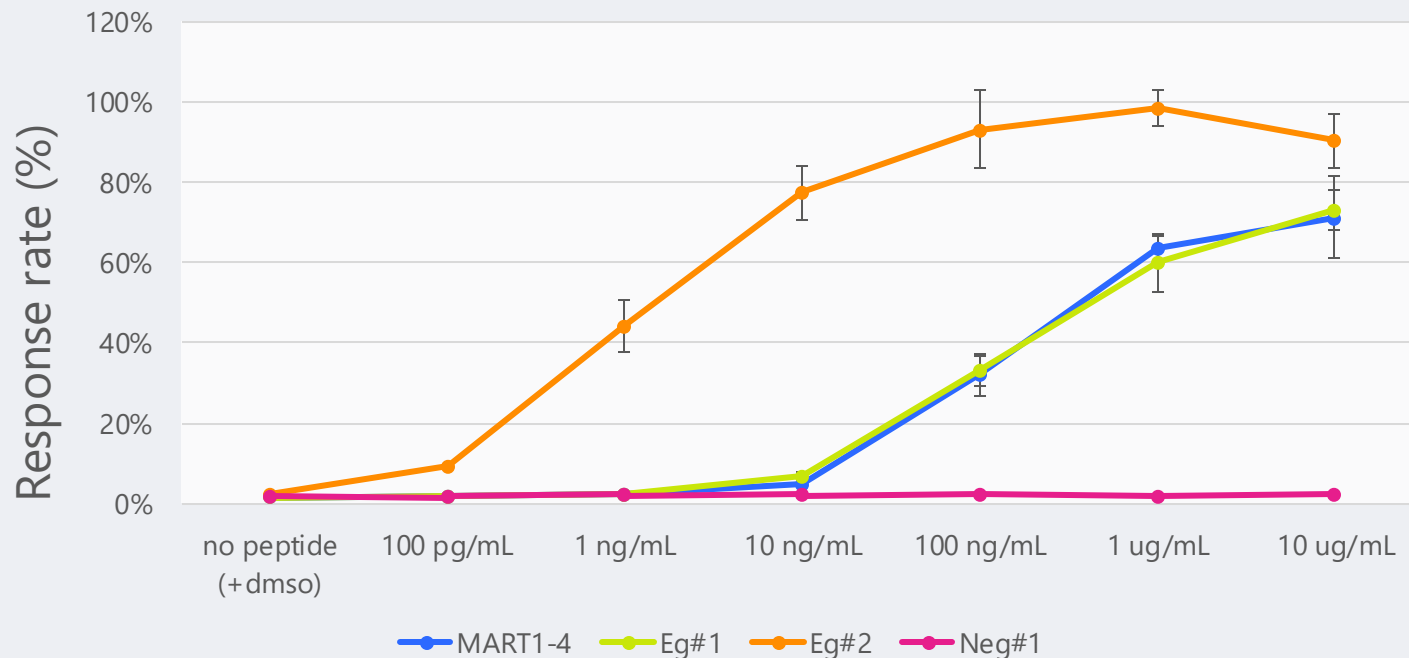


Chen *et al.*, T-Cell Receptor Optimization with Reinforcement Learning and Mutation Polices for Precision Immunotherapy. RECOMB 2023.
Min *et al.*, Design of enhanced TCR against cancer antigens using an AI system. SITC 2024.

In Vitro Validation Results

Mutate existing TCRs for better immunotherapy with reinforcement learning

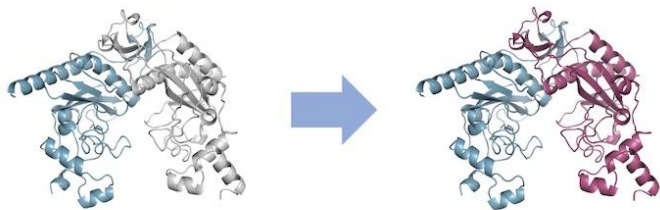
Cell activation after stimulation with different concentration of target peptide



From over 6.4 billion mutated TCR candidates, we selected 2 optimized TCRs targeting the melanoma antigen MART-1 for wet-lab validation, one engineered TCR is **over 1000 times more sensitive than the given template.**



Protein-Protein Complex Design

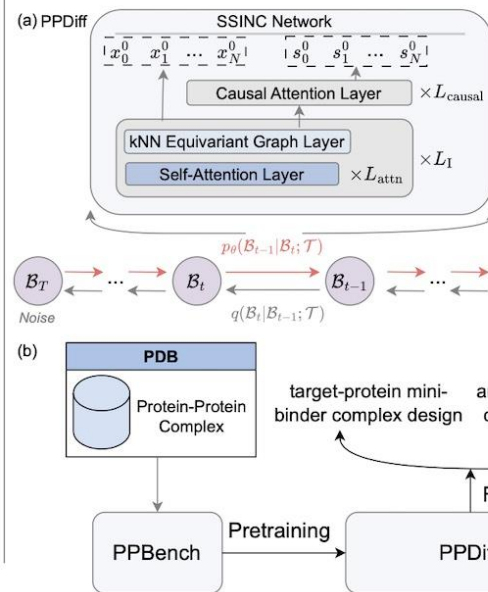


Given a target protein with its amino acid sequence and the alpha-carbon backbone structure, the goal is to designing proteins that bind to the target

Core idea

- Conditioning on target protein
- Co-designing binder protein sequence and backbone structure
- Hybrid (discrete-continuous) diffusion

PPDiff



B denotes binder protein

- x: alpha-carbon based backbone structure
- s: amino acid sequence

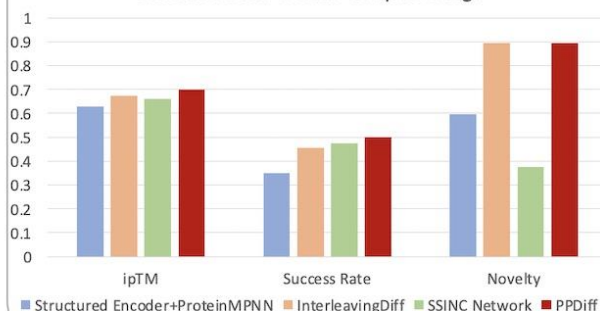
$L_{attn}: 11$ $L_I: 3$ $L_{causal}: 1$

PPBench

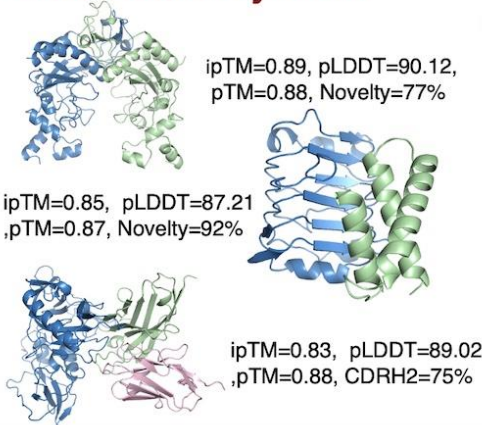
- 734,032 complexes
 - Training: 706149, Validation: 155, Test: 56
- Target Protein-Mini Binder Complex
 - 10 target proteins
 - 5 seen classes, 5 zero-shot
- Antigen-Antibody Complex
 - 4261 complexes

General Protein-Protein Complex Design

PPDiff achieves the highest success rate and novelty

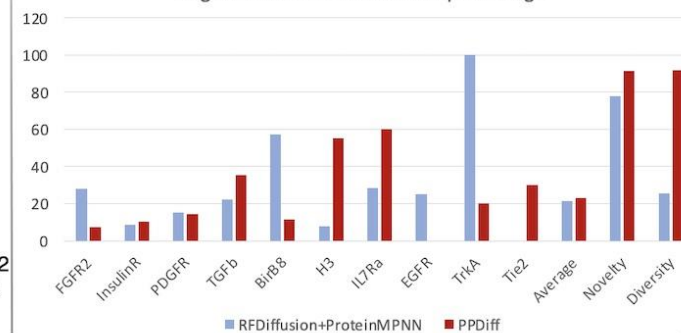


Cases designed by PPDiff



Target Protein-Mini Binder Complex Design

PPDiff achieves the higher success rate, diversity and novelty

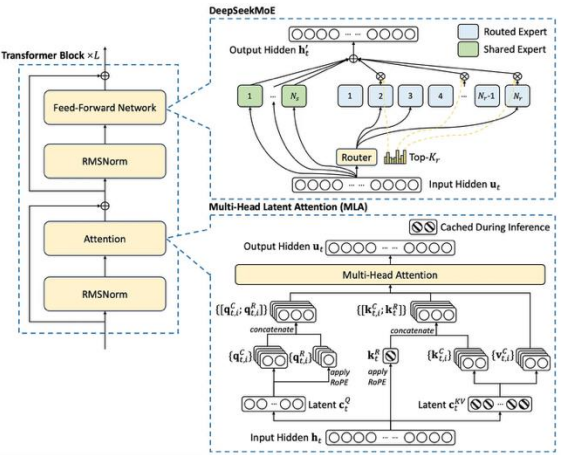
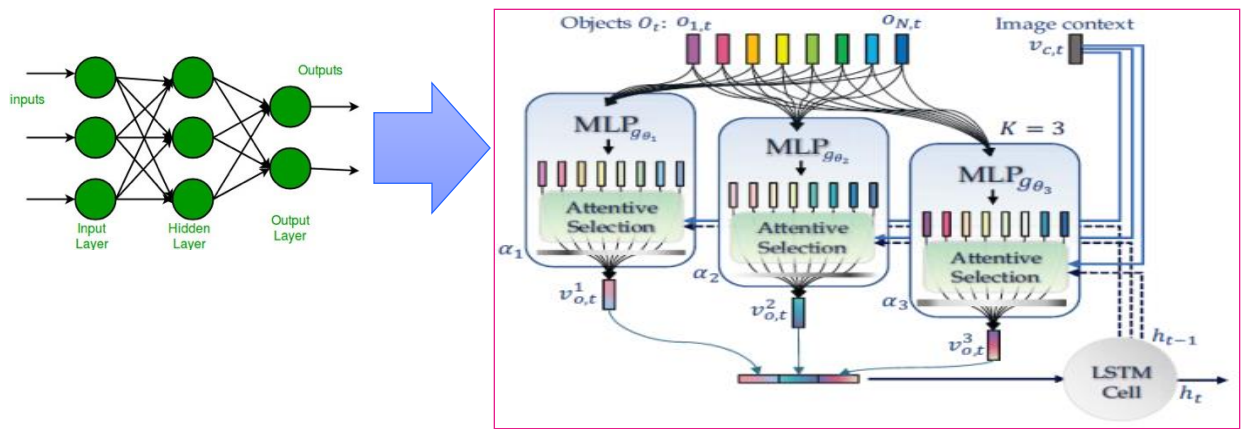


Future Directions

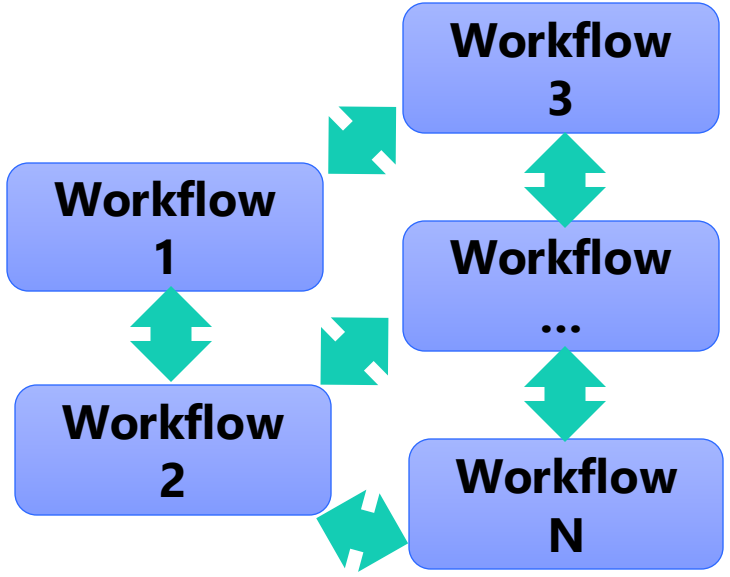
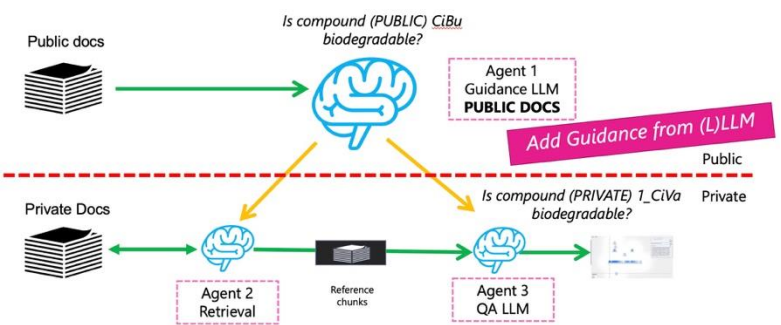
AI will make precision medicine scalable, accessible, and affordable worldwide.

What should we do?

Capture Tacit Knowledge into Workflows: From LLM to Structured Workflow AI




LLM



Digital Cell and Patient

Increasing efforts from both **pharmaceutical and IT companies** in building and utilizing digital twins of cells and patient models to reduce the trial-and-error of clinical research



Homepage > Stories > Our Science > Digital “Twinning”: Clinical Trials Powered by AI

Digital “Twinning”: Clinical Trials Powered by AI

Published on: May 22, 2024

Twin Health multiplies funding with \$50M for metabolic disease-reversing tech

By Andrea Park · Dec 14, 2023 12:45pm

Funding Summary

Last Round of Funding:

Series D \$50M in Dec 2023

Lead Investor: Temasek Holdings

Total Funding: \$248.5M

Digital twin generator Unlearn nets \$50M to bolster clinical trials with AI models

By Conor Hale · Apr 19, 2022 12:24pm

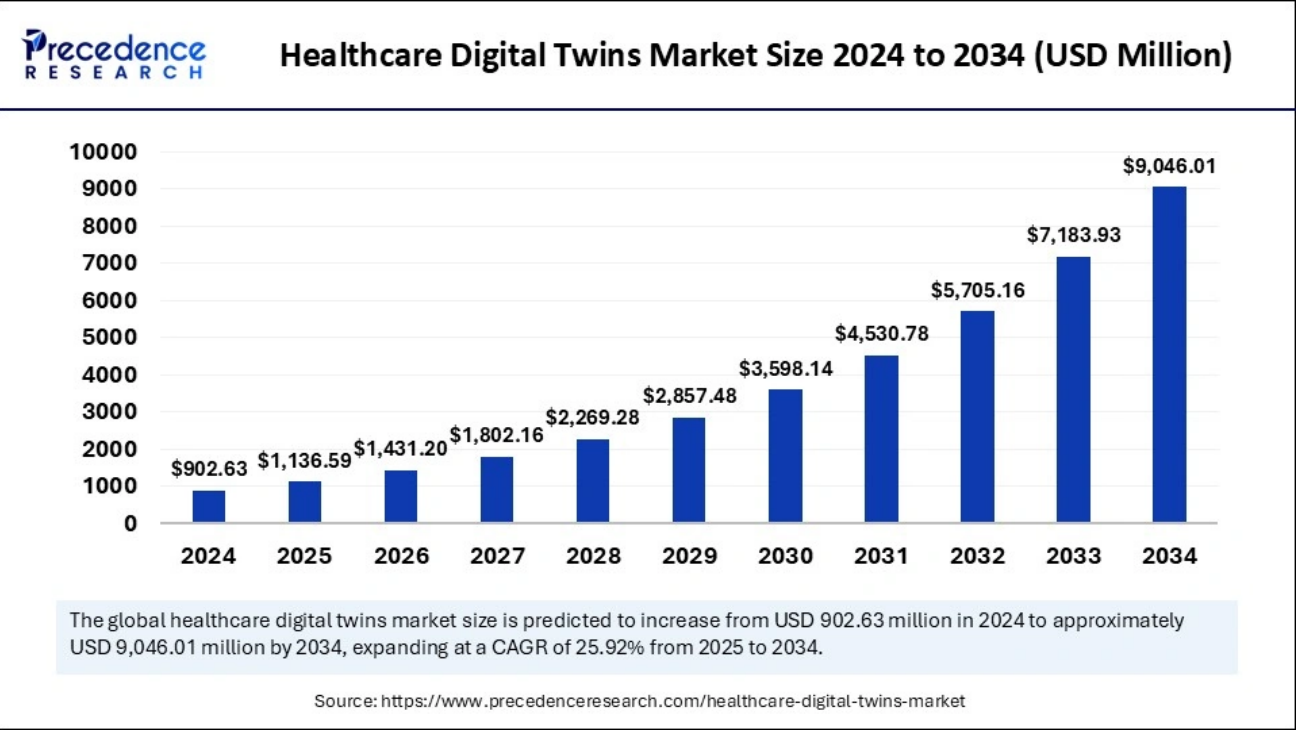
Funding Summary

Last Round of Funding:

Series C \$50M in Feb 2024

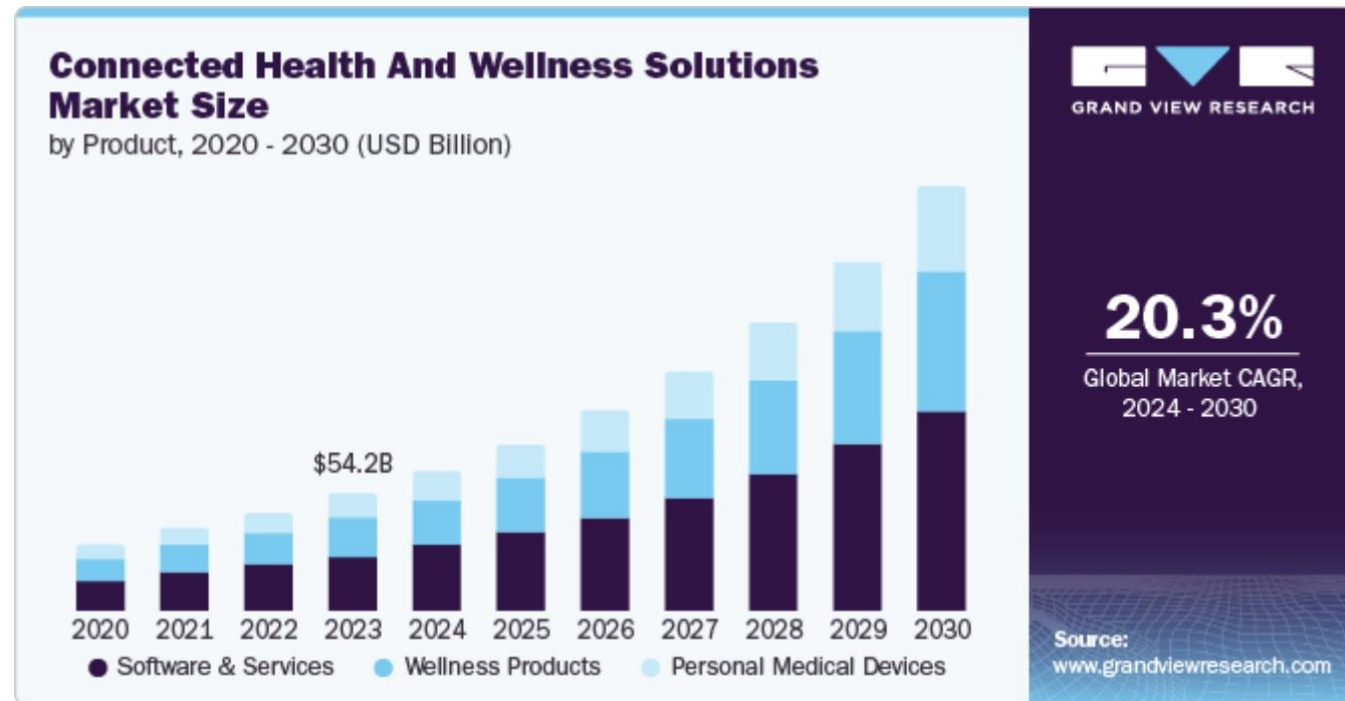
Lead Investor: Altimeter Capital

Total Funding: \$134.9M



Connected Health

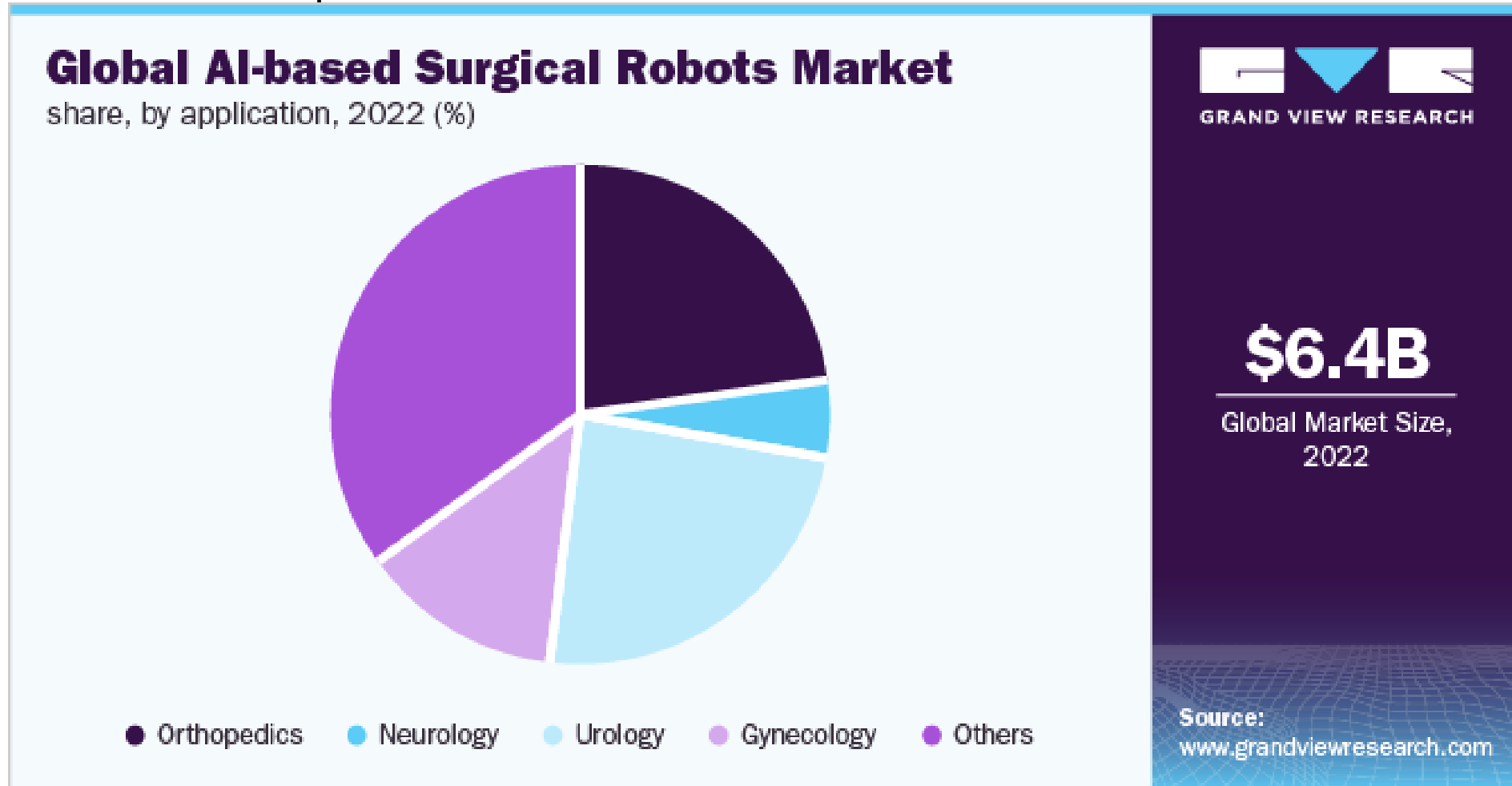
Primary care providers, virtual care providers, and specialist clinics will buy services to summarize incoming medical records, triage inquiries from patients and augment doctor-patient interaction



<https://www.grandviewresearch.com/industry-analysis/connected-health-wellness-devices-market>

AI-driven Robotic Surgery

AI-based surgical robots' segment was valued at \$6.4 billion in 2022 and is expected to expand at a CAGR of **18.9% from 2023 to 2030**



The Ultimate Future of Machine Learning with “Simulated Consciousness”

From Generative Digital Cell and Digital Human to

Mike Peter Robot



**Bill Gates Professor of Medicine,
Molecular Biophysics & Biochemistry
and Computer Science**

A.B. 2039, Harvard University

Ph.D. 2043, Cambridge University

Joined Yale Faculty 2047

Professor Robot does research in the new field of Rapid Human Disease Treatment and Prevention. He won the Nobel Prize in Physiology for discovering ABCBBCNBC007 for the no-pain one-smell treatment of cancer in 2067.

Most of this slide was shown during a talk at Yale in 2015

The Ultimate Future of Machine Learning with “Simulated Consciousness”

From Generative Digital Cell and Digital Human to
Of the Human, By the Human, For the Human



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Acknowledgements

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Youzhi Luo (TAMU)
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Prof. Lei Li (CMU)
Prof. Xia Ning (OSU)
Prof. Dongjin Song (UConn)
Prof. Jing Zhang (UCI)

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