



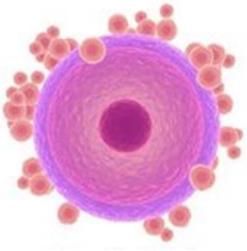
2026 PCRS ANNUAL MEETING

REPRODUCTIVE FRONTIERS: BRIDGING BIOLOGY,
PRACTICE, AND POSSIBILITY

MARCH 18-22 | RANCHO MIRAGE, CA



PACIFIC COAST
REPRODUCTIVE
SOCIETY



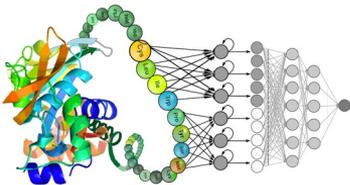
AI-Generated Therapeutics for Fertility and Medicine

Pranam Chatterjee, PhD

Assistant Professor of Bioengineering

Assistant Professor of Computer and Information Science

University of Pennsylvania



Disclosures

- Nothing to Disclose

Needs Assessment Statement

- Rapid advances in stem cell engineering, genome editing, and generative artificial intelligence are reshaping molecular therapeutics. Reproductive specialists require a clear understanding of how these technologies integrate to enable programmable control of genes, proteins, and cellular states in fertility and related diseases.

Expected Learning Outcomes

- **Describe** how ovarian support cell technologies improve oocyte maturation and clinical outcomes.
- **Interpret** clinical trial data comparing Fertilo with traditional IVM.
- **Analyze** how cell-state engineering can enhance reproductive success.
- **Evaluate** the potential role of AI-designed peptides and molecules in fertility and other reproductive diseases.

My hybrid lab!

Computational Team



Liz



Yinuo



Sophie



Tong



Aastha



Rosie



Sophia



Shrey

Experimental Team



Lin



Zach



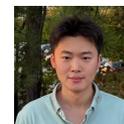
Lauren



Tian



Divya



Howard



Yesol

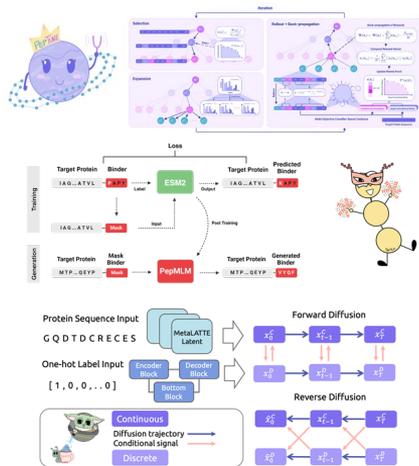


Sumi



Jiale

Generative Algorithms



My hybrid lab!

Computational Team



Liz Yinuo Sophie Tong



Aastha Rosie Sophia Shrey

Experimental Team

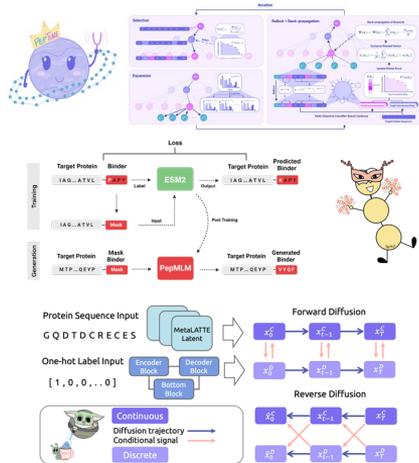


Lin Zach Lauren Tian Divya

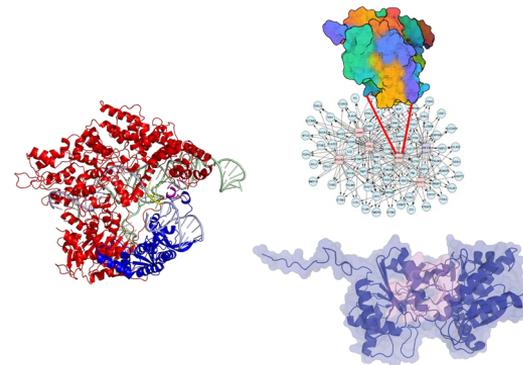


Howard Yesol Sumi Jiale

Generative Algorithms

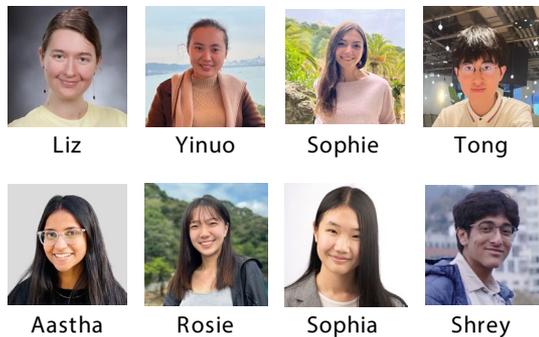


Protein Therapeutics

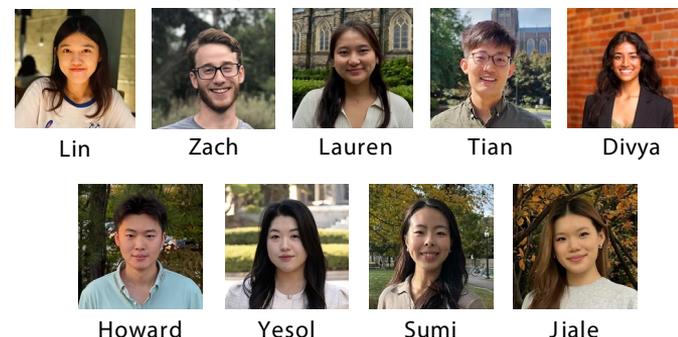


My hybrid lab!

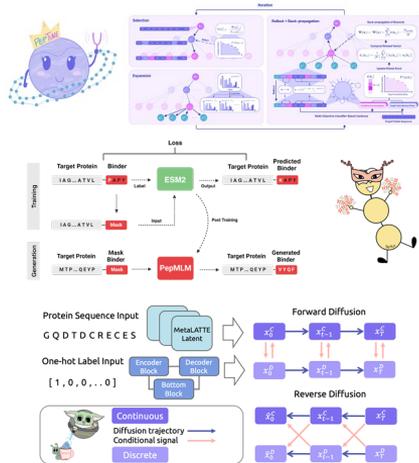
Computational Team



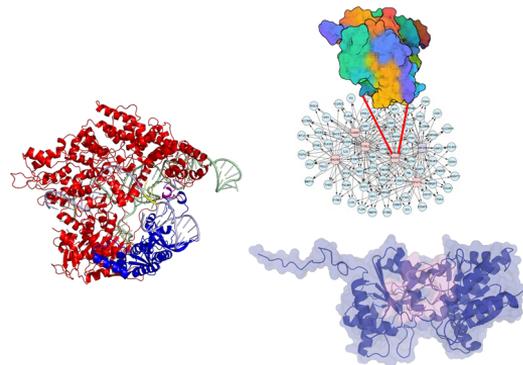
Experimental Team



Generative Algorithms



Protein Therapeutics



My hybrid lab!

Theoretical Frameworks

Applied Algorithms

Translation-Specific Models

Protein Biochemistry

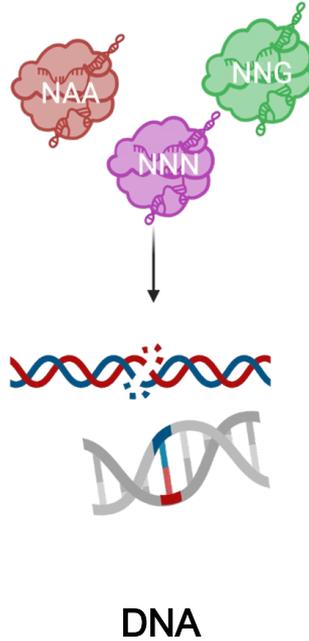
High-Throughput Screening

In vivo Disease Models

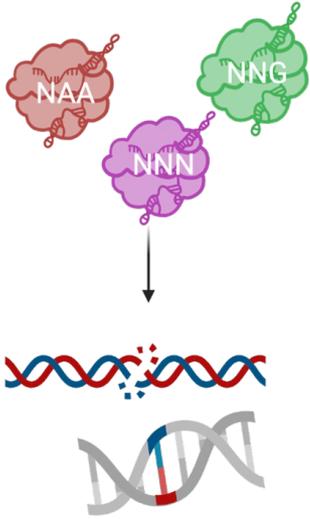


Proteins to Engineer Biology Programmably

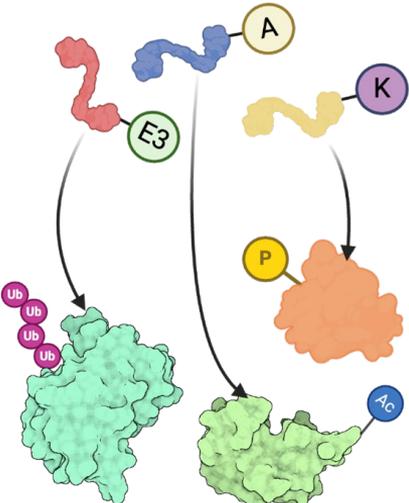
Proteins that can edit any DNA sequence



Proteins that can bind and modify any other protein

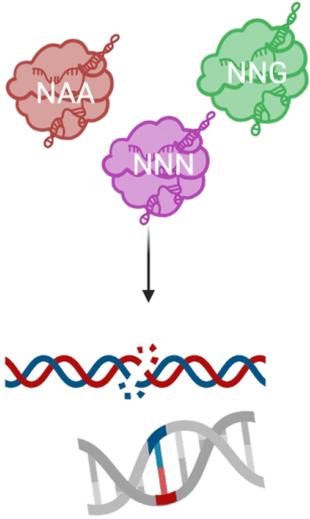


DNA

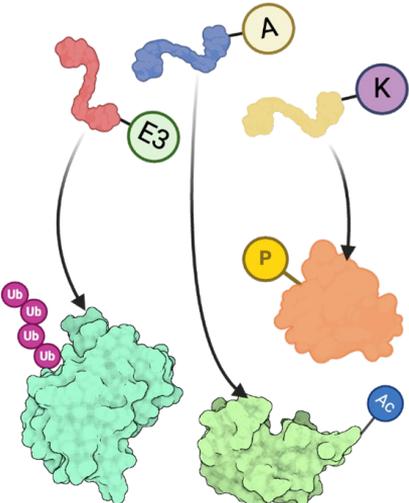


Proteins

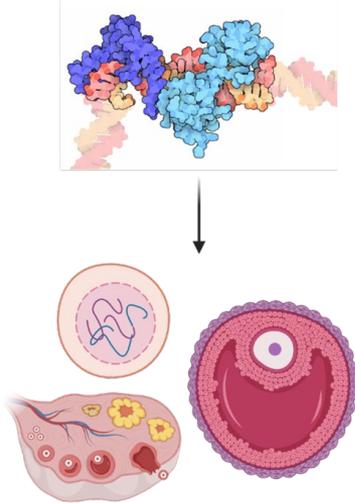
Proteins that can generate any new cell type



DNA

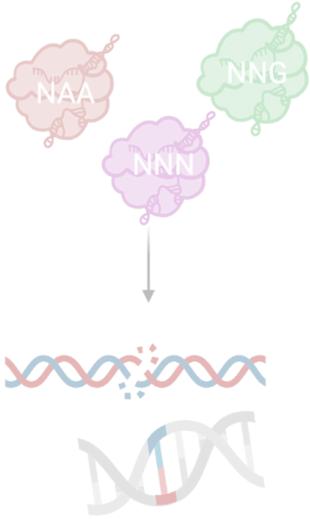


Proteins

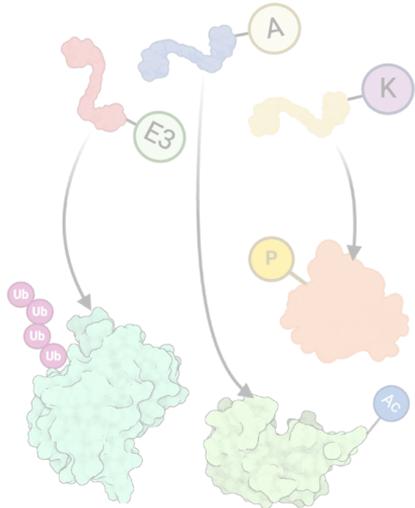


Cells/Tissues

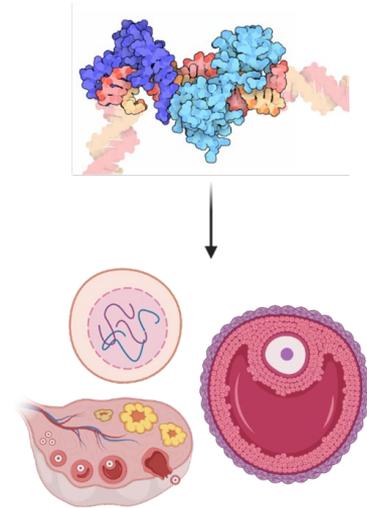
Proteins that can generate any new cell type



DNA



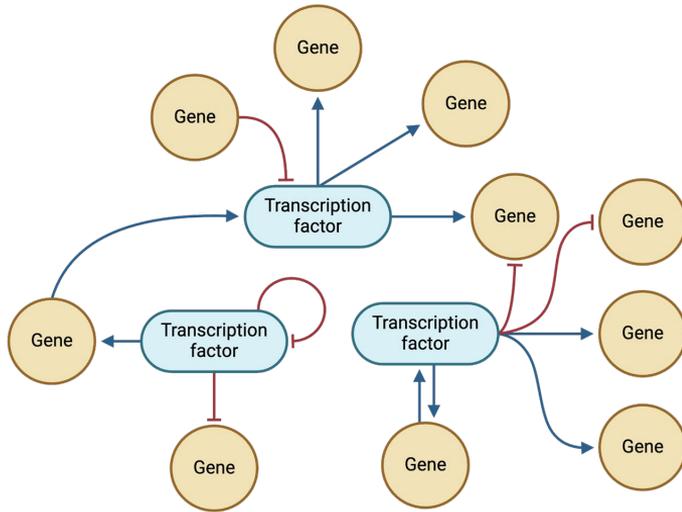
Proteins



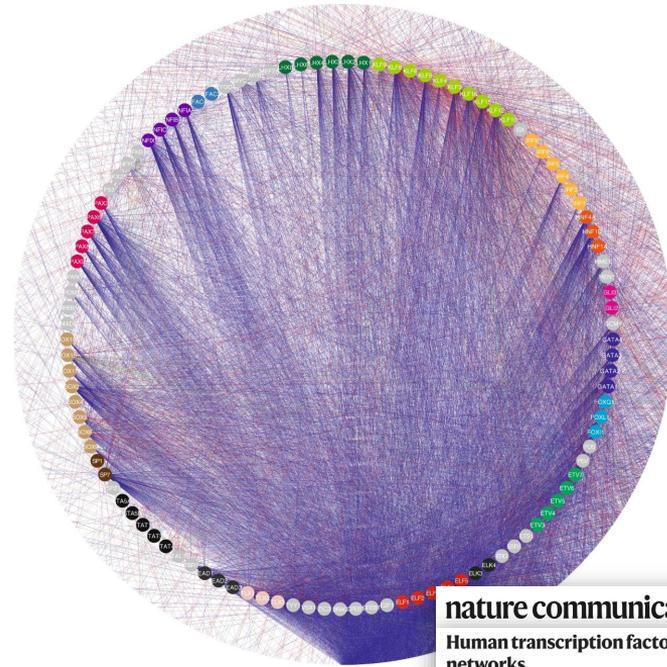
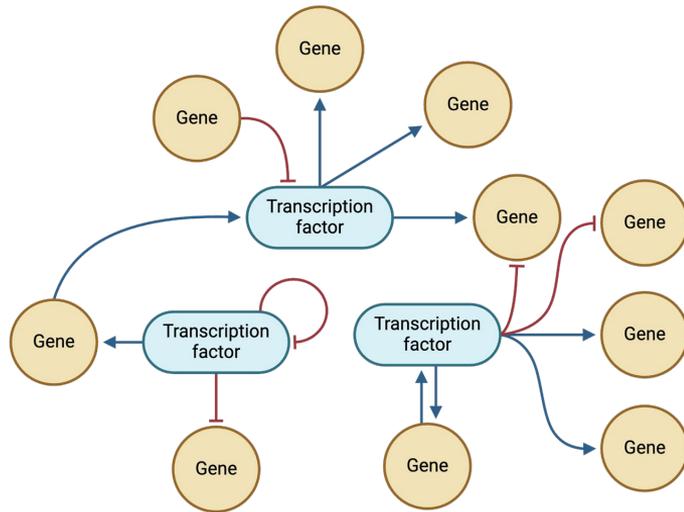
Cells/Tissues

Transcription Factors: Regulators of Cell State

Transcription Factors: Regulators of Cell State



Transcription Factors: Regulators of Cell State

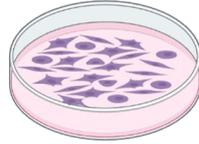


nature communications
Human transcription factor protein interaction networks
Helka Góös, Matias Kinnunen, Kari Salokas, Zenolai Tan, Xiaonan Liu, Leena Yadav, Qin Zhang, Gong-Hong Wei & Markku Varjosalo

TFs can be great tools for cell engineering

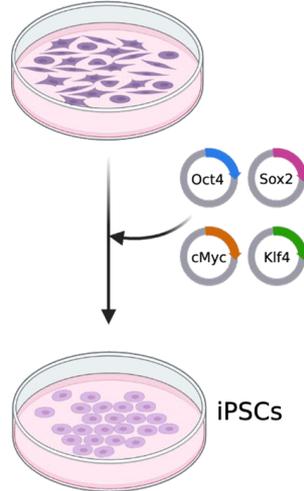
TFs can be great tools for cell engineering

Adult Fibroblast Cells

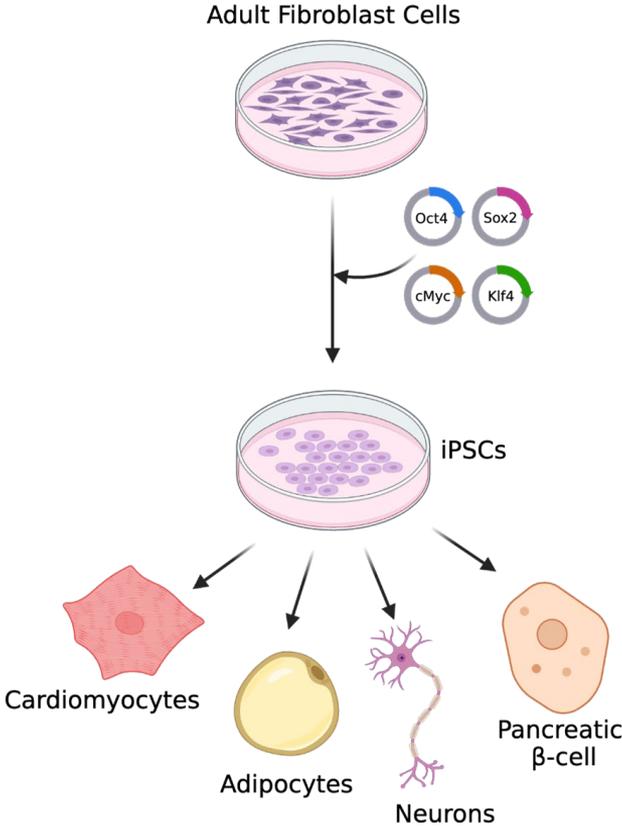


TFs can be great tools for cell engineering

Adult Fibroblast Cells



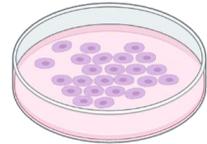
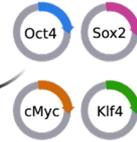
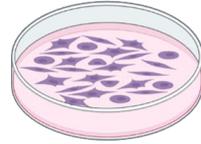
TFs can be great tools for cell engineering



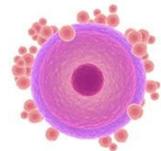
Fibroblasts		POU5F1, SOX2, KLF4 POU5F1, SOX2	
Keratinocytes		POU5F1, SOX2, LIN28A, NANOG POU5F1, SOX2 POU5F1, SOX2, KLF4 POU5F1, SOX2, KLF4, MYC	
Neural stem cells		POU5F1 POU5F1, KLF4	
Hematopoietic Stem cells		POU5F1, SOX2, KLF4 POU5F1, SOX2	
Fibroblasts		HNF1A, HNF4A, ONECUT1, CEBPA, ATFS, PROX1, MYC, TP53-siRNA HNF1A, HNF4A, FOXA3 FOXA2, HNF4A, CEBPB, MYC FOXA2, HNF4A, CEBPB	
Foreskin Fibroblasts		CBX2, HES1, ID1, TFAP2A, ZFP42, ZNF423 ZNF521	
Fibroblasts (NHDF)		SOX2, PAX6	
Hematopoietic Stem cells		SOX2	
Fibroblasts		POU3F2, ASCL1, MYT1L POU3F2, ASCL1, NEUROD1	
Embryonic stem cells		NEUROD2 POU3F2, ASCL1, MYT1L	
Fibroblasts		MYO1D1	
Fibroblasts		MITF, PAX3, SOX10	
Keratinocytes		MITF, LEF1, SOX10, SOX9 MITF, LEF1, SOX10, SOX9, PAX3, SOX2	
Mesenchymal stem cells		CEBPB PPARG CEBPB, PPARG	

TFs can be great tools for cell engineering

Adult Fibroblast Cells

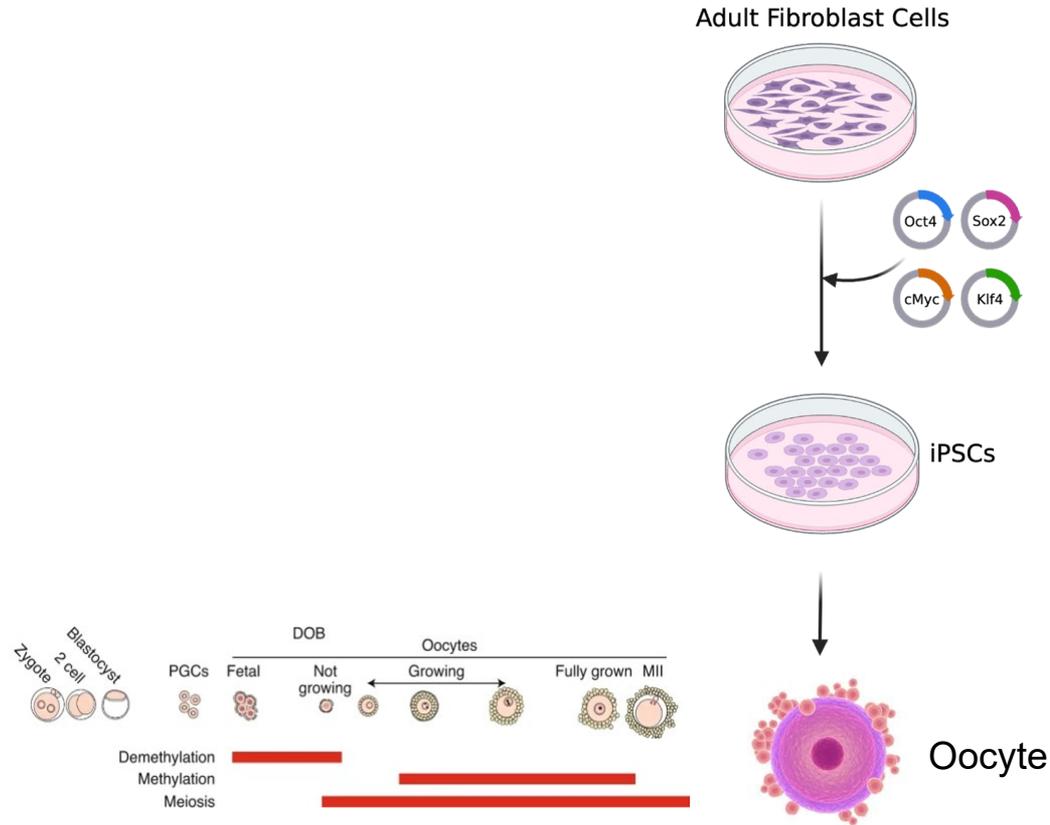


iPSCs

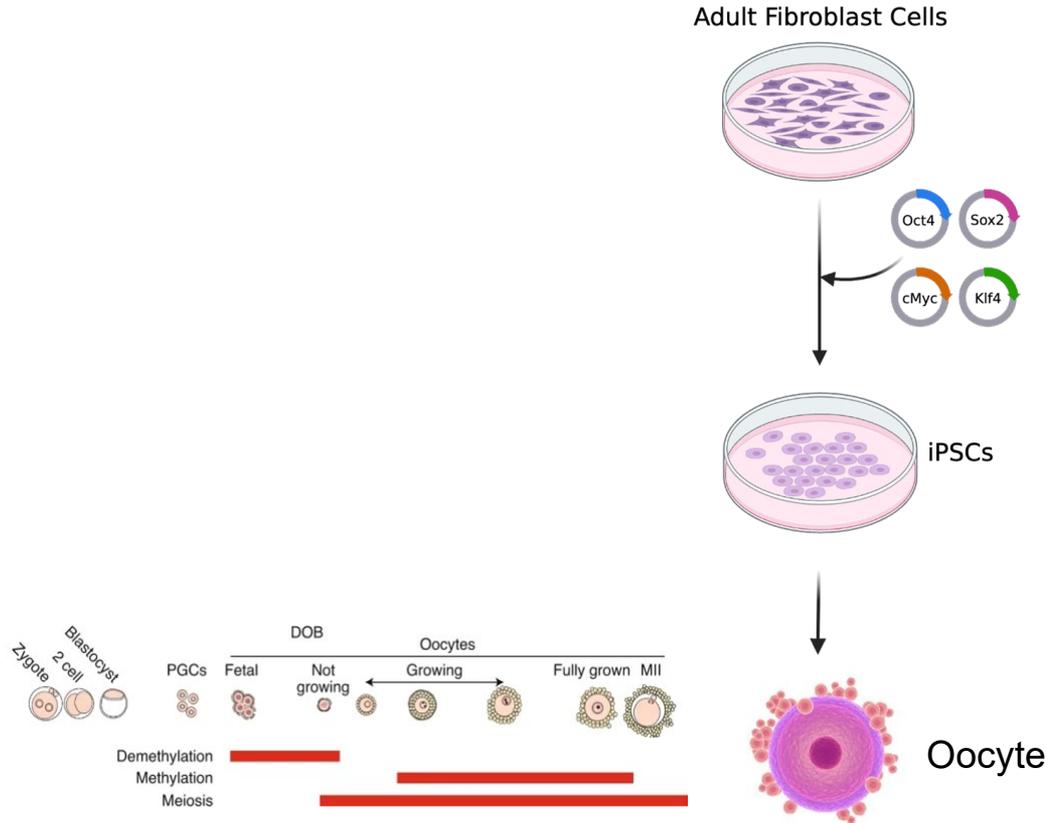


Oocyte

TFs can be great tools for cell engineering

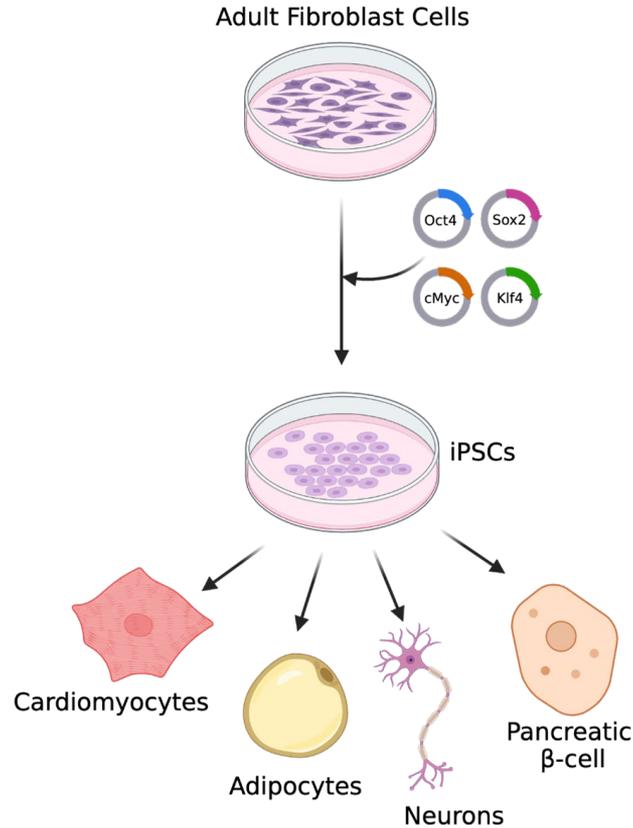


TFs can be great tools for cell engineering



Research Question:
What are the sufficient set of transcription factors for oogenesis?

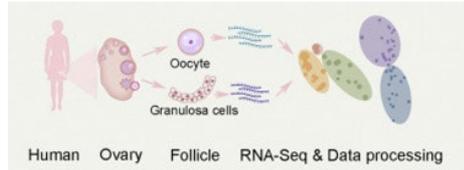
TFs can be great tools for cell engineering



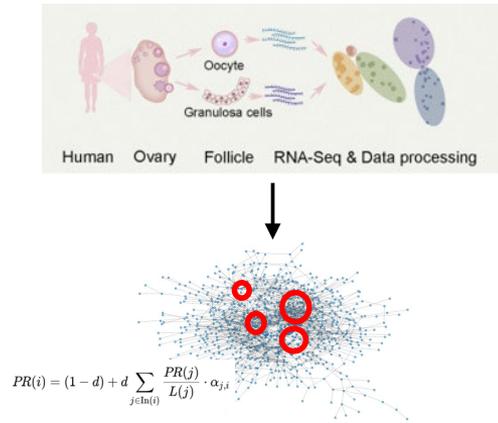
But first:

How do we systematically identify transcription factors that induce new cell states from iPSCs?

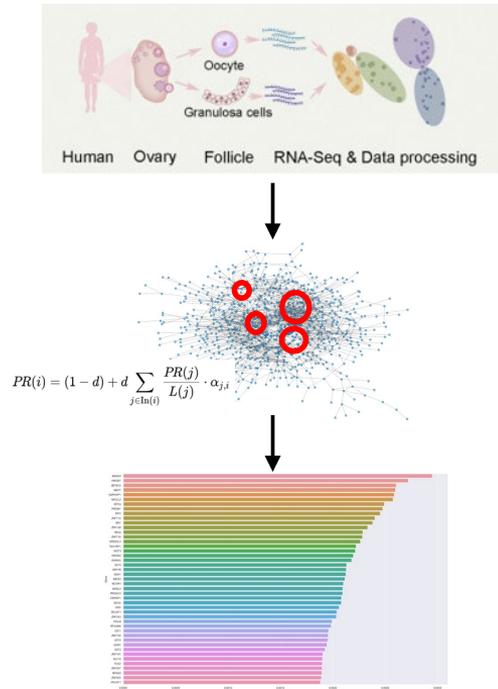
RNA-seq data of natural oocyte differentiation



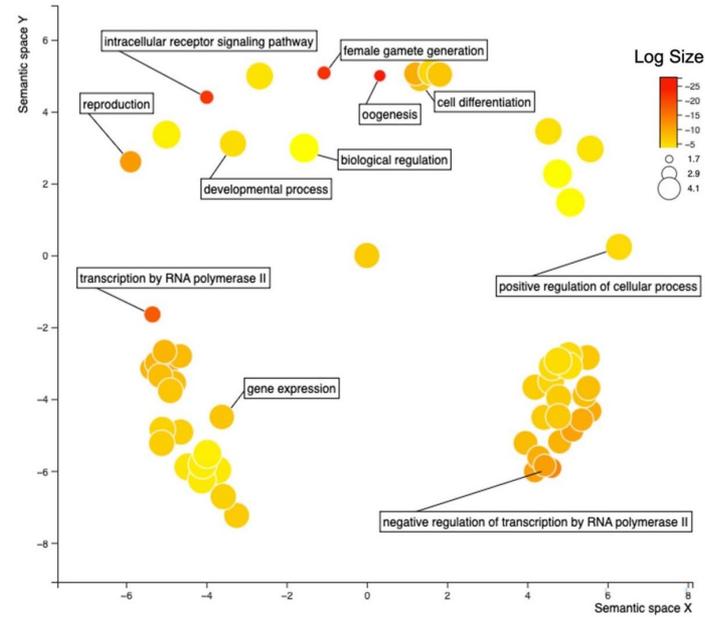
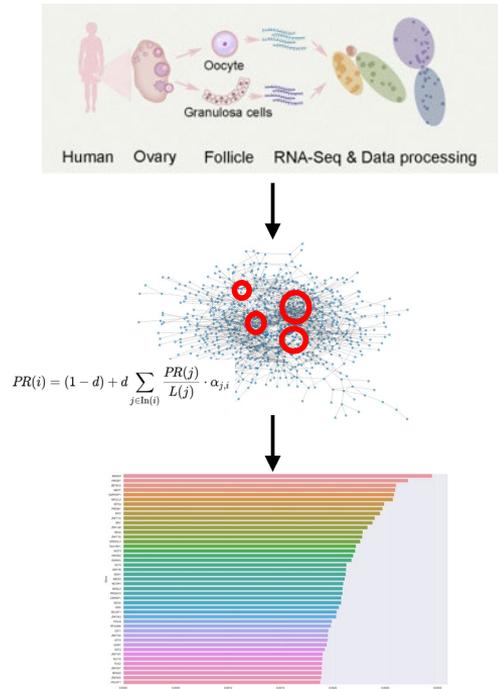
Graph theory algorithm to prioritize regulators



Prioritized list of TFs that drive oocyte formation

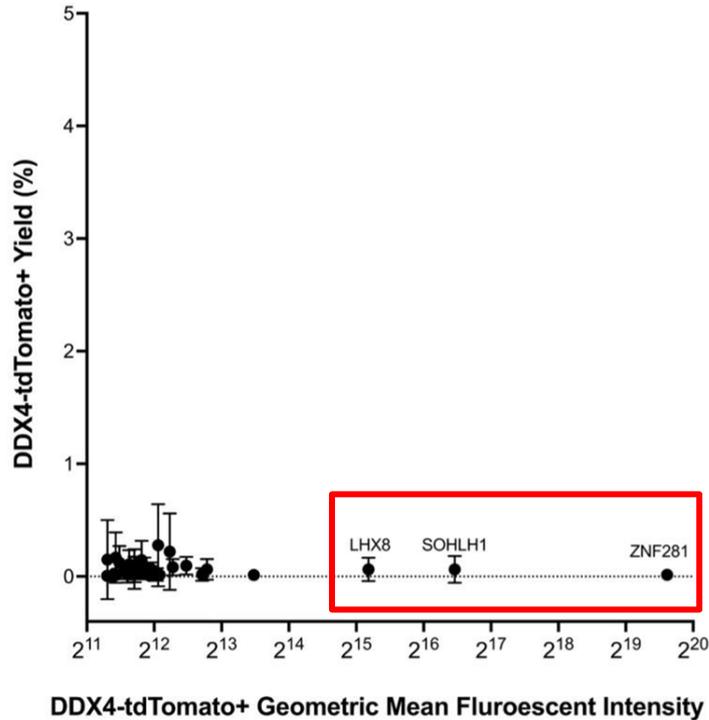


TFs with critical roles in germ cell development and cell differentiation!



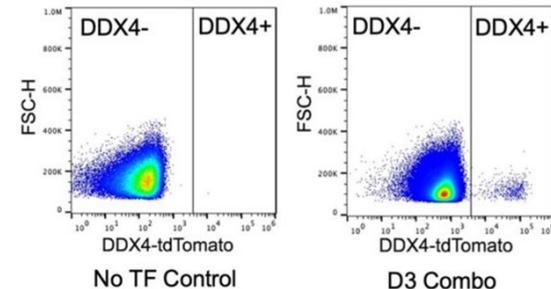
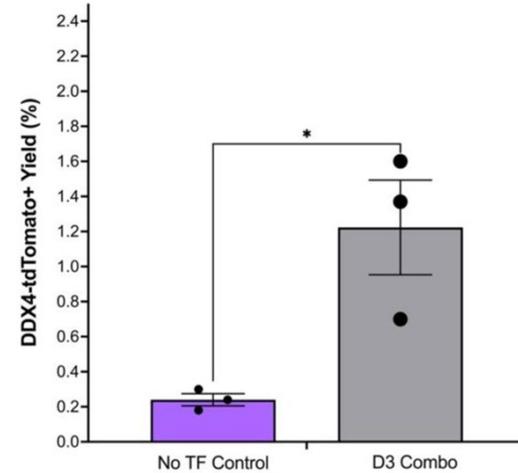
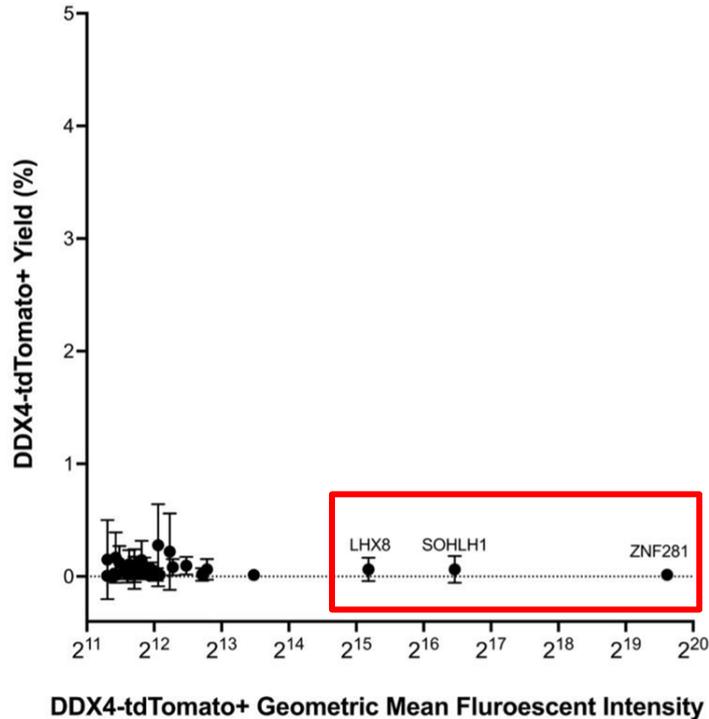


Three TFs that drive DDX4 expression!



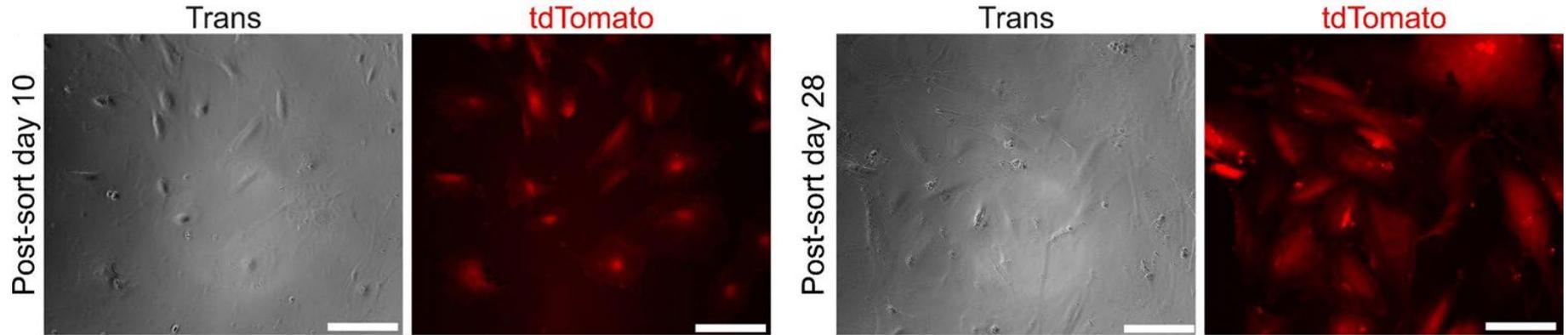


Combining them yielded higher percentages of these induced oogonia-like cells (iOLCs)



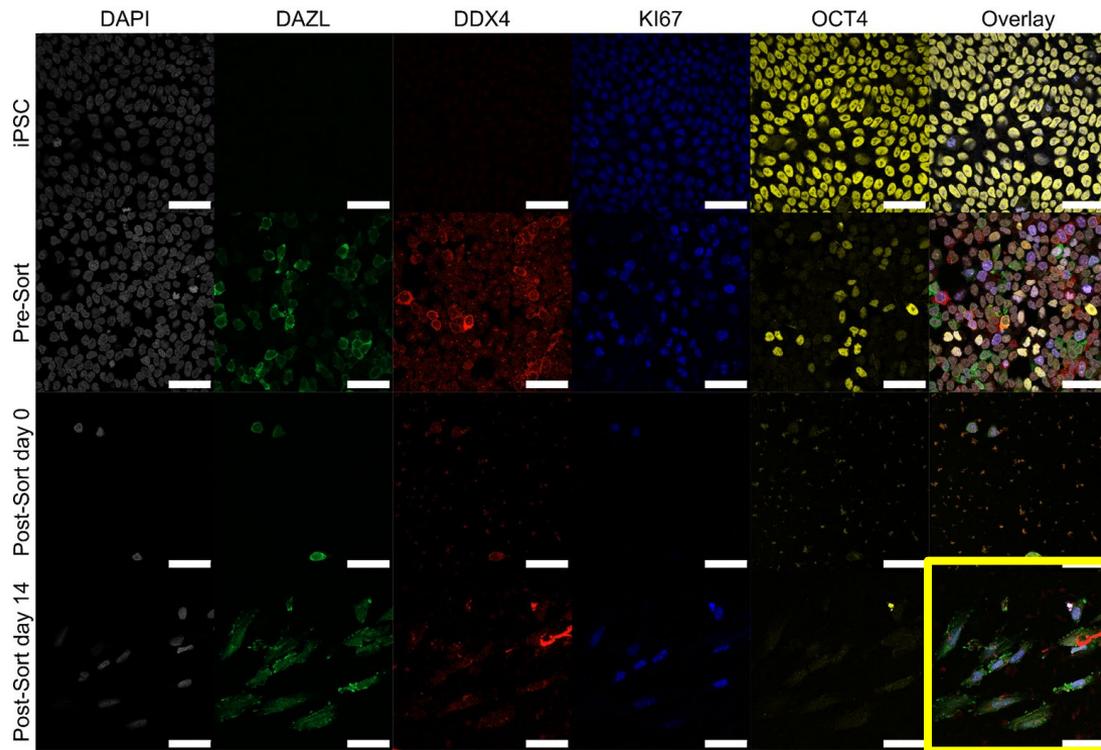


iOLCs expand post-isolation and display continued DDX4 protein expression





iOLCs show continued oogonia -like protein expression



Read about our story to make iOLCs!

Article



SOURCE
DATA



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PROCESS



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EMBO
reports

Rapid human oogonia-like cell specification via transcription factor-directed differentiation

Merrick Pierson Smela^{1,2}, Christian C Kramme ^{1,2}✉, Patrick R J Fortuna^{1,2}, Bennett Wolf^{1,2}, Shrey Goel³, Jessica Adams^{1,2}, Carl Ma ^{1,2}, Sergiy Velychko ^{1,2}, Ursula Widocki ⁴, Venkata Srikar Kavirayuni³, Tianlai Chen³, Sophia Vincoff³, Edward Dong^{1,2}, Richie E Kohman ^{1,2}, Mutsumi Kobayashi ⁵, Toshi Shioda ⁶, George M Church^{1,2} & Pranam Chatterjee^{3,7,8}



We have also generated ovarian support cells (OSCs) from iPSCs!



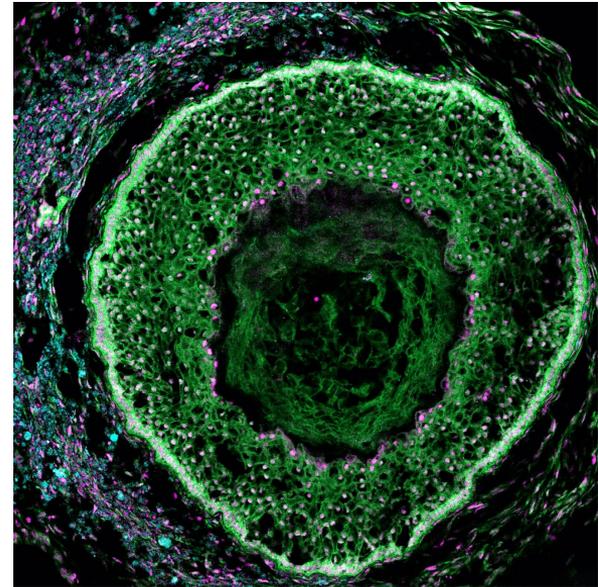
Research Article

[Developmental Biology](#), [Stem Cells and Regenerative Medicine](#)

Directed differentiation of human iPSCs to functional ovarian granulosa-like cells via transcription factor overexpression

Merrick D Pierson Smela, Christian C Kramme, Patrick RJ Fortuna, Jessica L Adams, Rui Su, Edward Dong, Mutsumi Kobayashi, Garyk Brixi, Venkata Srikar Kavirayuni, Emma Tysinger, Richie E Kohman, Toshi Shioda, Pranam Chatterjee, George M Church [see less](#)

Wyss Institute, Harvard University, United States; Department of Genetics, Harvard Medical School, United States; Massachusetts General Hospital Center for Cancer Research, Harvard Medical School, United States; Department of Biomedical Engineering, Duke University, United States; Department of Computer Science, Duke University, United States



With these technologies, we started **Gameto**.



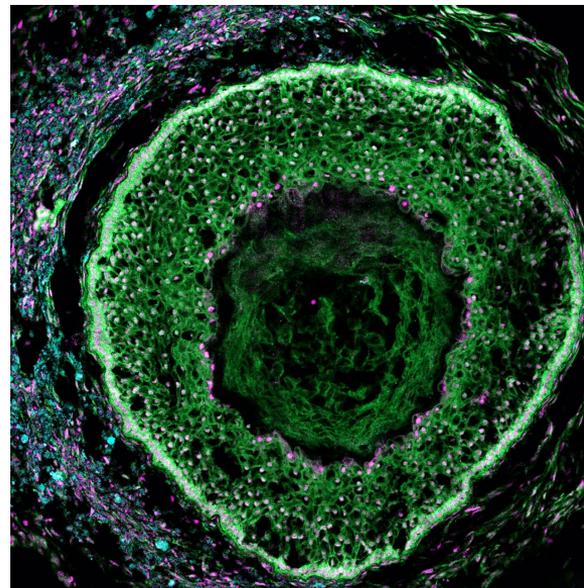
Research Article

[Developmental Biology, Stem Cells and Regenerative Medicine](#)

Directed differentiation of human iPSCs to functional ovarian granulosa-like cells via transcription factor overexpression

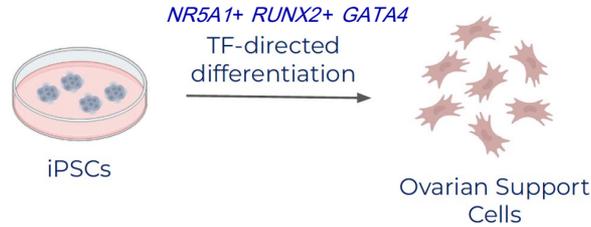
Merrick D Pierson Smela, Christian C Kramme, Patrick RJ Fortuna, Jessica L Adams, Rui Su, Edward Dong, Mutsumi Kobayashi, Garyk Brixi, Venkata Srikar Kavirayuni, Emma Tysinger, Richie E Kohman, Toshi Shioda, Pranam Chatterjee, George M Church [see less](#)

Wyss Institute, Harvard University, United States; Department of Genetics, Harvard Medical School, United States; Massachusetts General Hospital Center for Cancer Research, Harvard Medical School, United States; Department of Biomedical Engineering, Duke University, United States; Department of Computer Science, Duke University, United States





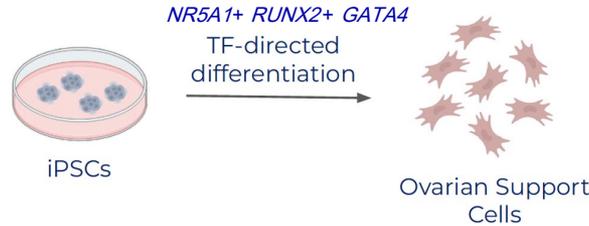
Gameto's **Fertilo** program uses proprietary OSCs to rapidly mature oocytes with minimal hormonal stimulations





Gameto's **Fertilo** program uses proprietary OSCs to rapidly mature oocytes with minimal hormonal stimulations

- OSGIVMs are primarily FOXL2+ AMHR2+ NR2F2+ granulosa-like cells.



Research Article
Developmental Biology, Stem Cells and Regenerative Medicine

Directed differentiation of human iPSCs to functional ovarian granulosa-like cells via transcription factor overexpression

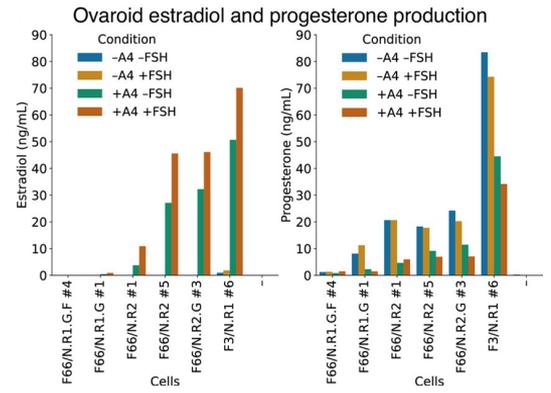
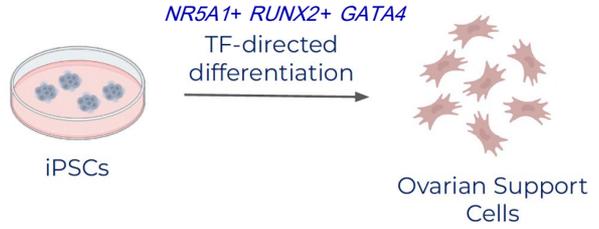
Merrick D Pierson Smela, Christian C Kramme, Patrick RJ Fortuna, Jessica L Adams, Rui Su, Edward Dong, Mutsumi Kobayashi, Garyk Brixi, Venkata Srikar Kavirayuni, Emma Tysinger, Richie E Köhman, Toshi Shioda, Pranam Chatterjee, George M Church [see less](#)

Wyss Institute, Harvard University, United States; Department of Genetics, Harvard Medical School, United States; Massachusetts General Hospital Center for Cancer Research, Harvard Medical School, United States; Department of Biomedical Engineering, Duke University, United States; Department of Computer Science, Duke University, United States



Gameto's Fertilo program uses proprietary OSCs to rapidly mature oocytes with minimal hormonal stimulations

- OSGIVMs are primarily FOXL2+ AMHR2+ NR2F2+ granulosa-like cells.
- They are steroidogenic, producing aromatase in response to FSH stimulation




Research Article
Developmental Biology, Stem Cells and Regenerative Medicine

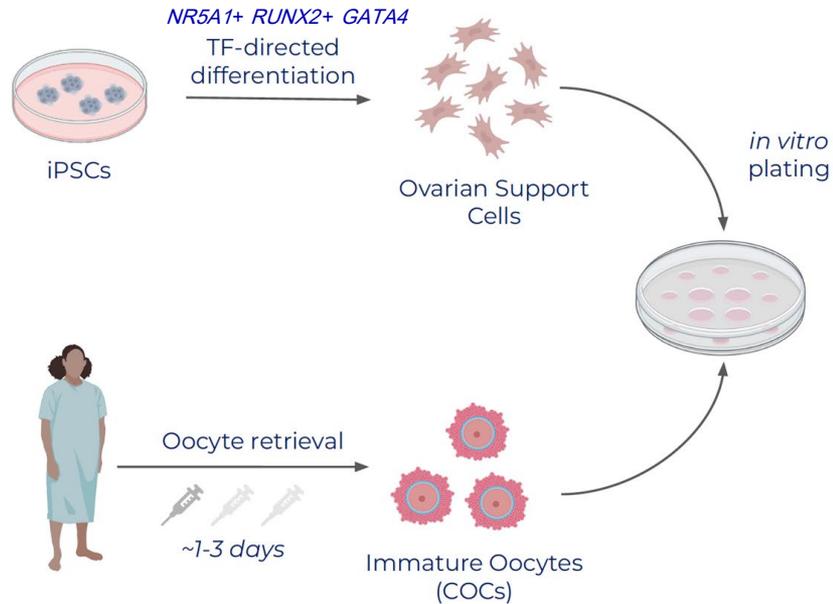
Directed differentiation of human iPSCs to functional ovarian granulosa-like cells via transcription factor overexpression

Merrick D Pierson Smela, Christian C Kramme, Patrick Rj Fortuna, Jessica L Adams, Rui Su, Edward Dong, Mutsumi Kobayashi, Garyk Brixi, Venkata Srikar Kavirayuni, Emma Tysinger, Richie E Kohman, Toshi Shioda, Pranam Chatterjee, George M Church [see less](#)

Wyss Institute, Harvard University, United States; Department of Genetics, Harvard Medical School, United States; Massachusetts General Hospital Center for Cancer Research, Harvard Medical School, United States; Department of Biomedical Engineering, Duke University, United States; Department of Computer Science, Duke University, United States

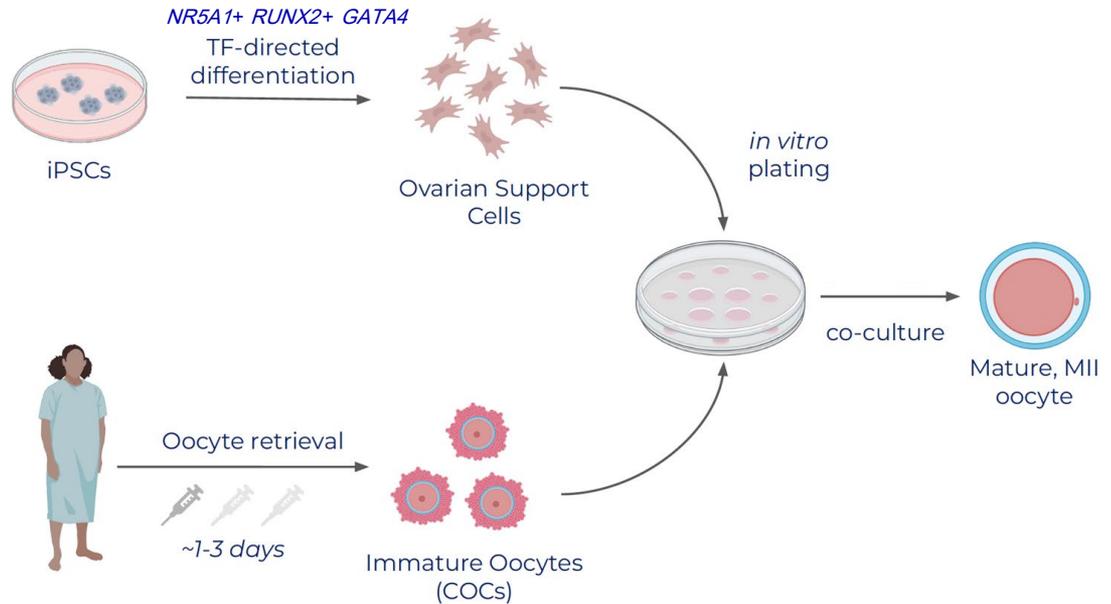
Gameto's **Fertilo** program uses proprietary OSCs to rapidly mature oocytes with minimal hormonal stimulations

- OSC-IVMs are primarily FOXL2+ AMHR2+ NR2F2+ granulosa-like cells.
- They are steroidogenic, producing aromatase in response to FSH stimulation
- Produce the necessary growth factors needed for robust paracrine interaction with oocytes and cumulus cells



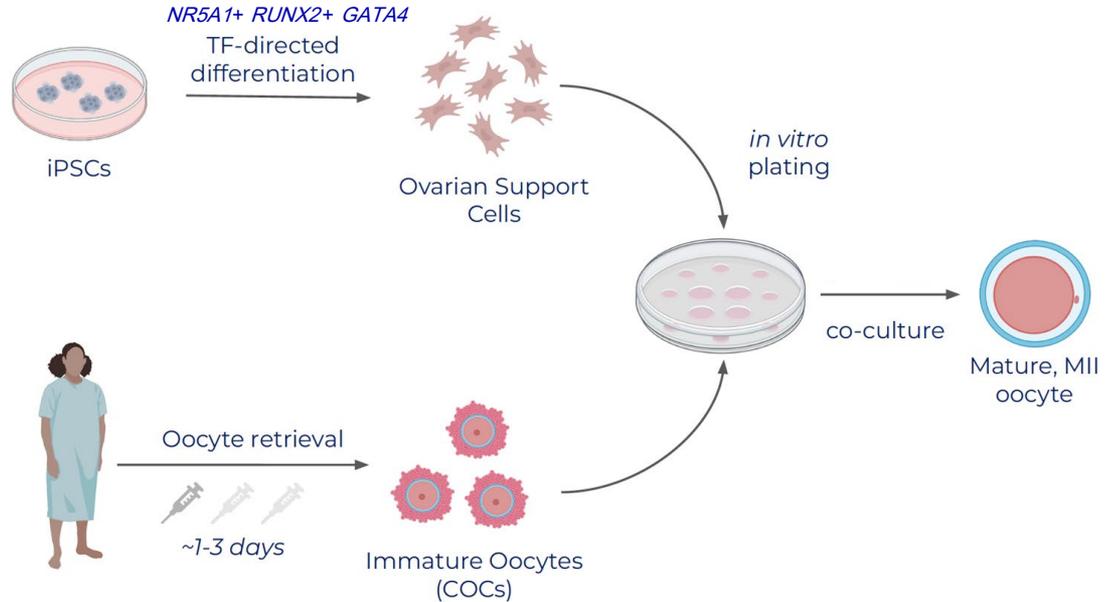
Gameto's **Fertilo** program uses proprietary OSCs to rapidly mature oocytes with minimal hormonal stimulations

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Let's see how this works!

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- They are steroidogenic, producing aromatase in response to FSH stimulation
- Produce the necessary growth factors needed for robust paracrine interaction with oocytes and cumulus cells



OSGIVM demonstrates improved oocyte maturation compared to commercially available IVM systems



Human Reproduction, 2023, 00(0), 1–14
<https://doi.org/10.1093/humrep/dead205>
Original Article

Infertility

Human-induced pluripotent stem cell-derived ovarian support cell co-culture improves oocyte maturation in vitro after abbreviated gonadotropin stimulation

Sabrina Piechota^{1,†}, Maria Marchante^{1,†}, Alexa Giovannini^{1,†}, Bruna Paulsen^{1,†}, Kathryn S. Potts³, Graham Rockwell¹, Caroline Aschenberger¹, Alexander D. Noblett⁴, Alexandra B. Figueroa¹, Marta Sanchez², Ferran Barrachina¹, Klaus Wiemer⁵, Luis Guzman¹, Pedro Belchín⁶, Merrick Pierson Smeola^{7,†}, Patrick K.J. Fortuna^{8,†}, Pranam Chatterjee^{9,†}, Nam D. Tran¹, Dawn A. Keik¹⁰, Marcy Forti¹⁰, Shelby Marcinyshyn¹⁰, Trozalla Smith¹⁰, David H. McCulloh^{1,11,12,13,14}, Marta-Julia Fernandez-Gonzalez¹, Baruch Abittan¹⁰, Silvia Ortiz¹, Joshua U. Klein¹⁰, Peter Klatsky⁹, Daniel Ordóñez-Pérez², and Christian C. Kramme^{1,14}

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⁴Francor Clinic, Lima, Peru

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⁶Department of Genetics, Harvard Medical School, Boston, MA, USA

⁷Department of Biomedical Engineering, Duke University, Durham, NC, USA

⁸Department of Computer Science, Duke University, Durham, NC, USA

⁹Spring Fertility, New York, NY, USA

¹⁰Extend Fertility, New York, NY, USA

¹¹Biogenetics Corporation, Mountainside, NJ, USA

¹²Sperm and Embryo Bank of New York, New York, NY, USA

¹³Biogenetics Laboratory, Brooklyn, NY, USA

¹⁴ReproART, Georgian American Center for Reproductive Medicine, Tbilisi, GA, USA

*Correspondence address: Gameto Inc., 286 Madison Ave., 1901, New York, NY 10016, USA. Tel: +1-515-720-1125; E-mail: christian@gametogen.com @ <https://orcid.org/0000-0002-7518-8111>

[†]These authors contributed equally to this work.

OSGIVM demonstrates improved oocyte maturation compared to commercially available IVM systems



Human Reproduction, 2023, 00(0), 1–14
<https://doi.org/10.1093/humrep/dead205>
Original Article

Infertility

Human-induced pluripotent stem cell-derived ovarian support cell co-culture improves oocyte maturation in vitro after abbreviated gonadotropin stimulation

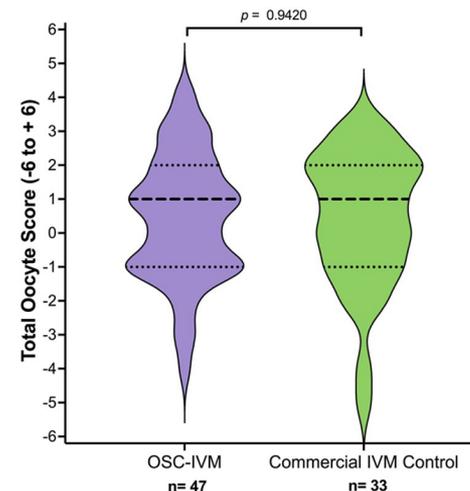
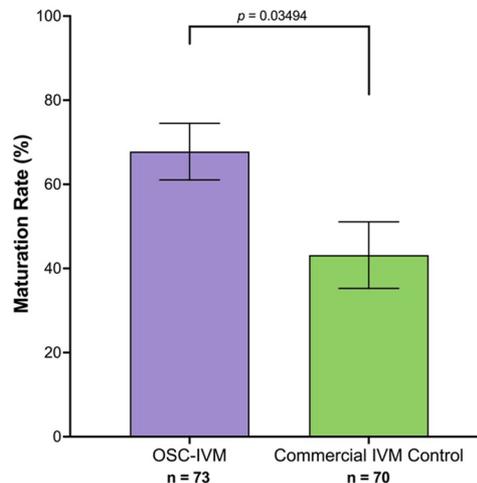
Sabrina Piechota^{1,†}, Maria Marchante^{1,†}, Alexa Giovannini^{1,†}, Bruna Paulsen^{1,†}, Kathryn S. Potts¹, Graham Rockwell¹, Caroline Aschenberger¹, Alexander D. Noblett¹, Alexandra B. Figueroa¹, Marta Sanchez², Ferran Bazzrachina³, Klaus Wiemer⁴, Luis Guzman⁵, Pedro Belchín⁶, Merrick Pierson Smeila^{7,8}, Patrick K.J. Fortuna⁹, Pranam Chatterjee⁹, Nam D. Tran¹⁰, Dawn A. Keik¹⁰, Marcy Forti¹⁰, Shelby Marcynishyn¹⁰, Trozalla Smith¹⁰, David H. McCulloh^{1,11,12,13,14}, Marta-Julia Fernandez-Gonzalez¹, Baruch Abittan¹⁰, Silvia Ortiz¹, Joshua U. Klein¹⁰, Peter Klatsky⁹, Daniel Ordóñez-Pérez², and Christian C. Kramme^{1,†*}

- ¹Gameto Inc., New York, NY, USA
- ²Ruber Juan Bravo University Hospital, Eugén Group, Madrid, Spain
- ³KEW Technology, Seattle, WA, USA
- ⁴Francor Clinic, Lima, Peru
- ⁵Wyss Institute, Harvard Medical School, Boston, MA, USA
- ⁶Department of Genetics, Harvard Medical School, Boston, MA, USA
- ⁷Department of Biomedical Engineering, Duke University, Durham, NC, USA
- ⁸Department of Computer Science, Duke University, Durham, NC, USA
- ⁹Spring Fertility, New York, NY, USA
- ¹⁰Extend Fertility, New York, NY, USA
- ¹¹Biogenetics Corporation, Mountainside, NJ, USA
- ¹²Sperma and Embryo Bank of New York, New York, NY, USA
- ¹³Sperma and Embryo Bank of New York, New York, NY, USA
- ¹⁴Biogenetics Laboratory, Brooklyn, NY, USA

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[†]These authors contributed equally to this work.

After 28 h of IVM



Commercial IVM = Medicult (Cooper Surgical)

OSGIVM-assisted oocytes are developmentally competent for healthy embryo formation



Human Reproduction, 2023, 00(0), 1–14
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 Original Article

Infertility

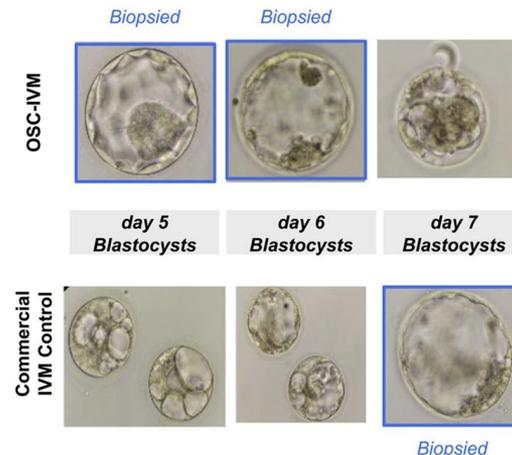
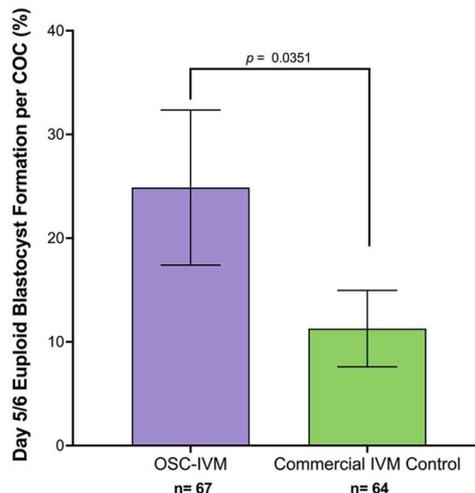
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The OSGIVM platform is a novel tool to rescue immature oocytes from COS cycles

GAMETE BIOLOGY



Rescue in vitro maturation using ovarian support cells of human oocytes from conventional stimulation cycles yields oocytes with improved nuclear maturation and transcriptomic resemblance to in vivo matured oocytes

Bruna Paulsen¹ · Sabrina Piechota¹ · Ferran Barrachina¹ · Alexa Giovannini¹ · Simone Kats¹ · Kathryn S. Potts¹ · Graham Rockwell¹ · Maria Marchante¹ · Samantha L. Estevez² · Alexander D. Noblett¹ · Alexandra B. Figueroa¹ · Caroline Aschenberger¹ · Dawn A. Kelk³ · Marcy Forti³ · Shelby Marcinyshyn³ · Klaus Wiemer⁴ · Marta Sanchez⁵ · Pedro Belchin⁵ · Joseph A. Lee⁶ · Erkan Buyuk^{2,6} · Rick E. Slifkin⁶ · Merrick Pierson Smela^{7,8} · Patrick R. J. Fortuna^{7,8} · Pranam Chatterjee^{9,10} · David H. McCulloh¹ · Alan B. Copperman^{2,6} · Daniel Ordonez-Perez⁵ · Joshua U. Klein⁹ · Christian C. Kramme¹ 

Received: 10 August 2023 / Accepted: 9 May 2024
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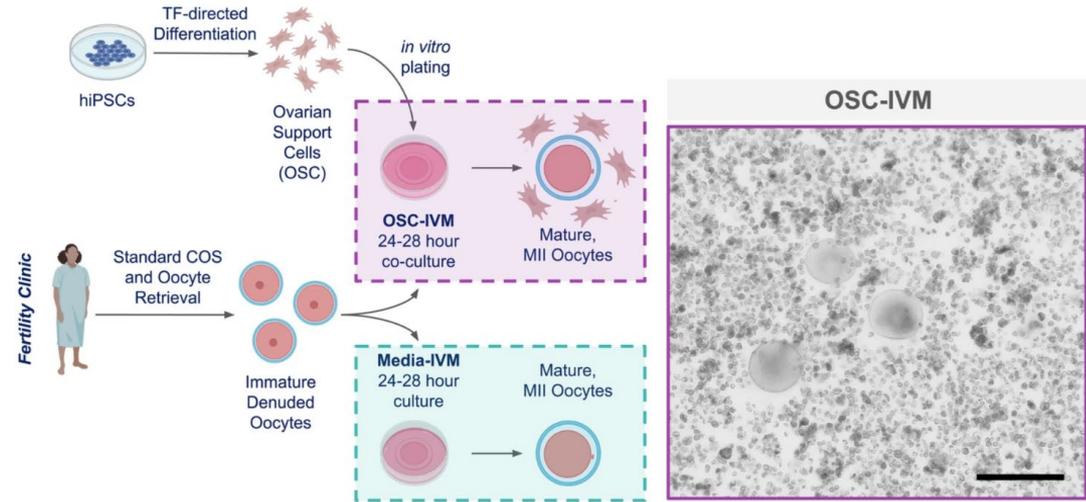
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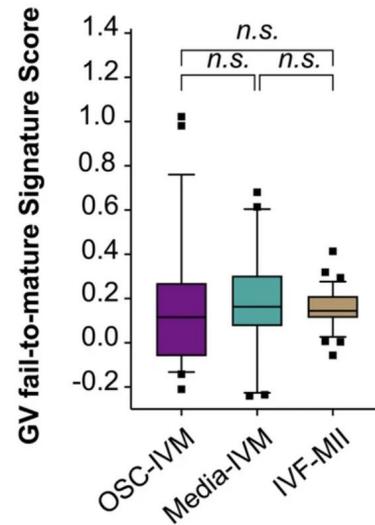
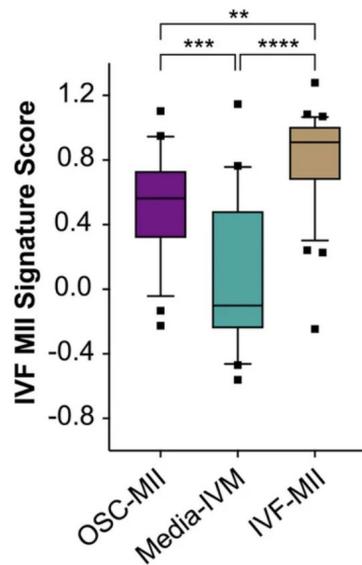
Oocytes co-cultured with OSCs (vs. IVM Media) are transcriptionally closer to IVF MII oocytes



Rescue in vitro maturation using ovarian support cells of human oocytes from conventional stimulation cycles yields oocytes with improved nuclear maturation and transcriptomic resemblance to in vivo matured oocytes

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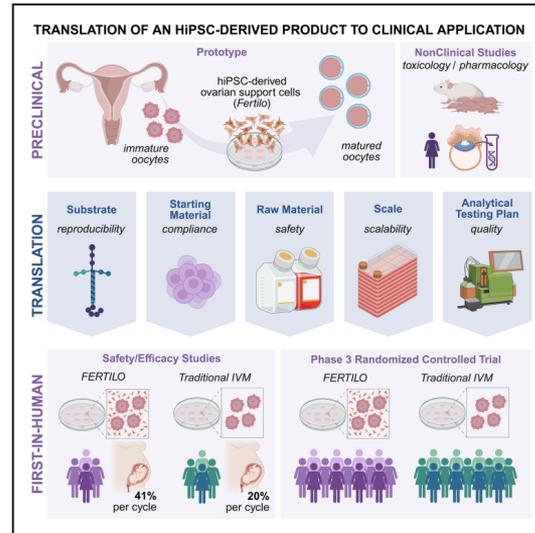


Our complete **Fertilo** results are now out!

Cell Stem Cell Clinical and Translational Report

Development of human induced pluripotent stem cell-derived ovarian support cells as a clinical-grade product for *in vitro* fertilization

Graphical abstract



Authors

Bruna Paulsen, Ferran Barrachina, Sabrina Piechota, ..., David F. Albertini, Michel De Vos, Christian C. Kramme

Correspondence

christian@gametogen.com

In brief

Paulsen et al. present the process development and clinical application of an hiPSC-derived OSC product, *Fertilo*. They describe the raw material upgrades, process consistency and reproducibility, and analytical assessment required for the generation of a clinically suitable product, as well as favorable outcomes from the first-in-human application of *Fertilo*.

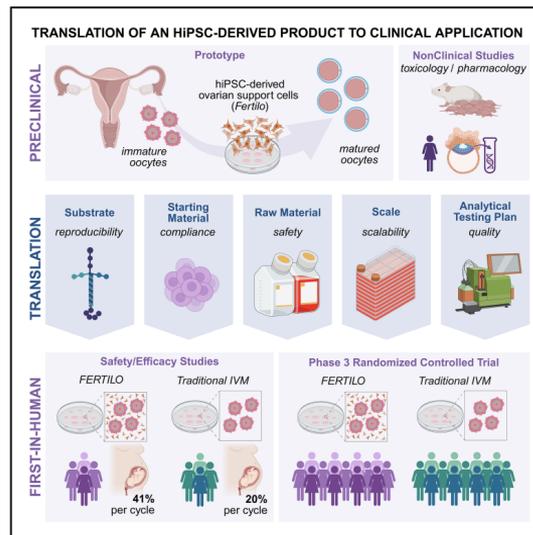
Let me walk you through the key points!

Cell Stem Cell

Clinical and Translational Report

Development of human induced pluripotent stem cell-derived ovarian support cells as a clinical-grade product for *in vitro* fertilization

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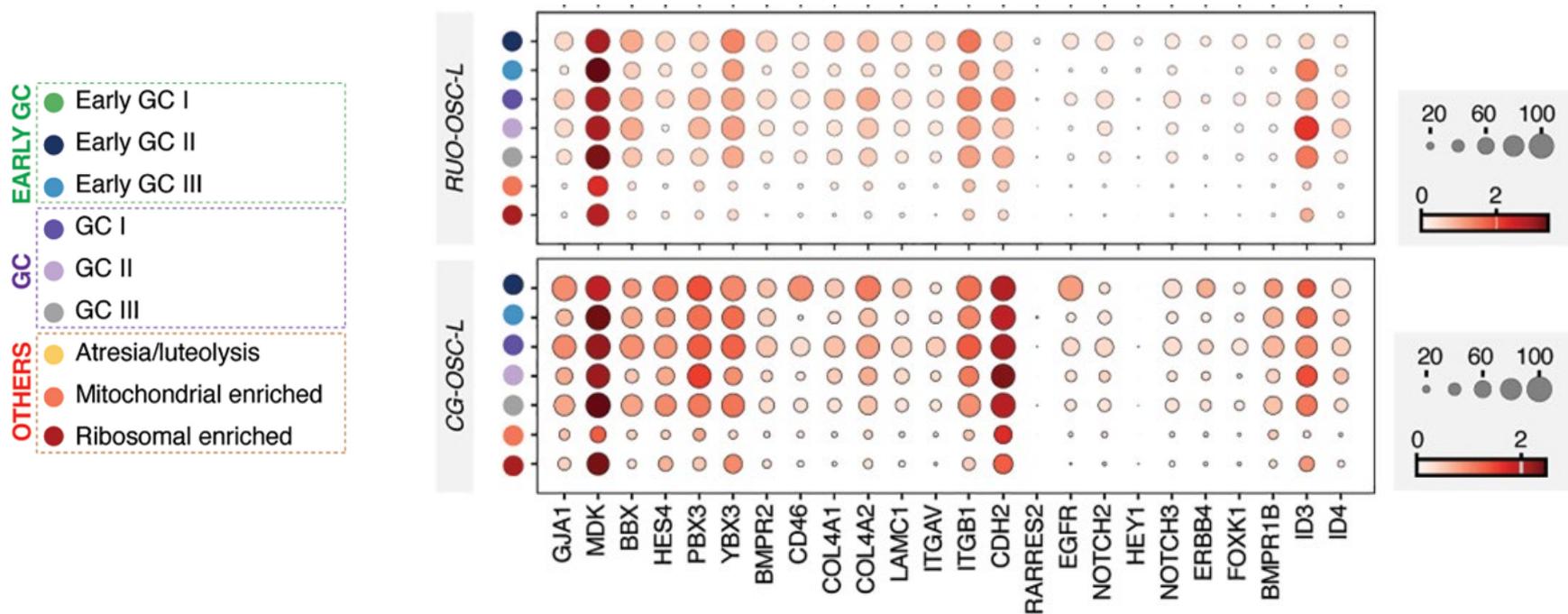
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Clinical-grade (CG) hiPSC → reproducible OSCs



A master cell bank (MCB) with GMP-compliant manufacturing and safety

Table 2. Safety testing panel

Test	Specification	CG-hiPSC MCB	CG-OSC production lots	CG-OSC product	CG-OSC product 6 months storage	CG-OSC product 12 months storage
Genomic integrity	no karyotypic abnormalities detected	pass	NT	pass	pass	pass
Endotoxin	≤0.1 EU/mL	pass	pass	pass	NT	NT
Mycoplasma	not detected	pass	pass	pass	NT	NT
Sterility	no growth	pass	pass	pass	pass	pass
Adventitious agents (human test panel)	not detected	Pass	pass	pass	NT	NT
Adventitious agents (<i>in vivo</i>)	negative	pass	NT	NT	NT	NT
Adventitious agents (<i>in vitro</i>)	negative	pass	NT	NT	NT	NT
Adventitious agents (species- specific)	negative	pass	NT	NT	NT	NT
Retroviral contamination	negative	pass	NT	NT	NT	NT
Residual hiPSC – RT-qPCR	OCT4 ≤ 0.1%; NANOG ≤ 0.1%	NT	NT	pass	NT	NT
Residual doxycycline	≤0.06 µg/mL	NT	NT	pass	NT	NT
Mouse embryo assay –embryotoxicity	blastocyst formation ≥ 80%	NT	NT	pass	pass	pass

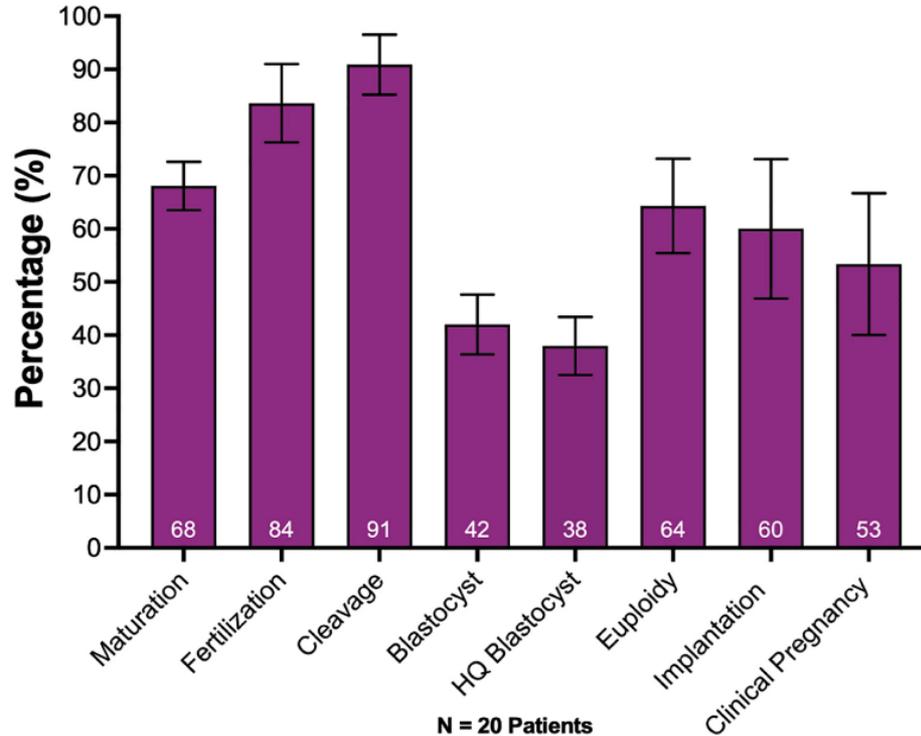
NT, not tested.



So we headed to clinical trials



Phase 1: High maturation, euploidy, and live birth rates



Our first **Fertilo** baby was born in Lima, Peru! PE



3255 g, 49.5 cm, 9/9 APGAR!



Welcome Mia! 🧡



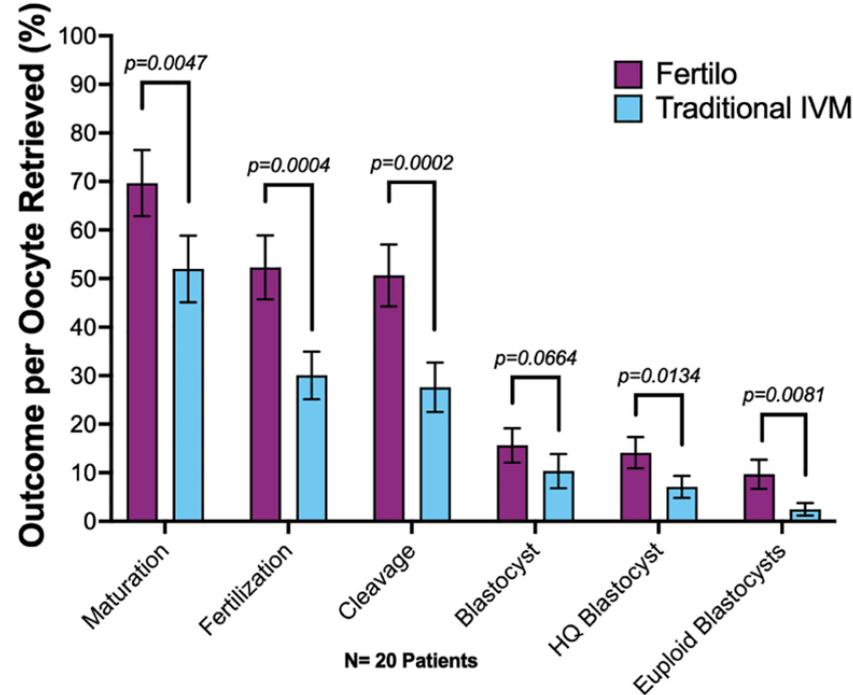
Welcome Mia! 🤗

We now have 5 additional births from this study! 🧒



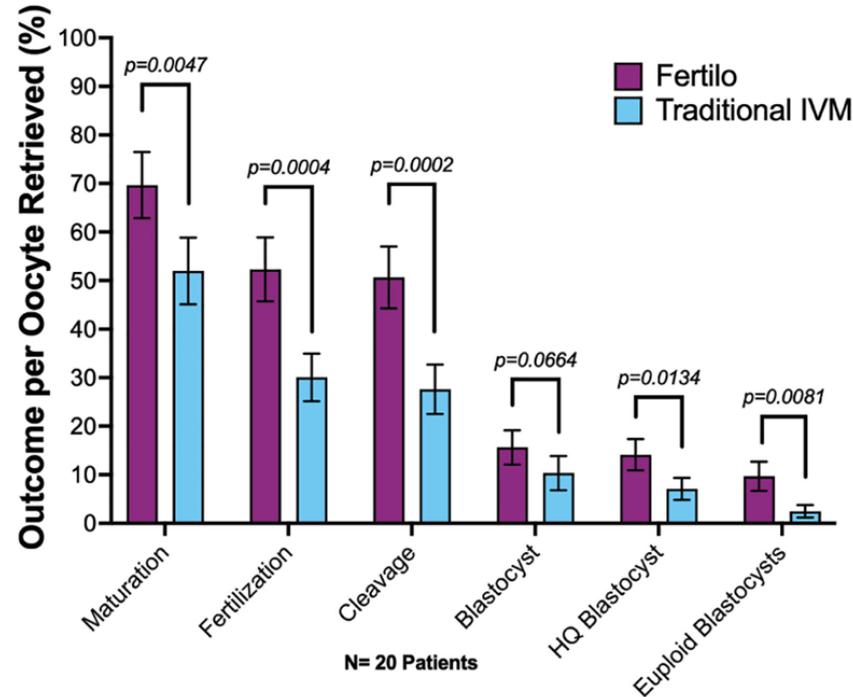


Phase 2: Fertilo improves outcomes vs. standard IVM



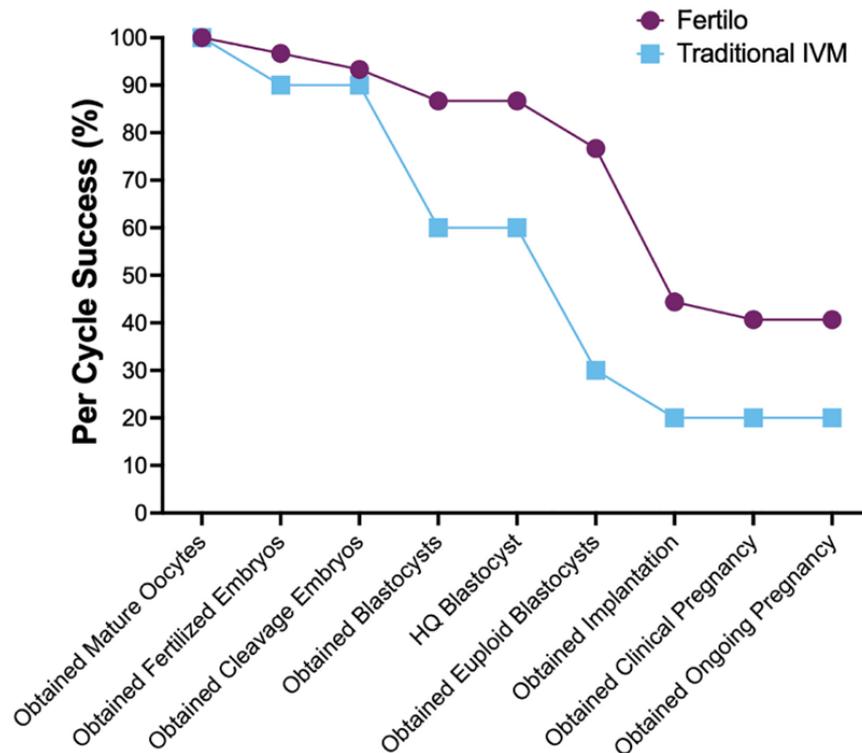


Phase 2: Two healthy, live births in this group!

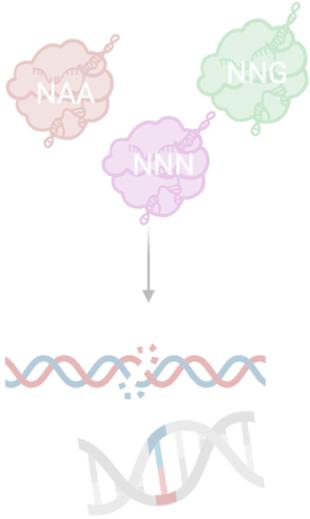




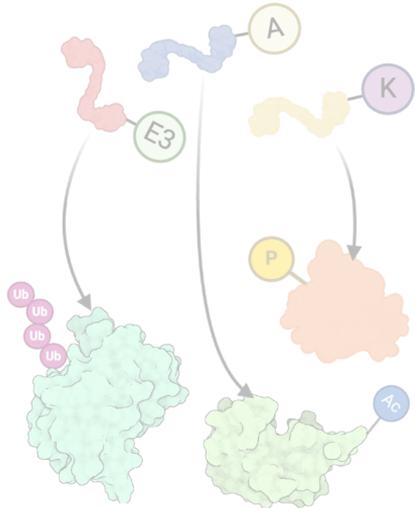
In total, **Fertilo** doubles ongoing pregnancy per cycle



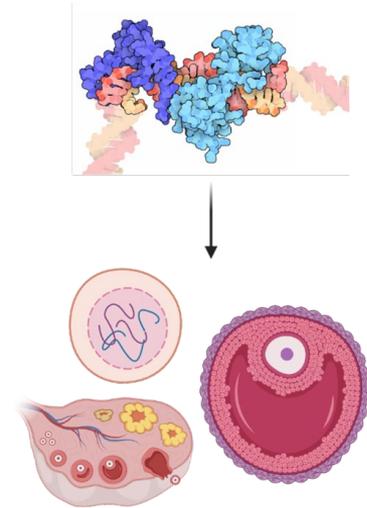
So this is the “near” future



DNA

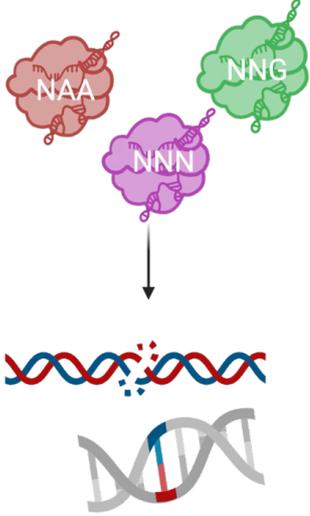


Proteins

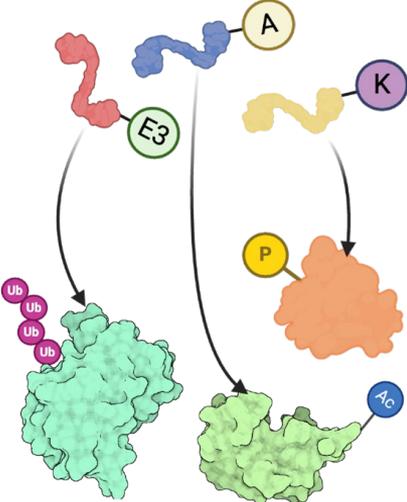


Cells/Tissues

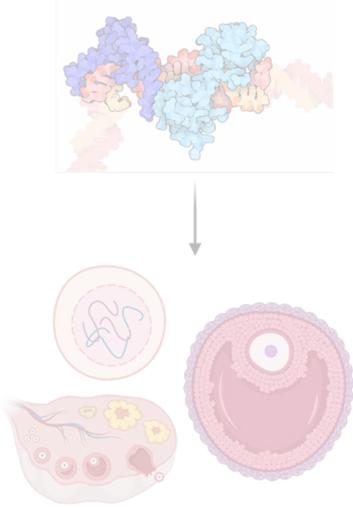
Let's look a little farther ahead!



DNA

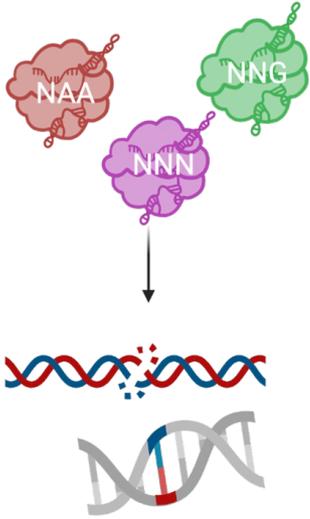


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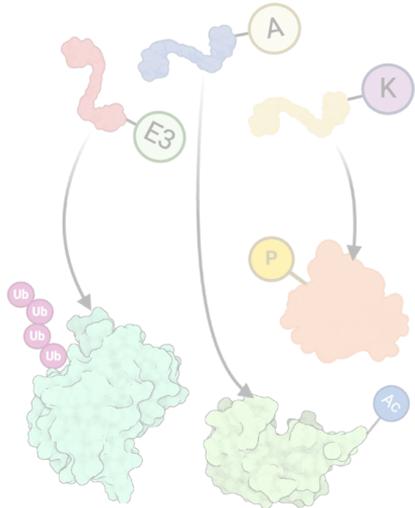


Cells/Tissues

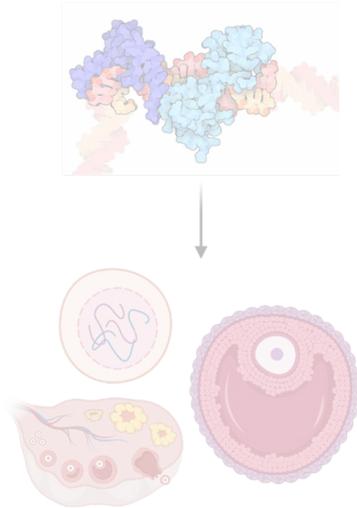
Proteins that can edit any DNA sequence



DNA

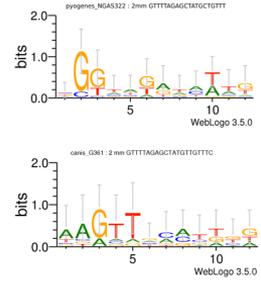
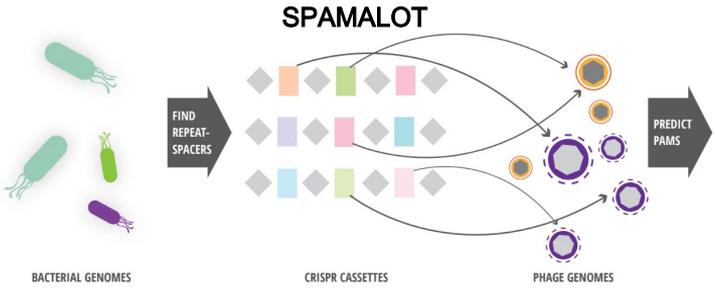
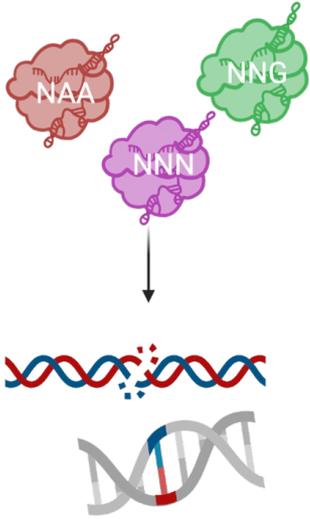


Proteins



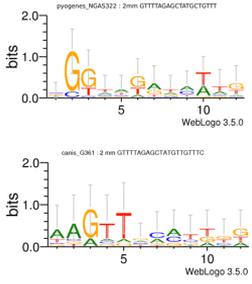
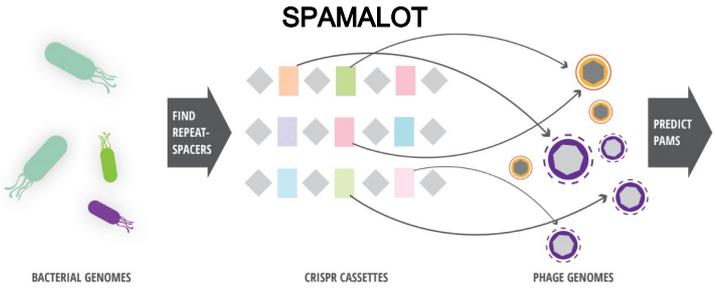
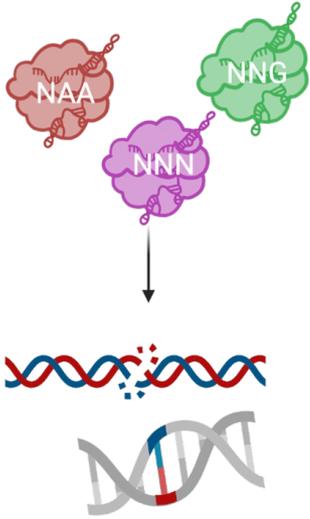
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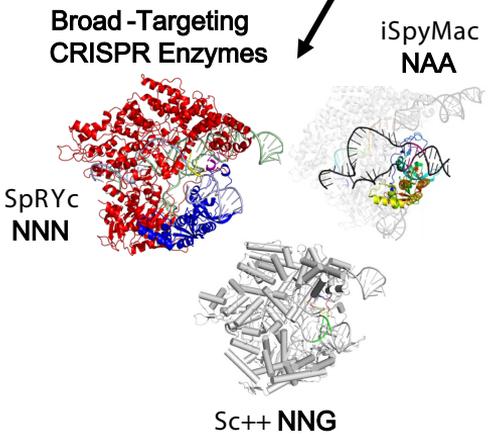


$$P_{\text{PAM}} = \bigcup_{i,k} \{p[\text{PAM region}] \mid p \in G_p, \text{Bowtie}(s_{i,k}, p) \leq 2 \text{ mismatches}\}$$

Proteins that can edit any DNA sequence



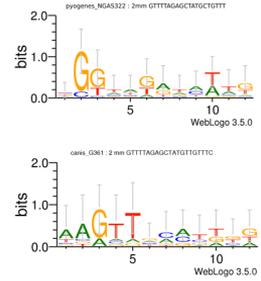
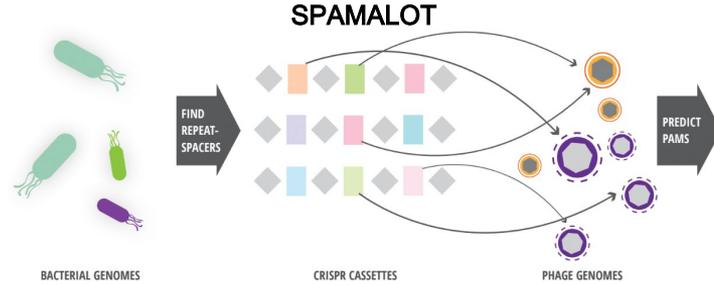
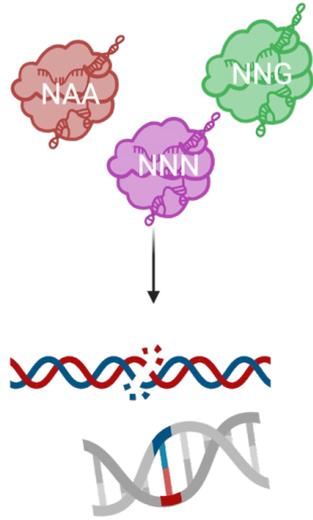
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Chatterjee, et al. *Nature Communications*(2020)
 Chatterjee, et al. *Nature Biotechnology* (2020)
 Zhao, Koseki,...,Chatterjee. *Nature Communications*(2023)



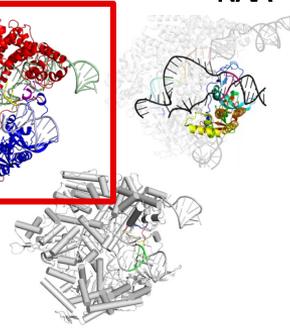
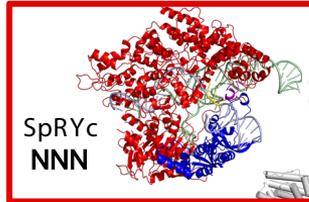
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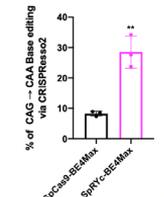
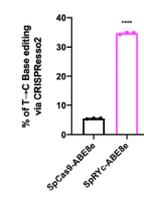
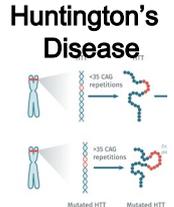
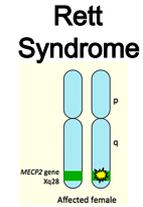
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Broad-Targeting
CRISPR Enzymes

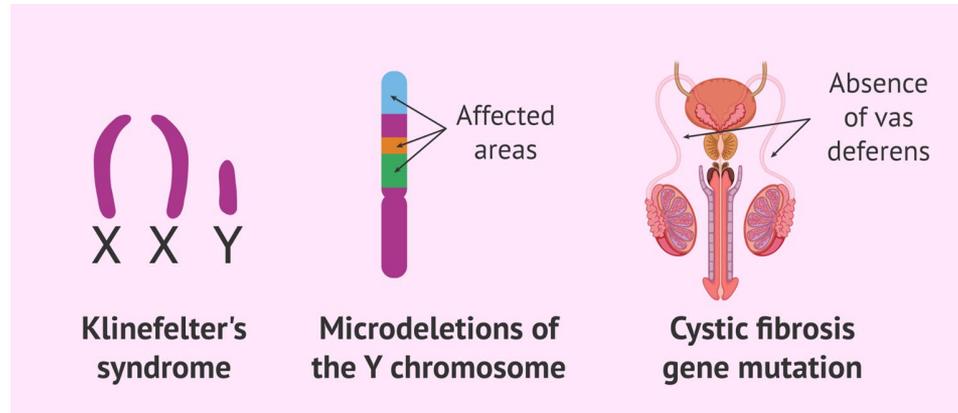
iSpyMac
NAA



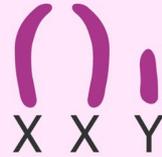
Sc++ NNG



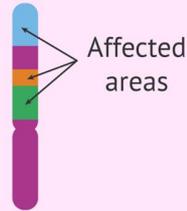
We are using SpRYc to address fertility-related disorders



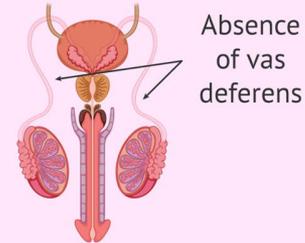
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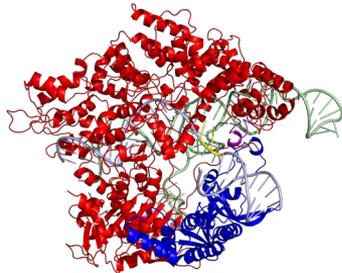
Klinefelter's syndrome



Microdeletions of the Y chromosome



Cystic fibrosis gene mutation

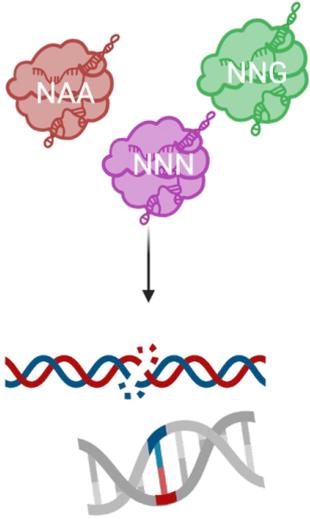


SpRYc-mediated chromosomal elimination

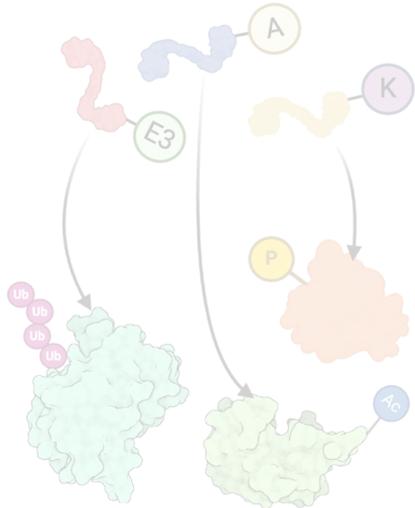
SpRYc-mediated restoration of microdeletions

SpRYc-mediated correction of mutations

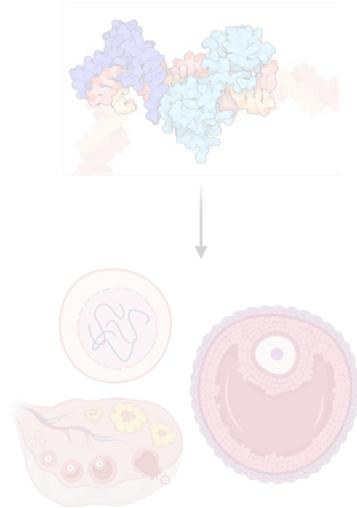
Proteins that can edit any DNA sequence



DNA

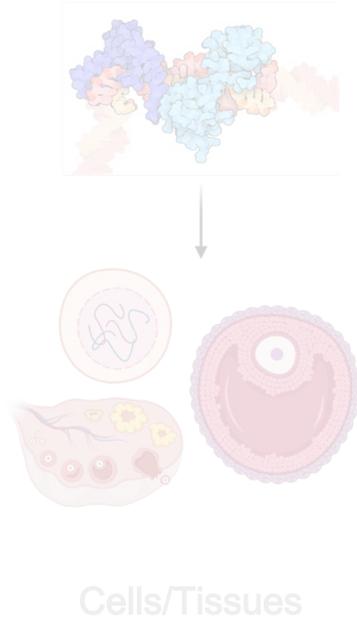
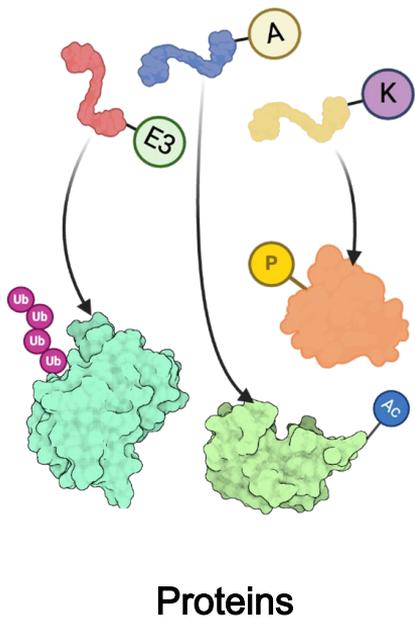
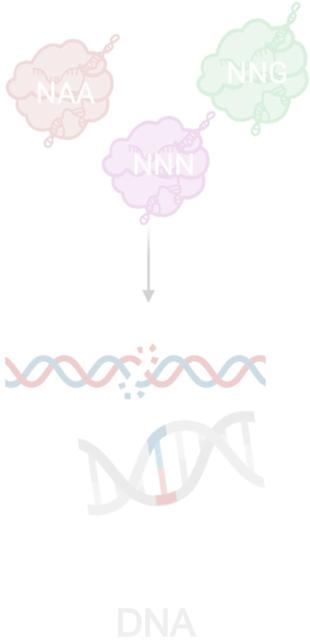


Proteins



Cells/Tissues

Proteins that can bind and modify any other protein

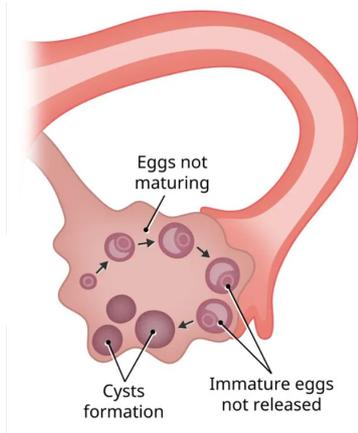


Fertility -related disorders are driven by dysregulated proteins



Fertility -related disorders are driven by dysregulated proteins

PCOS

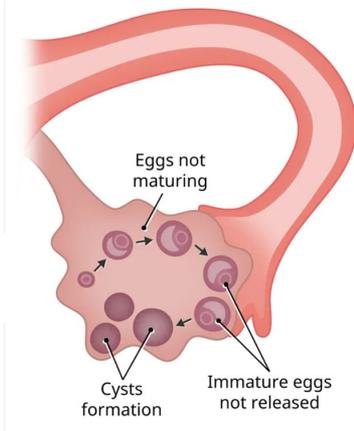


AR, StAR,
CYP17A1,
IRS-1



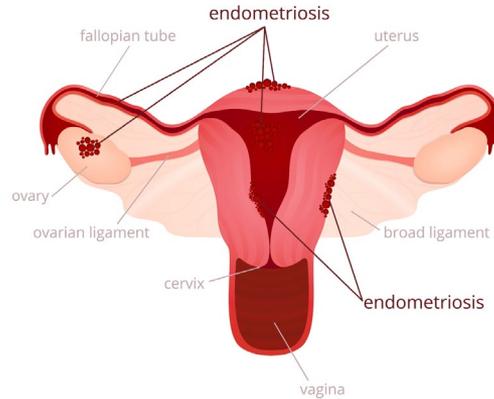
Fertility -related disorders are driven by dysregulated proteins

PCOS



AR, StAR,
CYP17A1,
IRS-1

Endometriosis

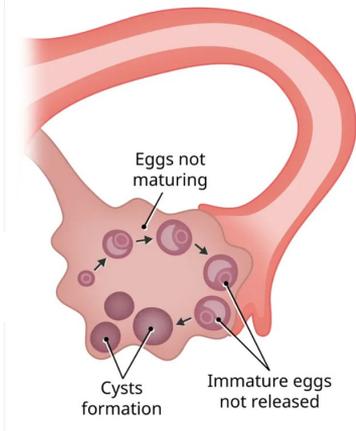


ER α , COX-2,
CYP19A1,
MMP-9



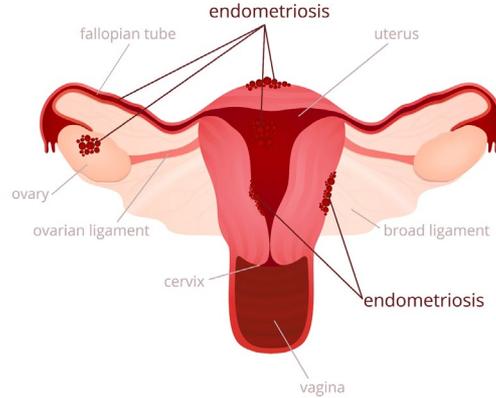
Fertility -related disorders are driven by dysregulated proteins

PCOS



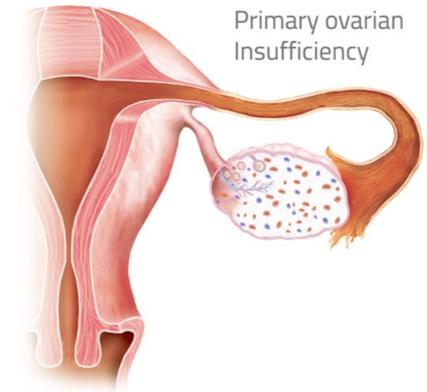
AR, StAR,
CYP17A1,
IRS-1

Endometriosis



ER α , COX-2,
CYP19A1,
MMP-9

POI



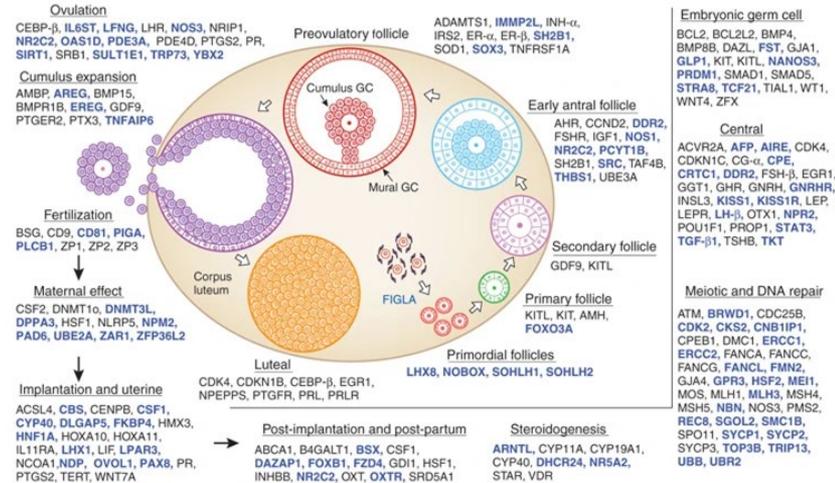
FOXO3a, p53,
BMP15,
GDF9

For proteins, unlike DNA, it's not so simple.

nature medicine

The biology of infertility: research advances and clinical challenges

Martin M. Matzuk & Dolores J. Lamb

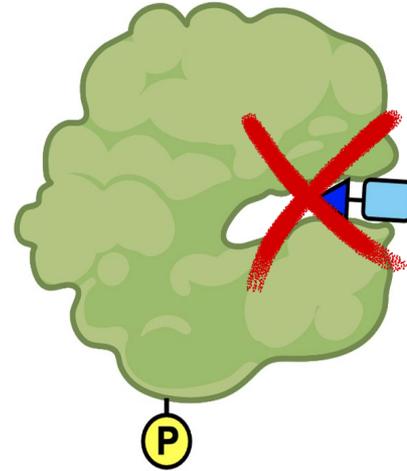
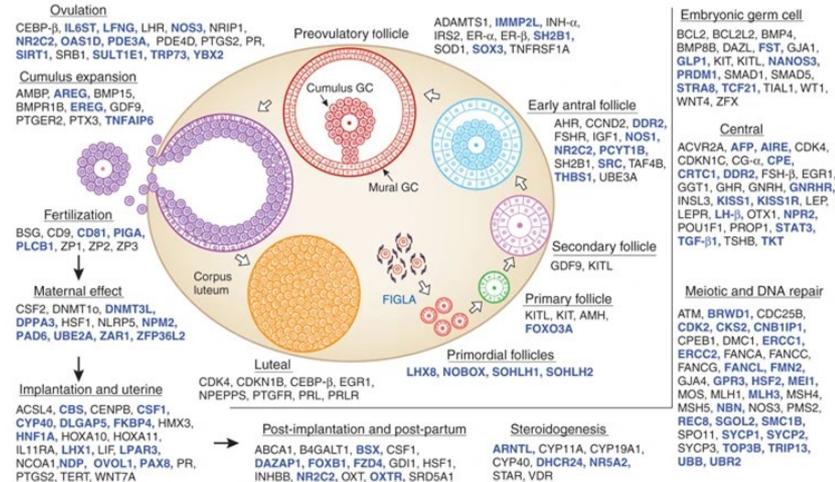


Many infertility -driving proteins are considered “undruggable”

nature medicine

The biology of infertility: research advances and clinical challenges

Martin M Matzuk & Dolores J Lamb

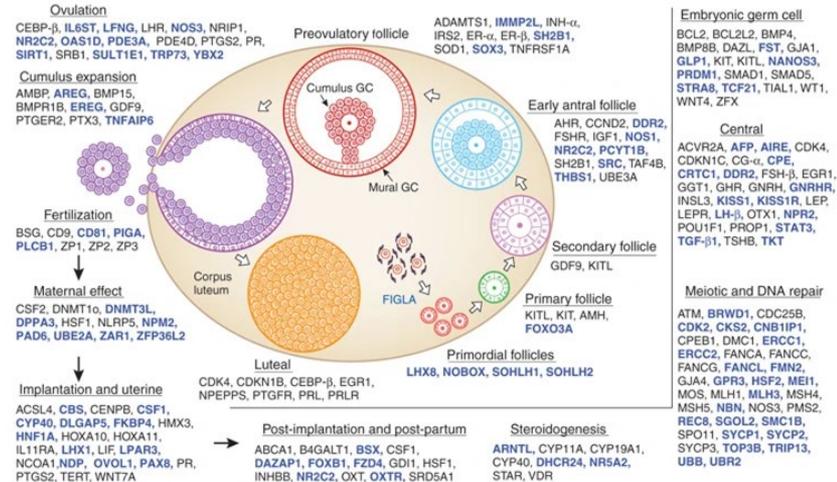


And are also conformationally disordered

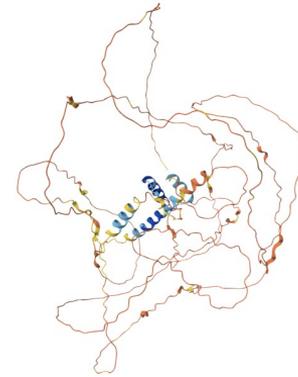
nature medicine

The biology of infertility: research advances and clinical challenges

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AlphaFold prediction: NOBOX

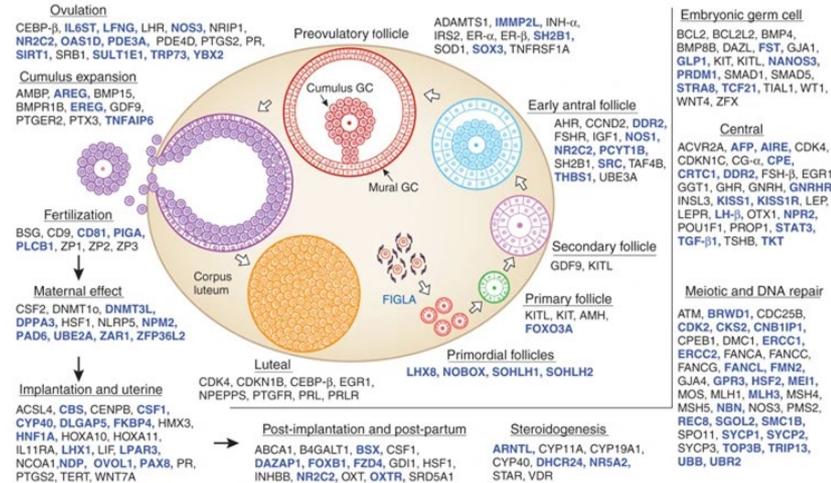


So you can't use small molecules or structure-based methods to hit these targets...

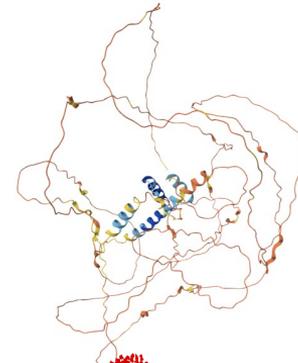
nature medicine

The biology of infertility: research advances and clinical challenges

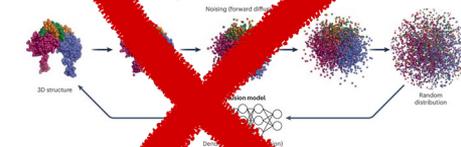
Martin M. Matzuk & Dolores J. Lamb



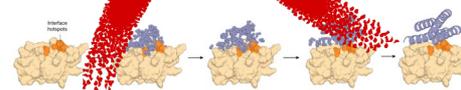
AlphaFold prediction: NOBOX



Accurate structure prediction of biomolecular interactions with AlphaFold 3



De novo design of protein structure and function with RFdiffusion

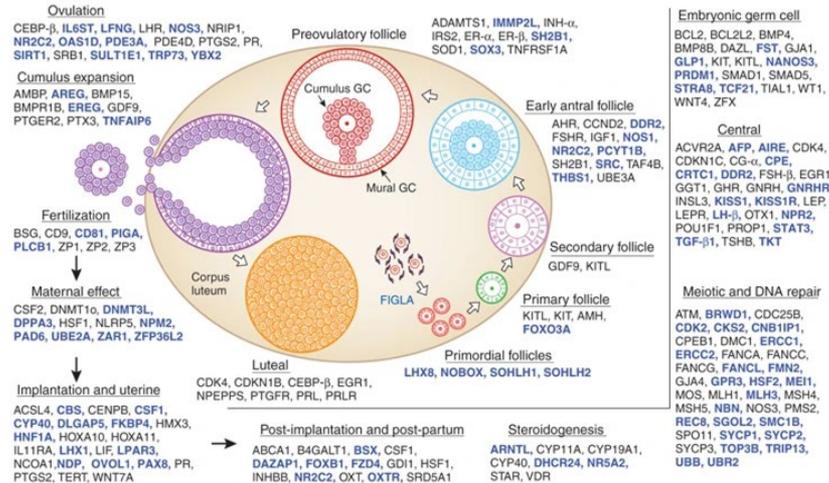


So you can't use small molecules or structure-based methods to hit these targets...

nature medicine

The biology of infertility: research advances and clinical challenges

Martin M Matzuk & Dolores J Lamb

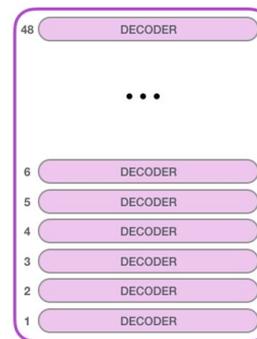


Can we design “guide” peptides to bind and degrade undruggable targets without needing their structure?

Our first instinct : language models

> 1 Trillion Sentences
~ 70 trillion tokens

Large Language Models (LLMs), such as GPT-3 and GPT-4, utilize a process called tokenization. Tokenization involves breaking down text into smaller units, known as tokens, which the model can process and understand. These tokens can range from individual characters to entire words or even larger chunks, depending on the model. For GPT-3 and GPT-4, a Byte Pair Encoding (BPE) tokenizer is used. BPE is a subword tokenization technique that allows the model to dynamically build a vocabulary during training, efficiently representing common words and word fragments. Although the core tokenization process remains similar across different versions of these models, the specific implementation can vary based on the model's architecture and training objectives.

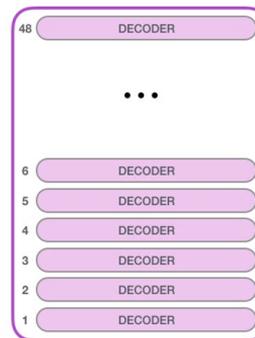


Our first instinct : language models

- ✓ Syntax
- ✓ Semantics
- ✓ Grammar

> 1 Trillion Sentences
~ 70 trillion tokens

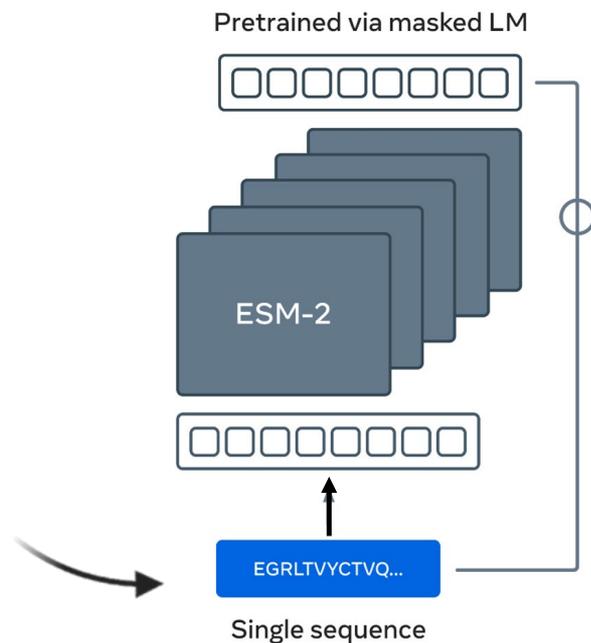
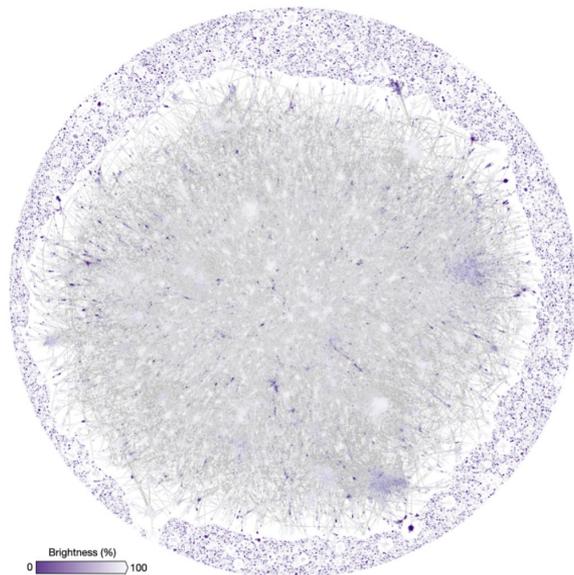
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Our first instinct : protein language models

- ✓ Syntax
- ✓ Semantics
- ✓ Grammar

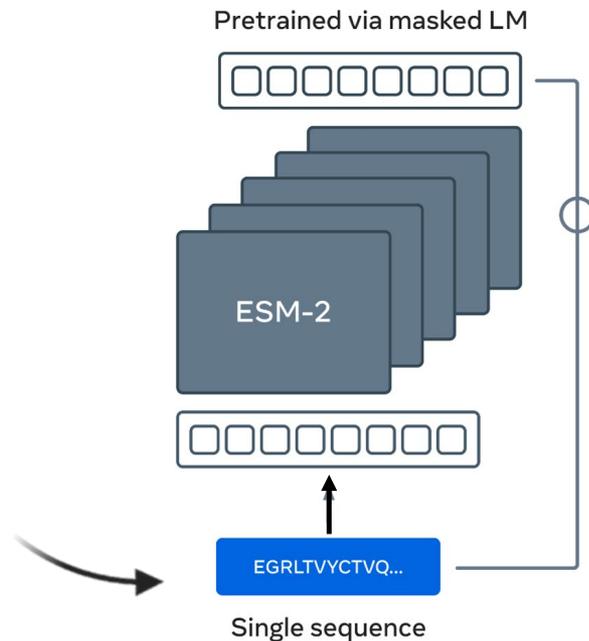
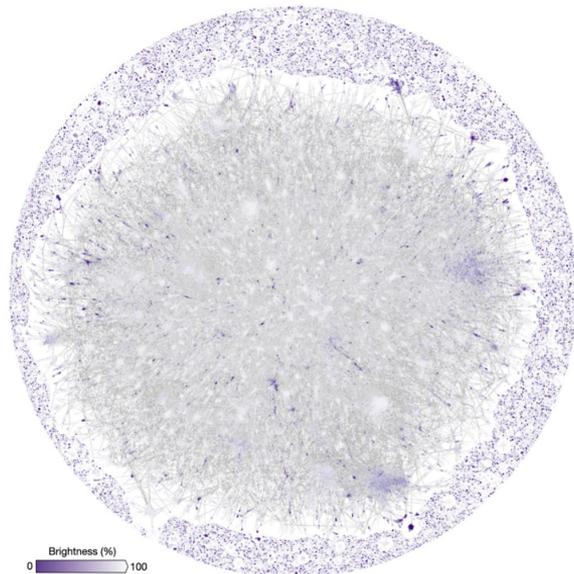
~100 Million Sequences



Our first instinct : protein language models

- ✓ ~~Syntax~~ Structure
- ✓ ~~Semantics~~ Function
- ✓ ~~Grammar~~ Interaction

~100 Million Sequences

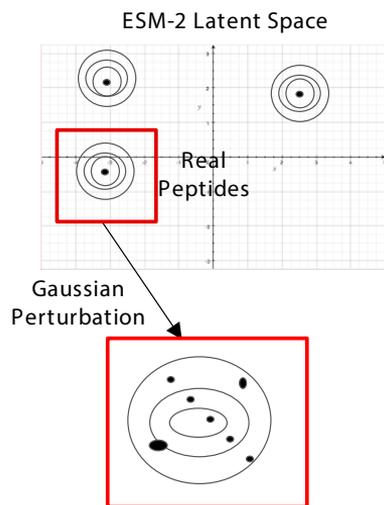


pLM-based generators for target-specific peptide design



pLM-based generators for target-specific peptide design

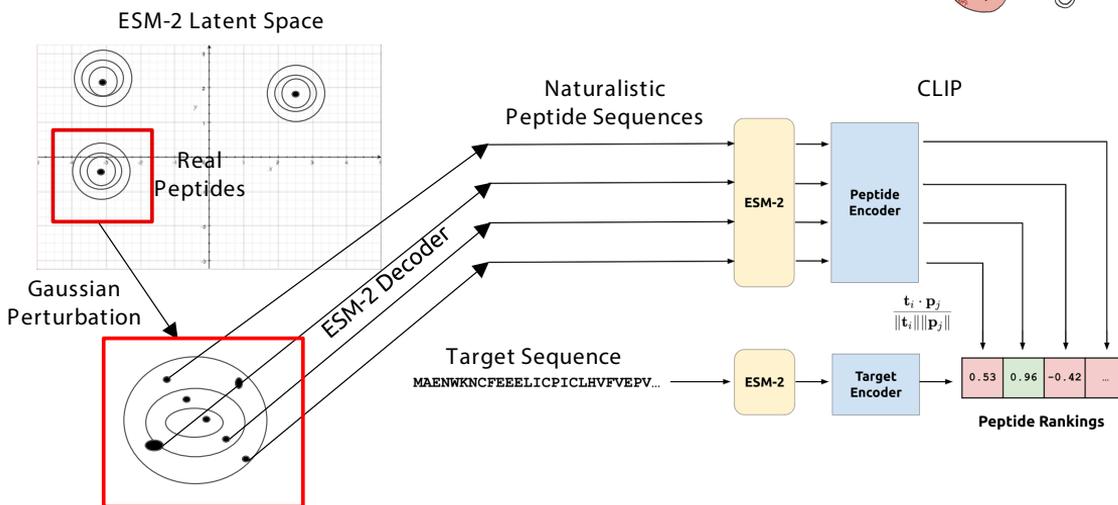
PepPrCLIP





pLM-based generators for target-specific peptide design

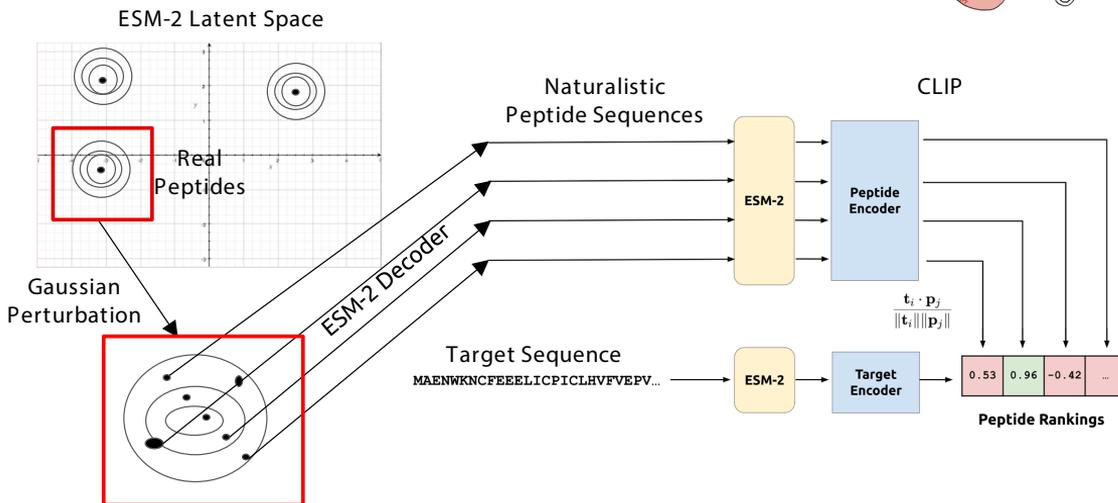
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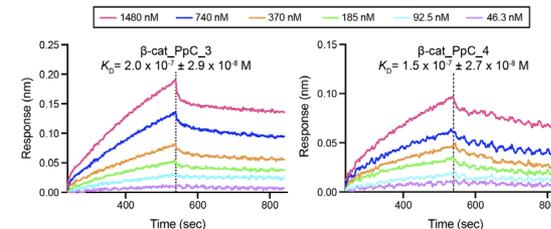


pLM-based generators for target-specific peptide design

PepPrCLIP



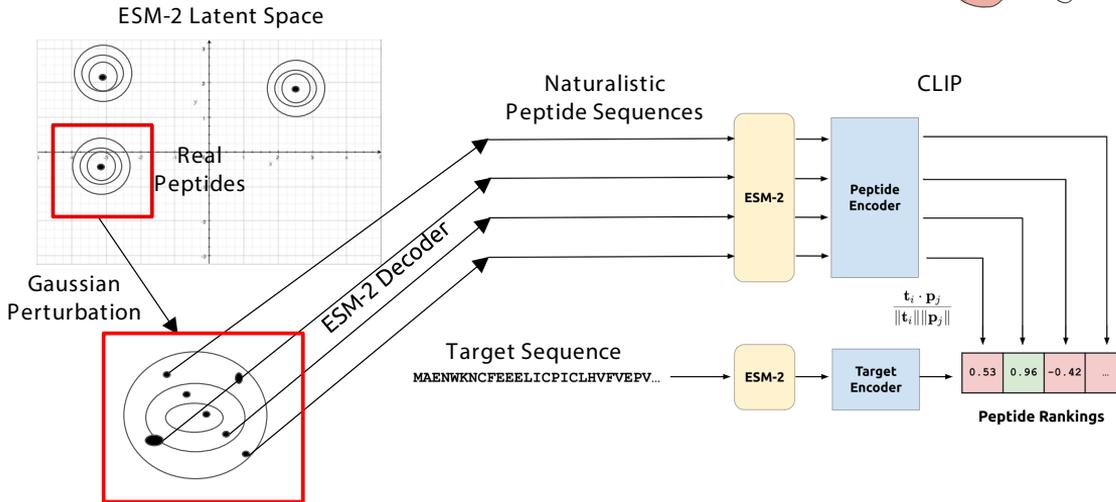
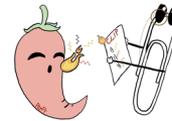
Nanomolar Binding



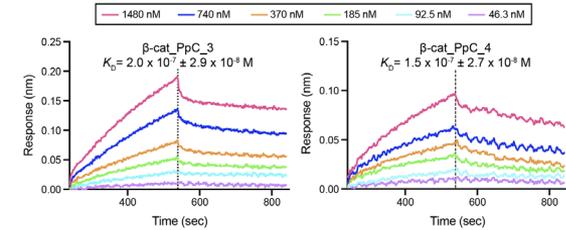


pLM-based generators for target-specific peptide design

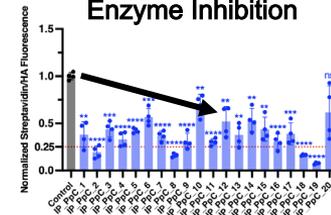
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Nanomolar Binding



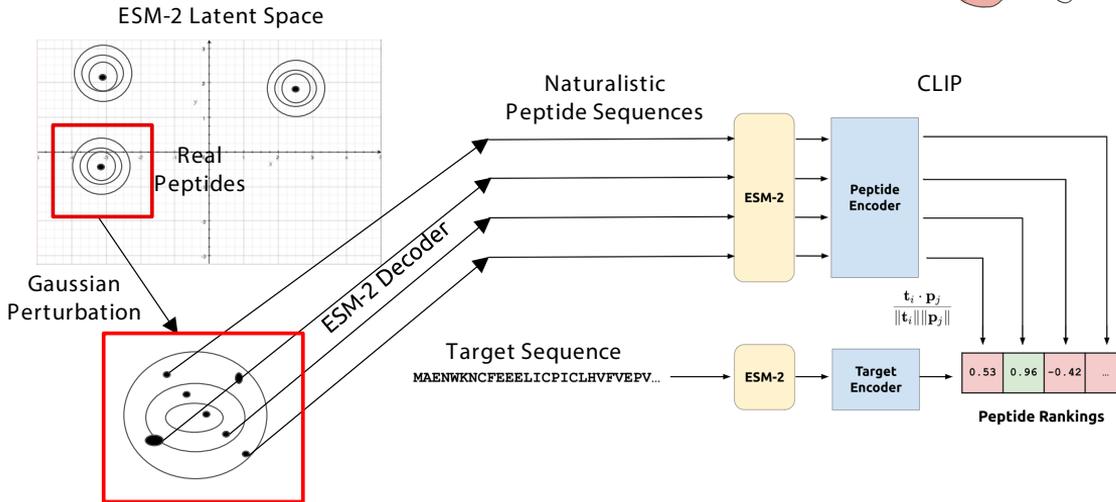
Enzyme Inhibition



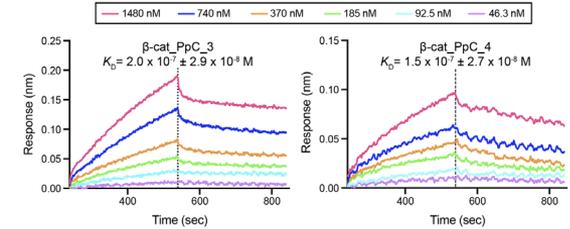


pLM-based generators for target-specific peptide design

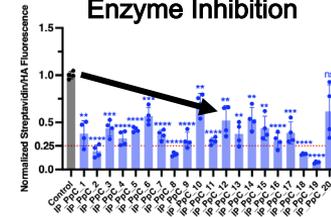
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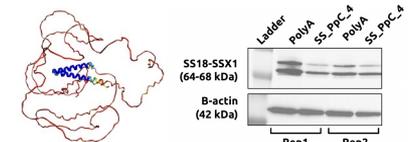
Nanomolar Binding



Enzyme Inhibition

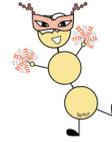


Disordered Protein Degradation

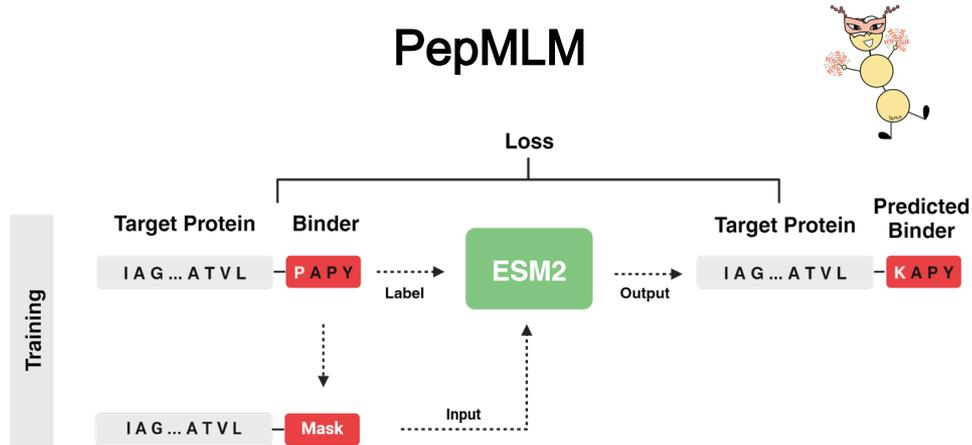




pLM-based generators for target-specific peptide design

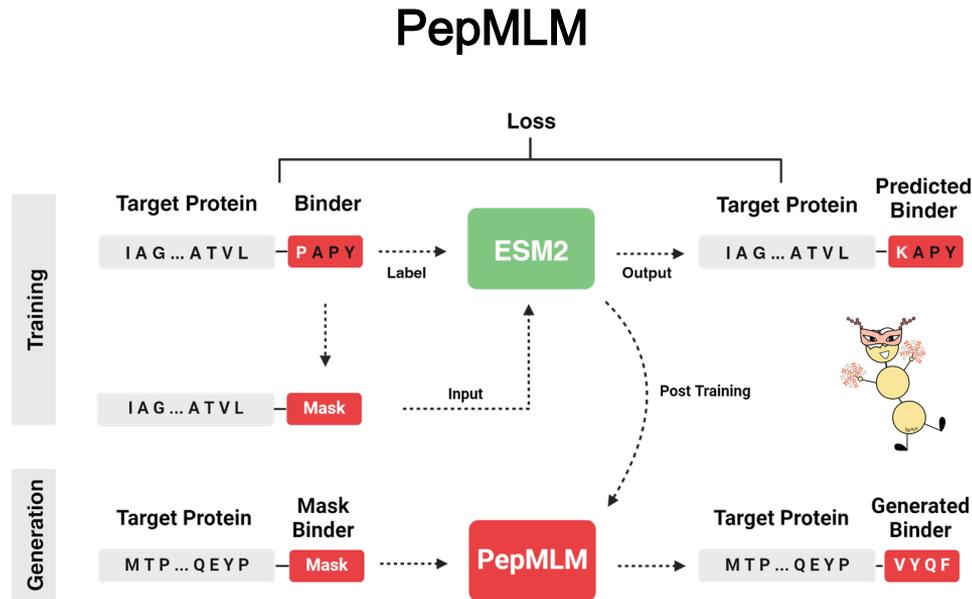


PepMLM



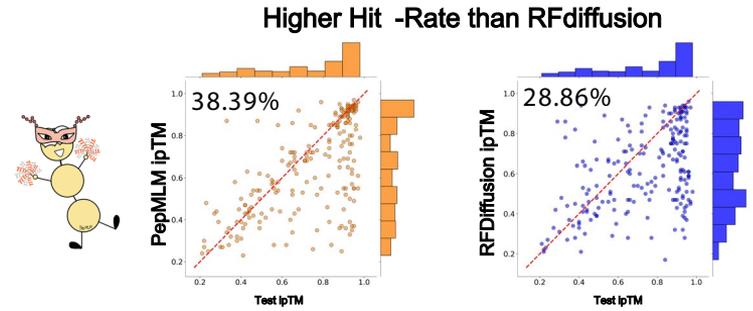
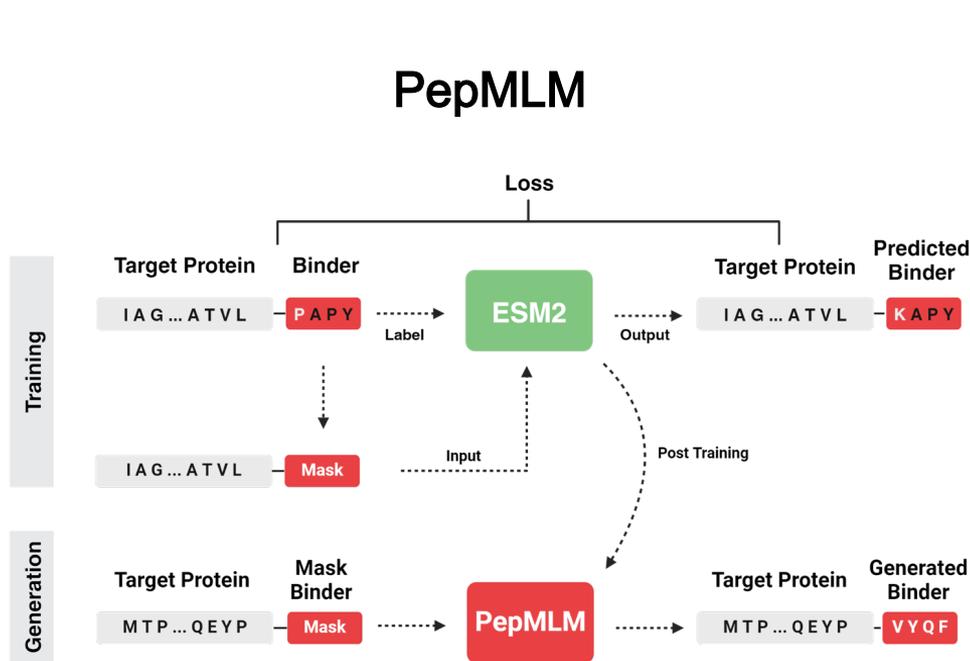


pLM-based generators for target-specific peptide design



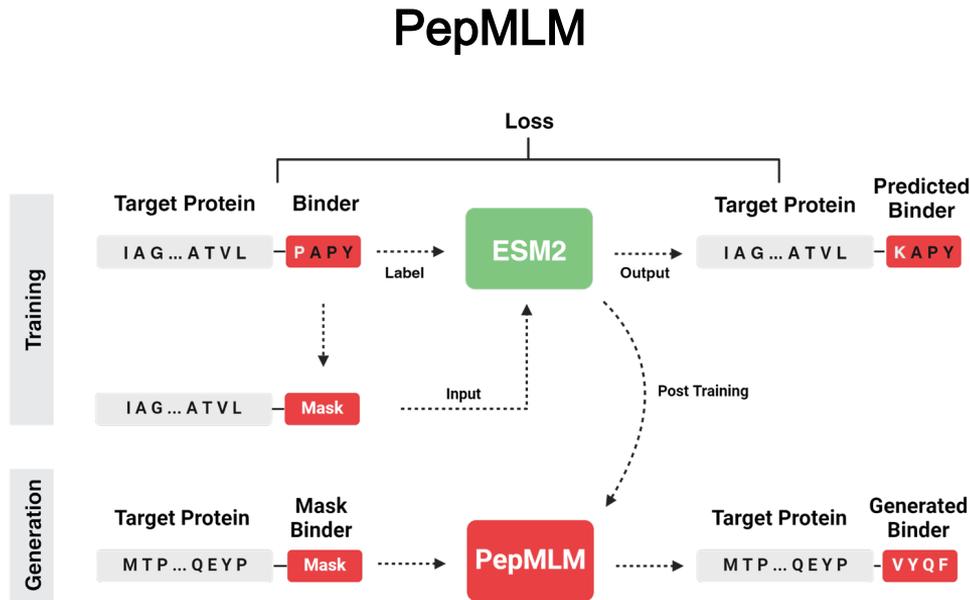


pLM-based generators for target-specific peptide design

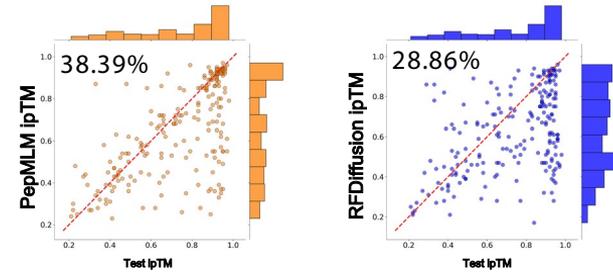




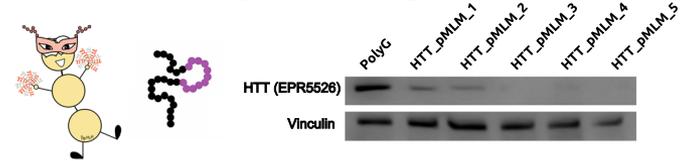
pLM-based generators for target-specific peptide design



Higher Hit Rate than RFdiffusion

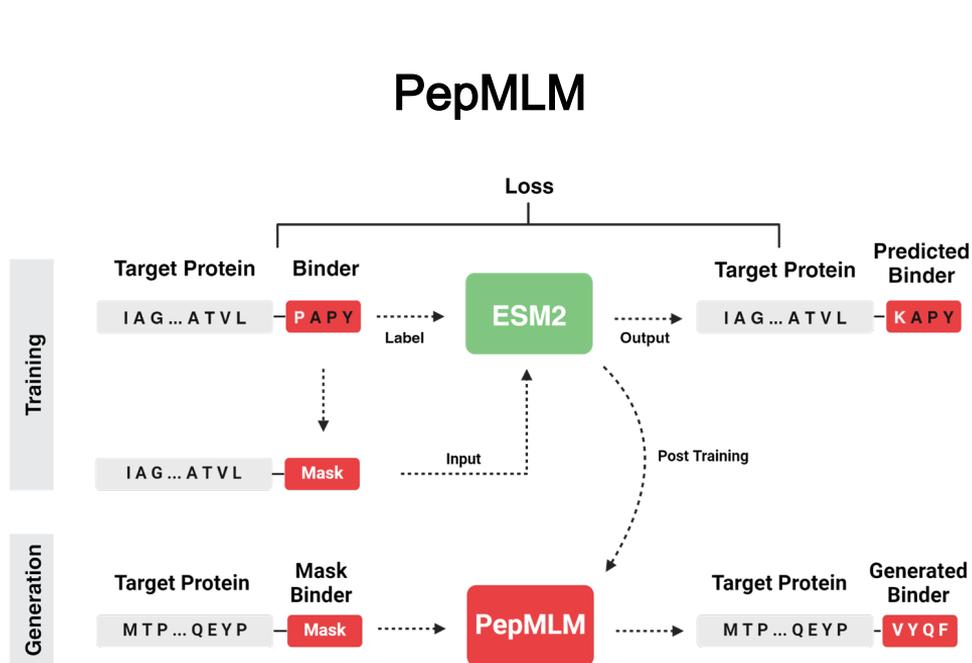


Degrades Huntingtin Protein

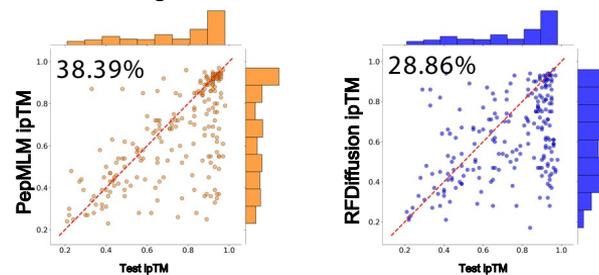




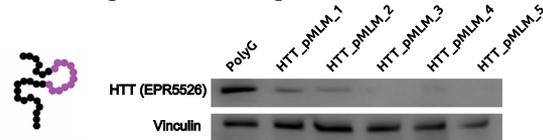
pLM-based generators for target-specific peptide design



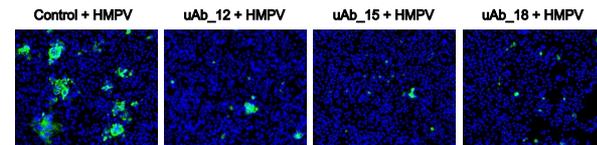
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Degrades Huntingtin Protein

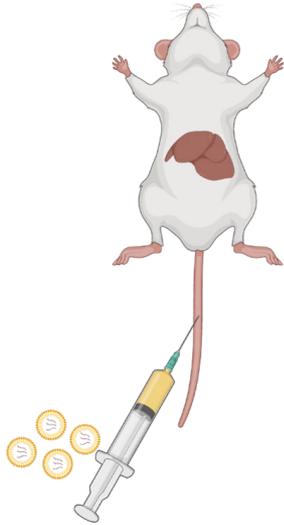


Inhibits Viral Infection



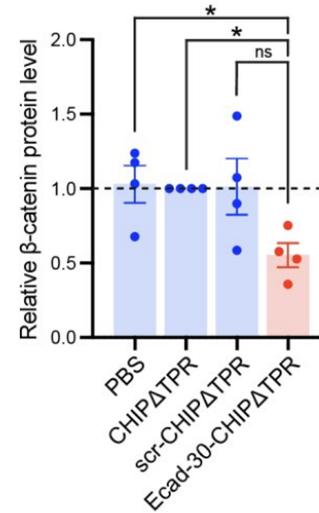
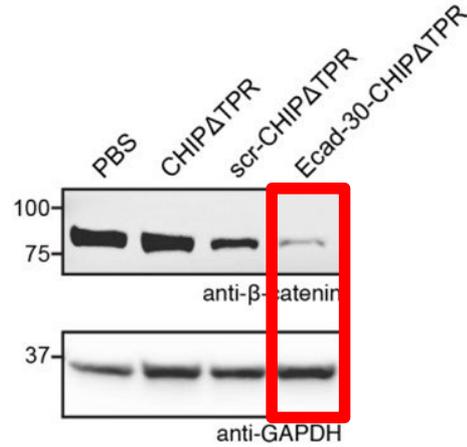
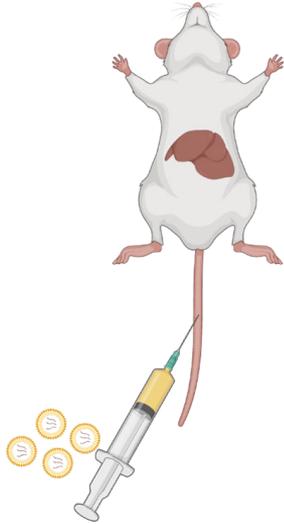


pLM-designed peptides work *in vivo*!

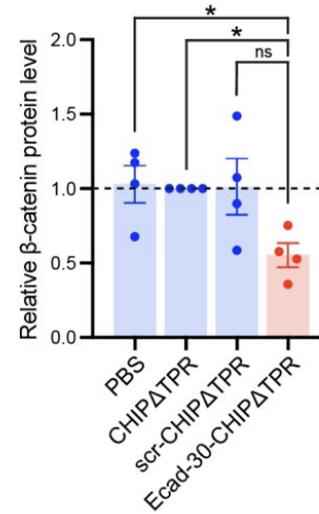
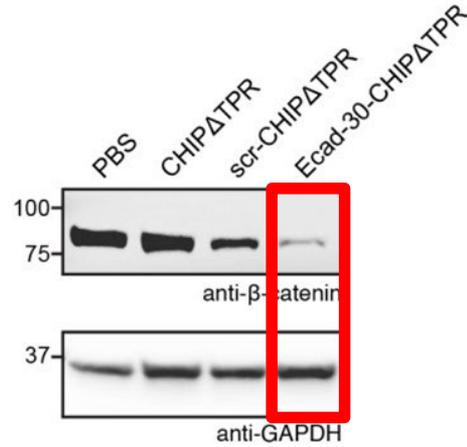
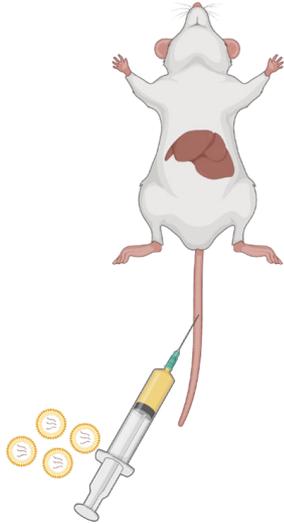




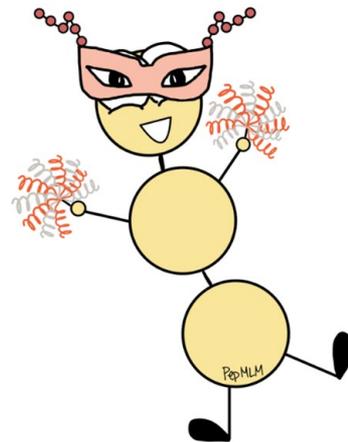
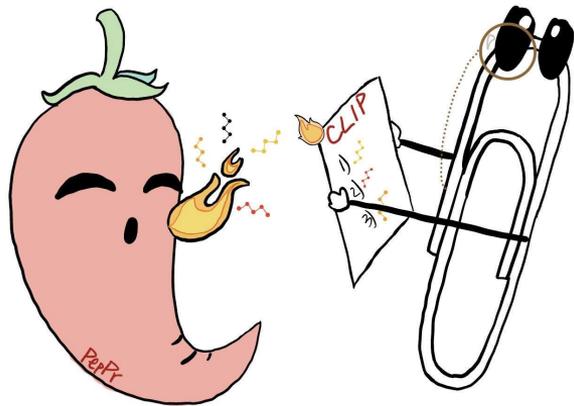
pLM-designed peptides work *in vivo*!



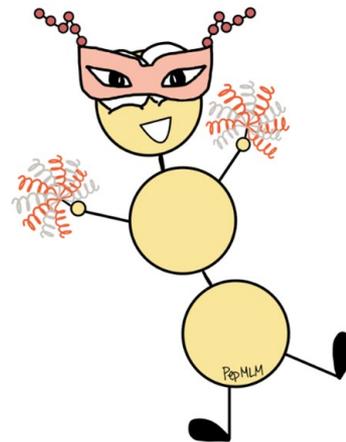
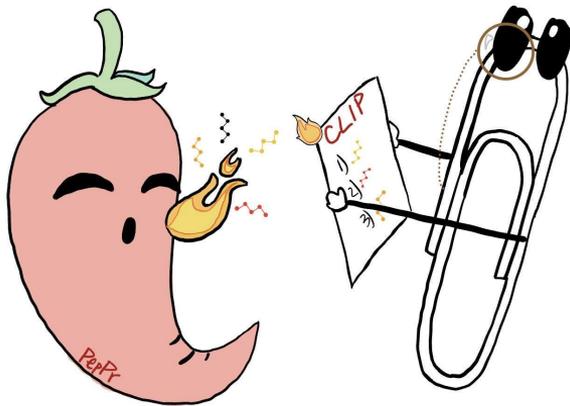
pLM-designed peptides work *in vivo*!



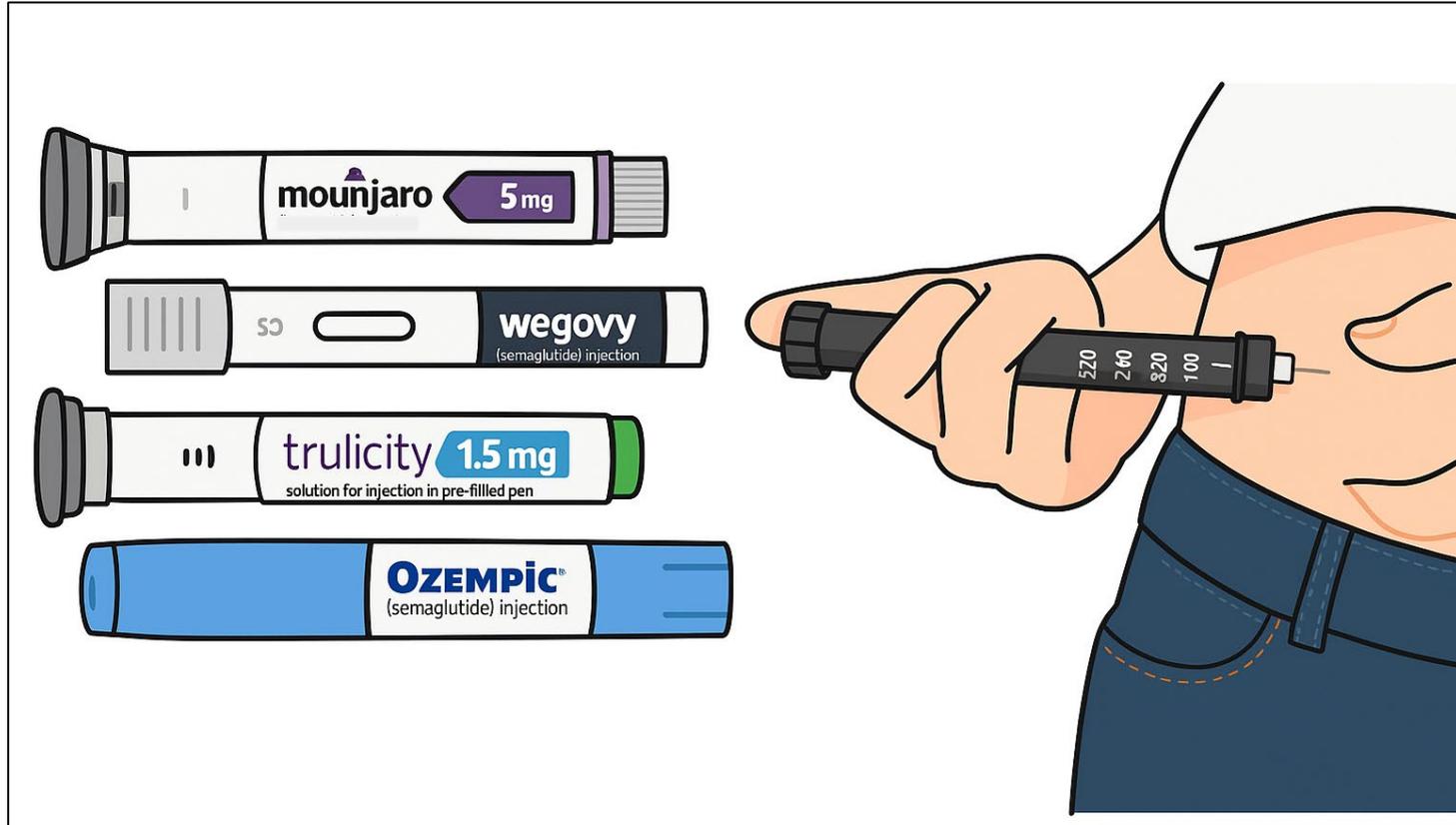
So clearly, sequence is enough to generate potent binders



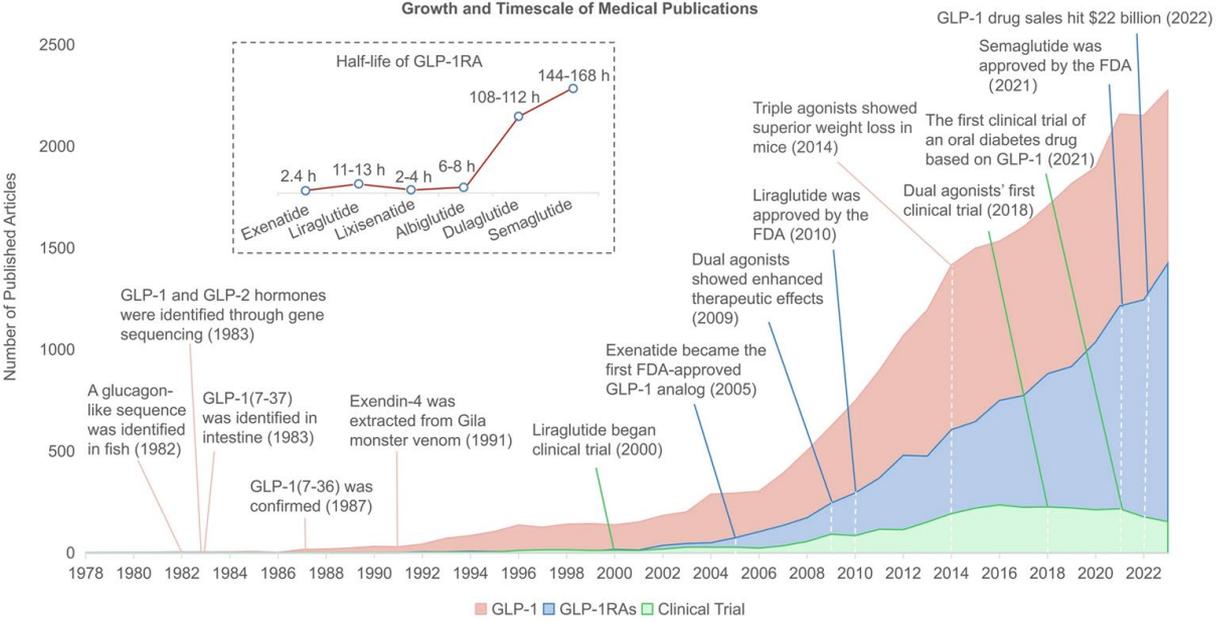
But just binding isn't enough...



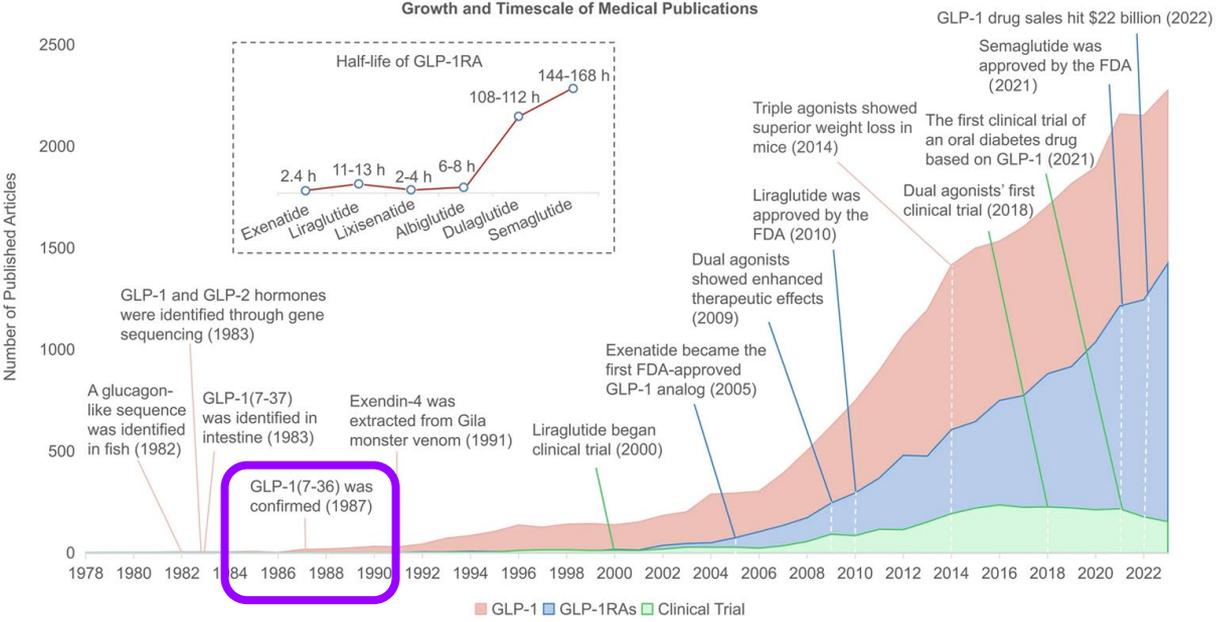
Consider the GLP-1 agonists



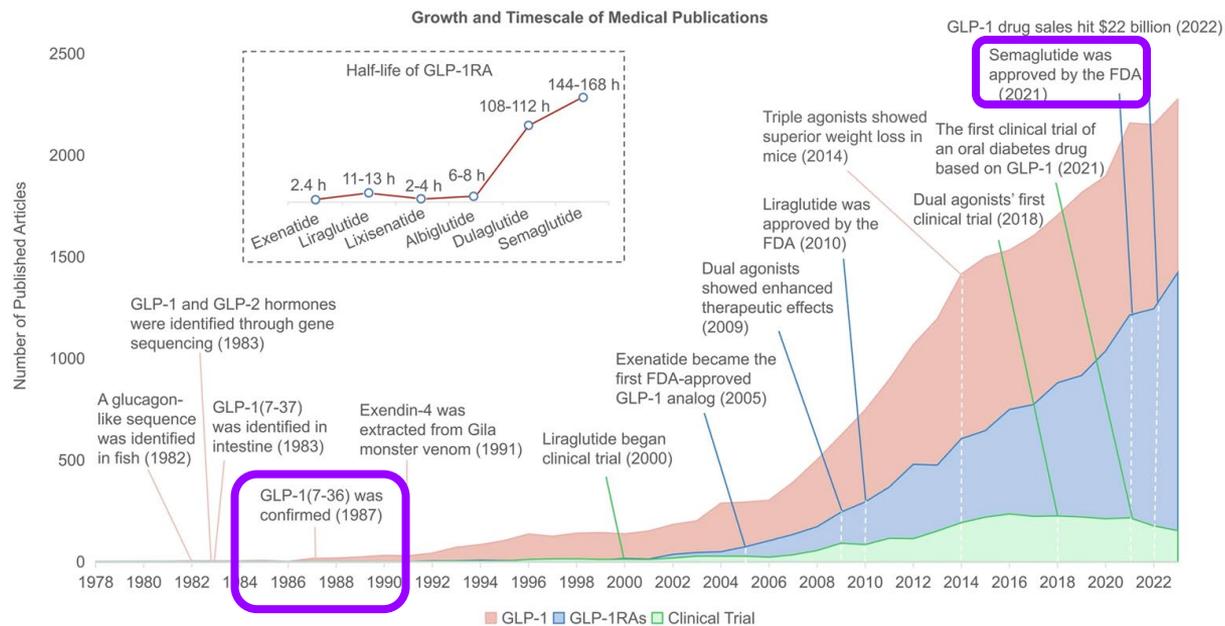
GLP-1R agonist peptides have been around for **DECADES**



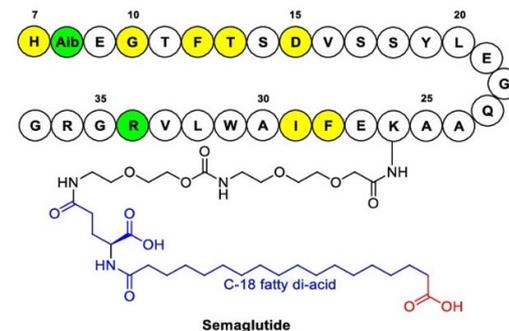
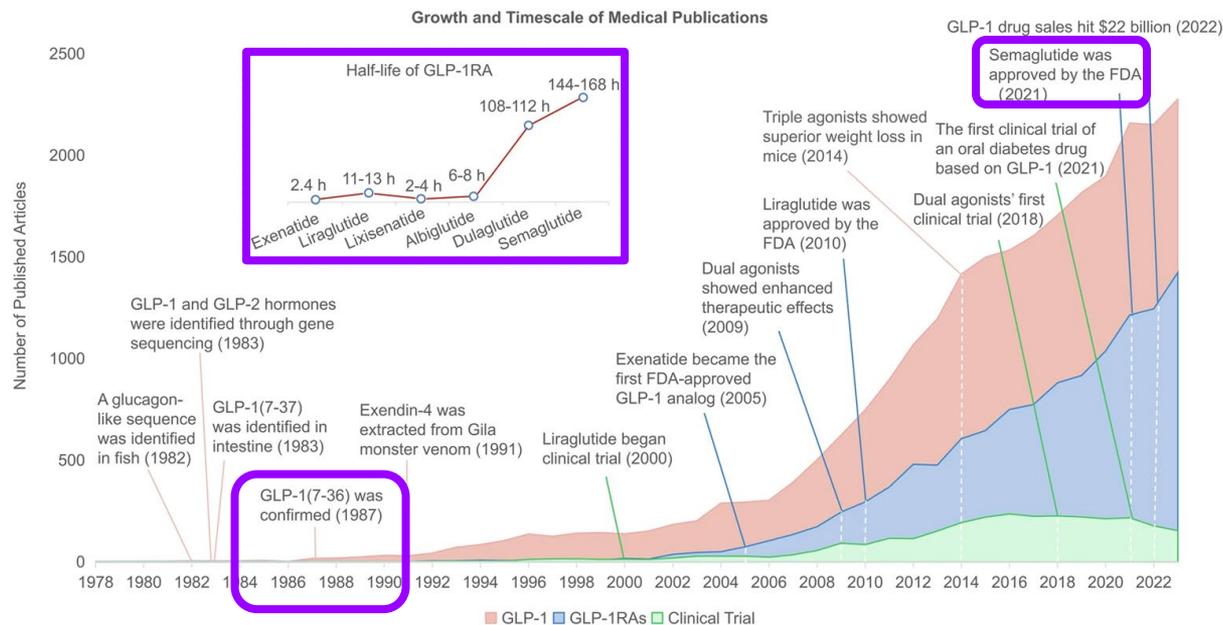
GLP-1R agonist peptides have been around for **DECADES**



What's taken so long??



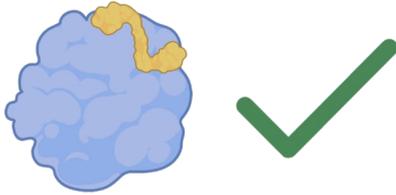
It took *years* to perfectly adjust these peptides to get optimal therapeutic properties !



Therapeutic design is **multi -objective**

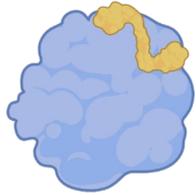
Therapeutic design is **multi -objective**

Binds Target

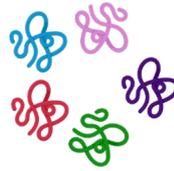


Therapeutic design is **multi -objective**

Binds Target

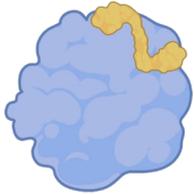


Soluble



Therapeutic design is **multi -objective**

Binds Target



Soluble



Non-Toxic

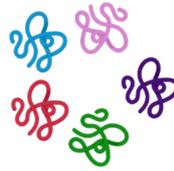


Therapeutic design is multi-objective

Binds Target



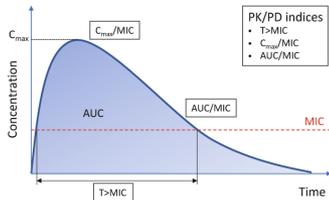
Soluble



Non-Toxic



Long Half-Life



Therapeutic design is multi-objective

Binds Target



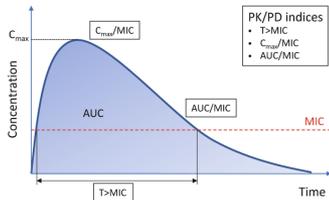
Soluble



Non-Toxic



Long Half-Life



Non-Fouling



Therapeutic design is multi-objective

Binds Target



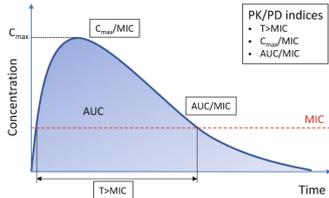
Soluble



Non-Toxic



Long Half -Life



Non-Fouling



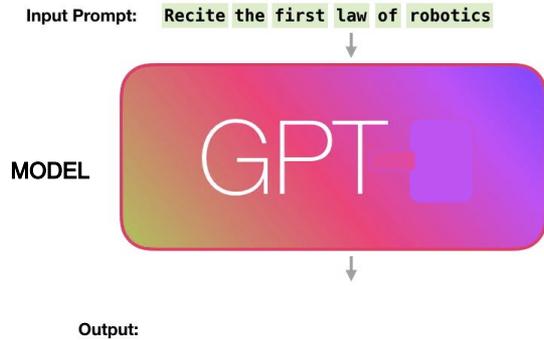
Orally Bioavailable



Current generation frameworks do not support this...

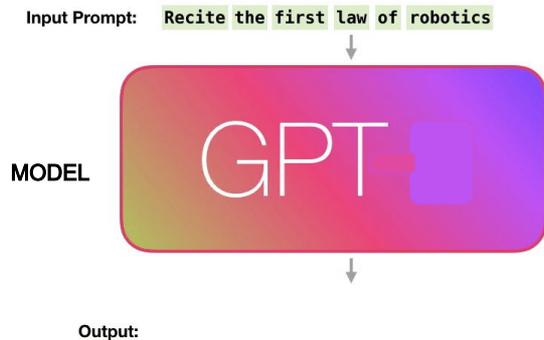
Current generation frameworks do not support this...

Zero (0)
Objective -Guided
Generation

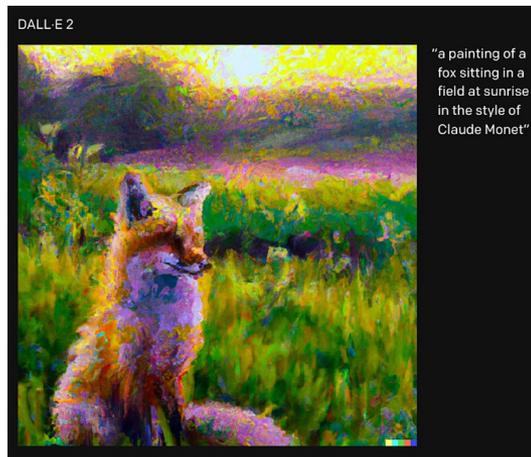


Current generation frameworks do not support this...

Zero (0) Objective -Guided Generation

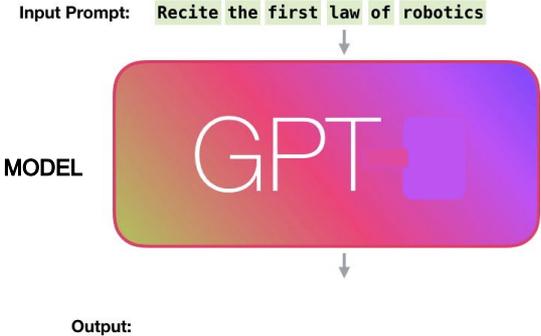


Single (1) Objective -Guided Generation

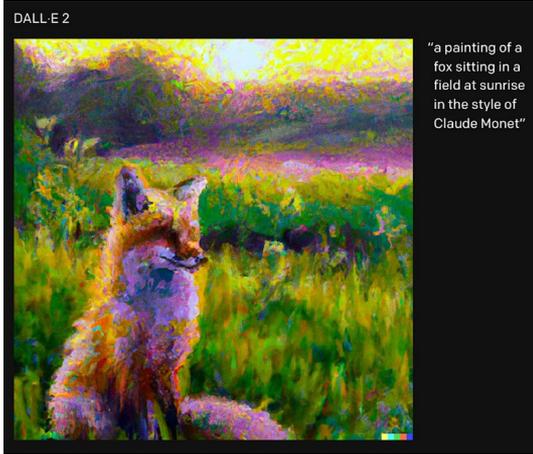


But maybe we can still get clues from these?

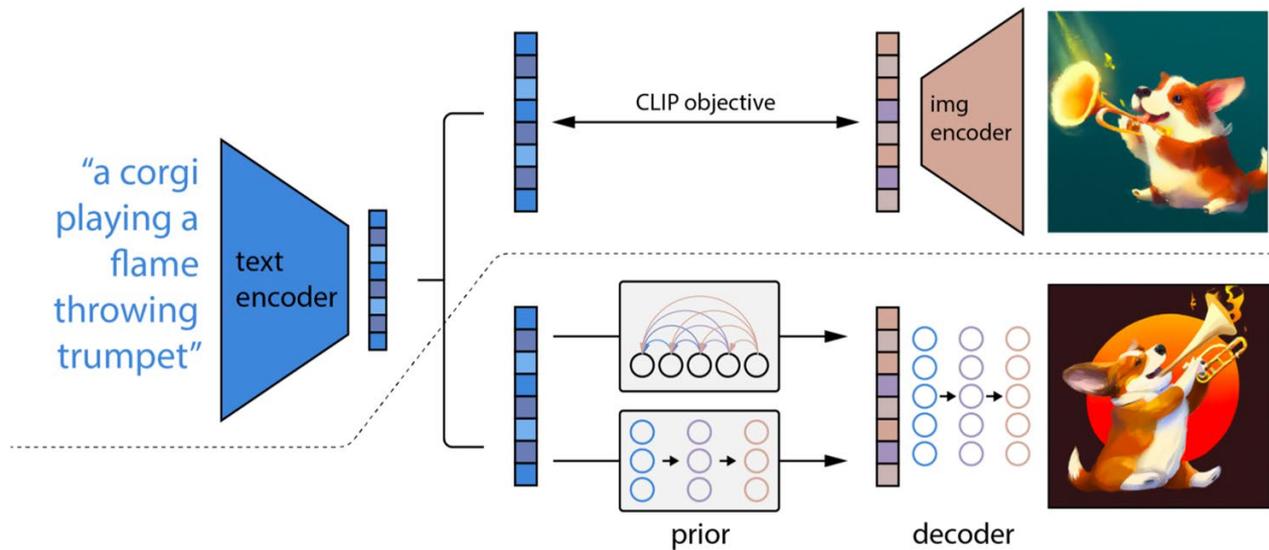
Zero (0) Objective -Guided Generation



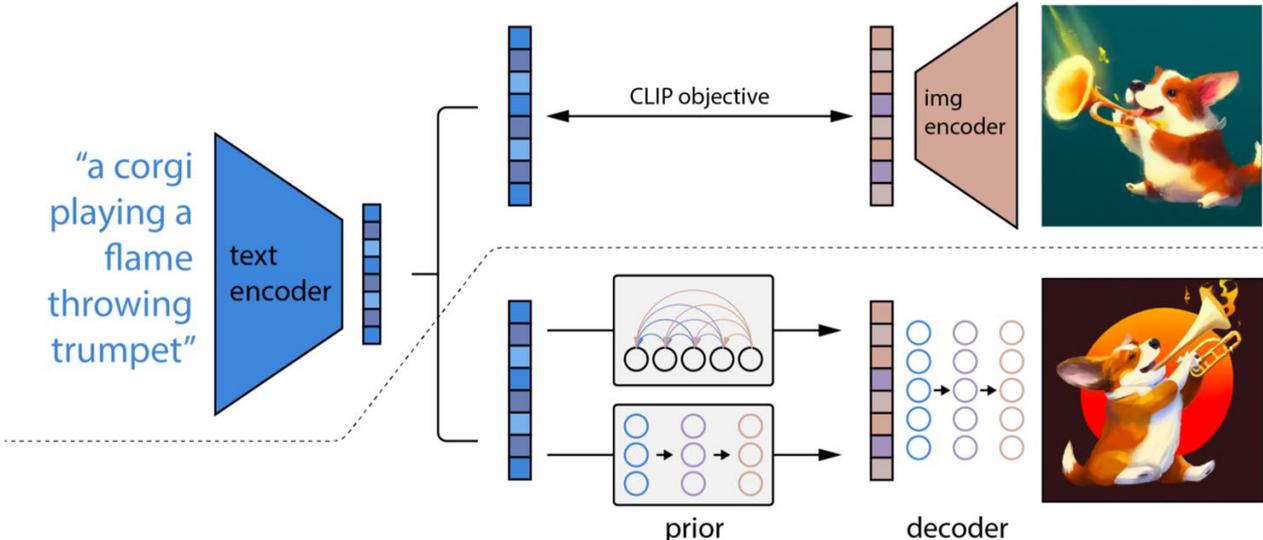
Single (1) Objective -Guided Generation



Language-Guided Diffusion

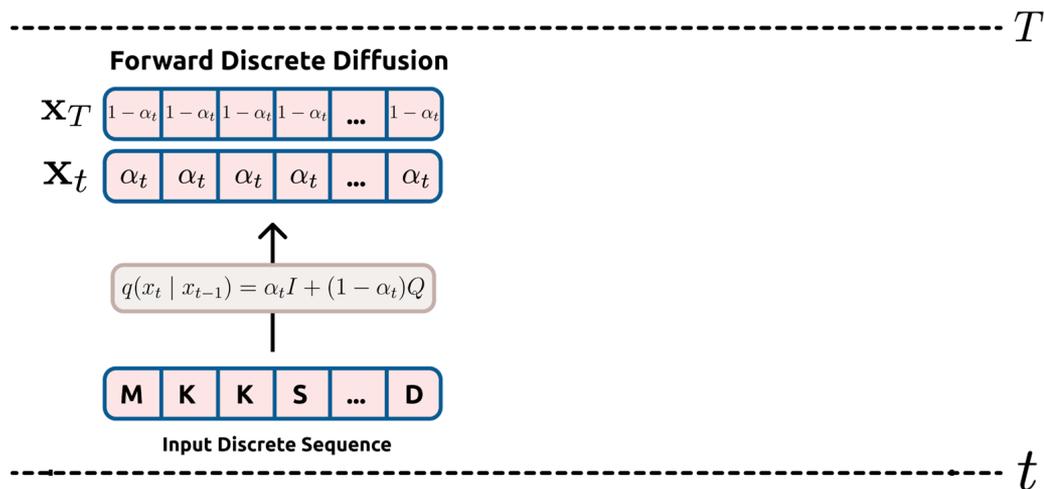


Can we instead do this for **sequence** generation?



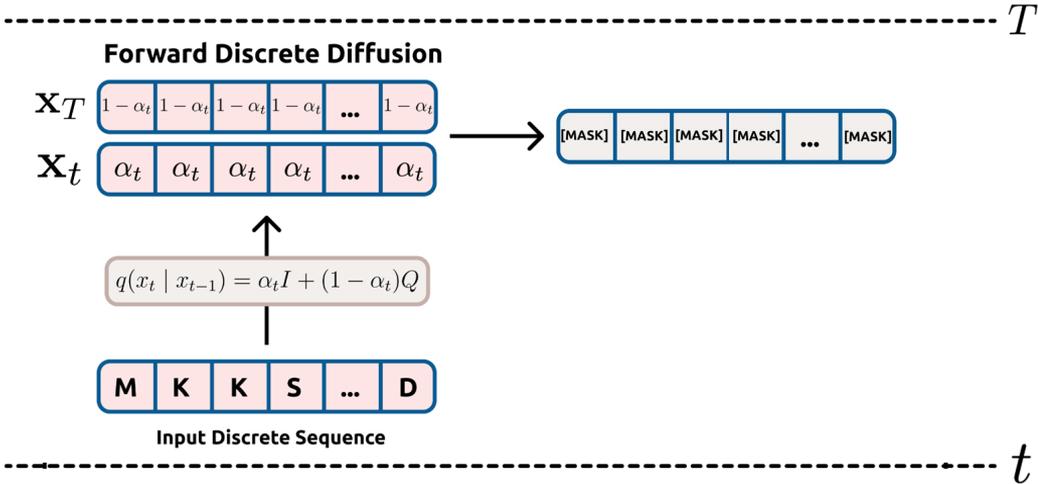


Discrete Diffusion



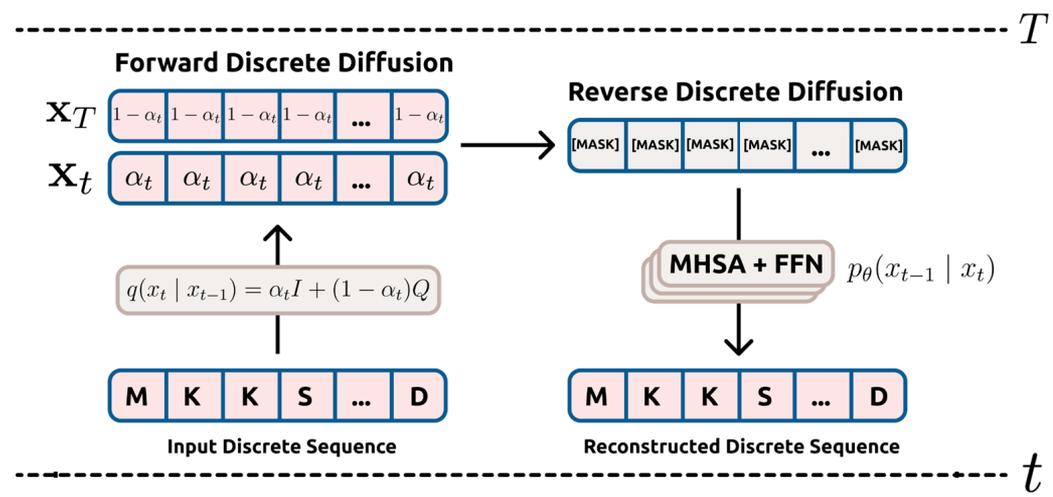


Discrete Diffusion



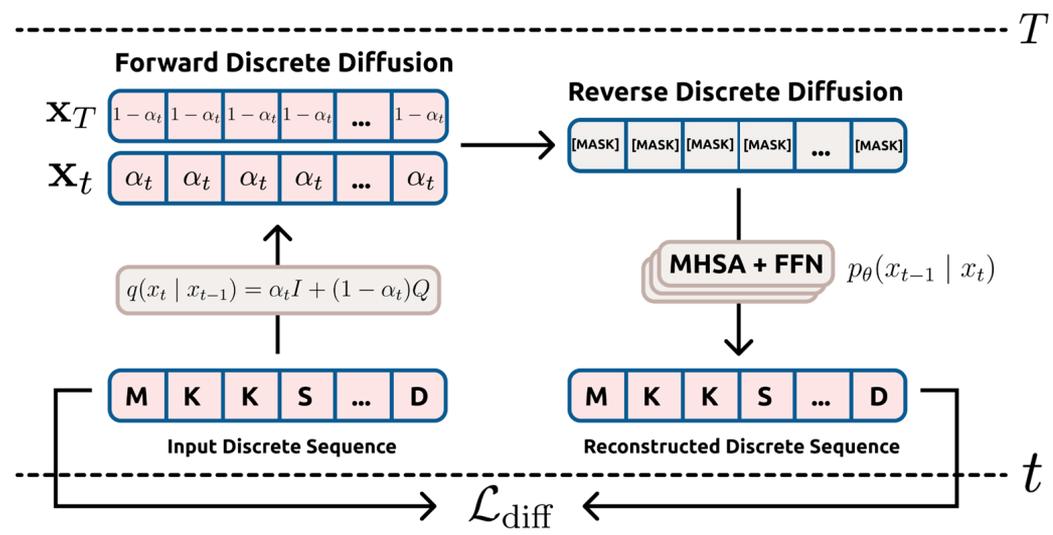


Discrete Diffusion



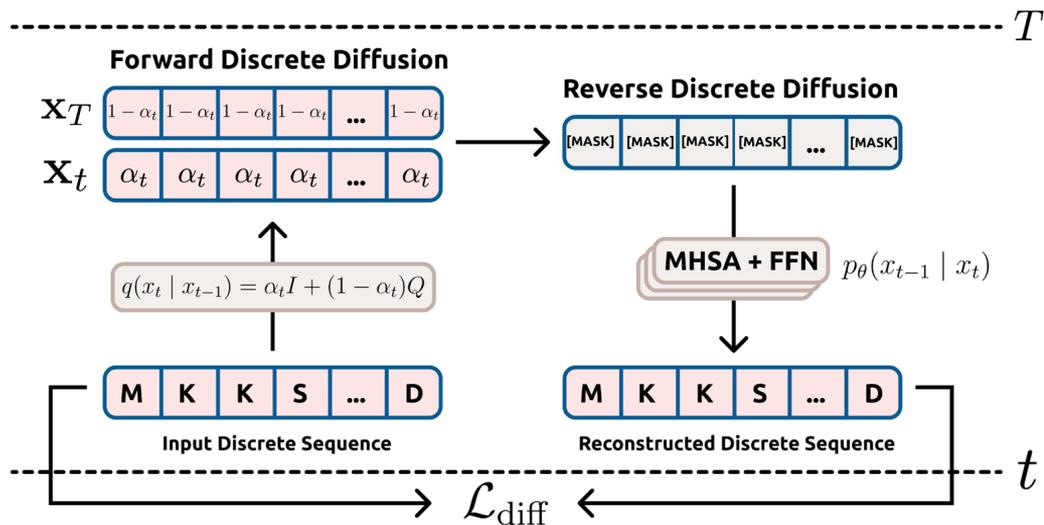


Discrete Diffusion





Masked Diffusion Language Model (MDLM)



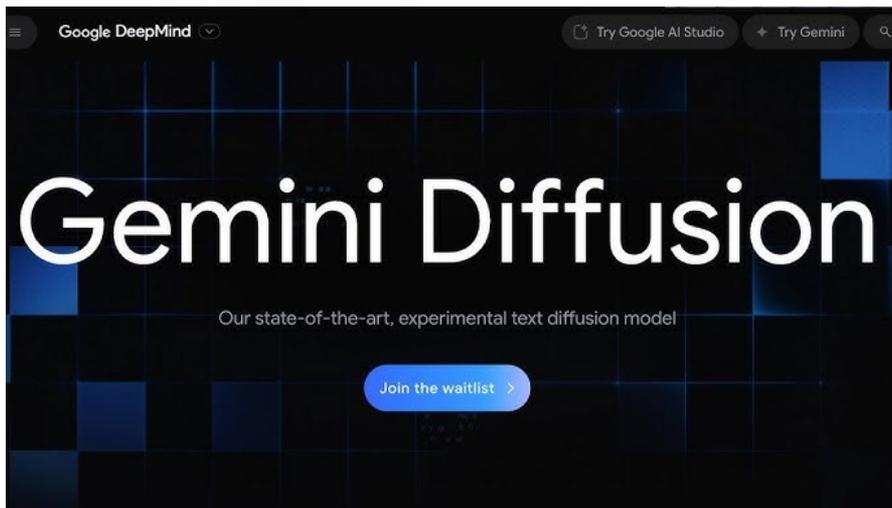


Masked Diffusion Language Model (MDLM)

said

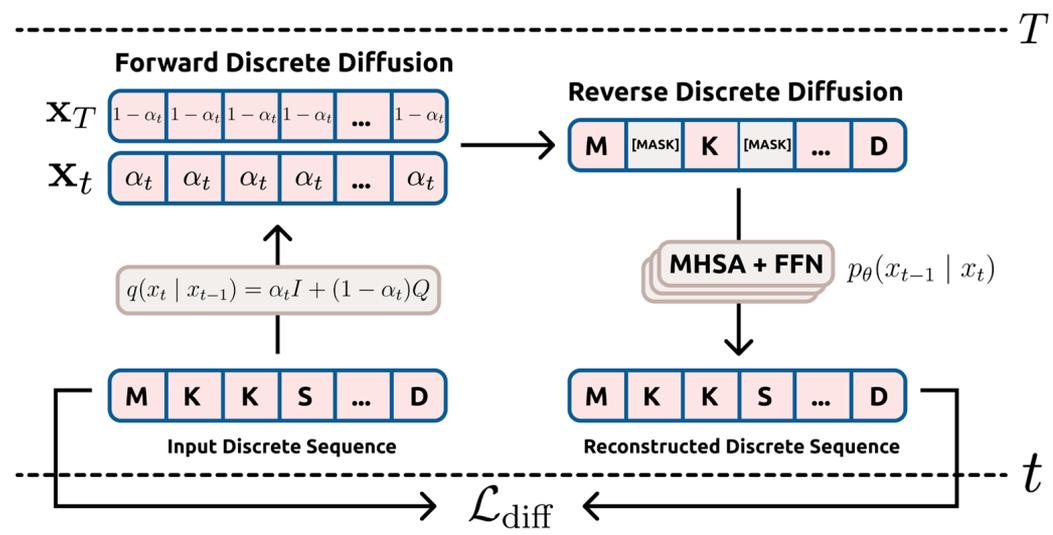


MDLMs are the future of language models



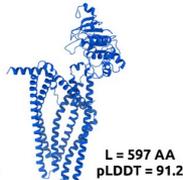
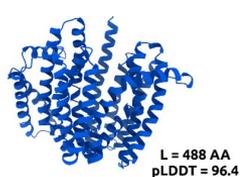
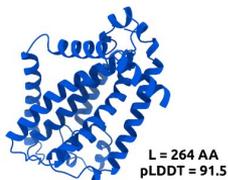
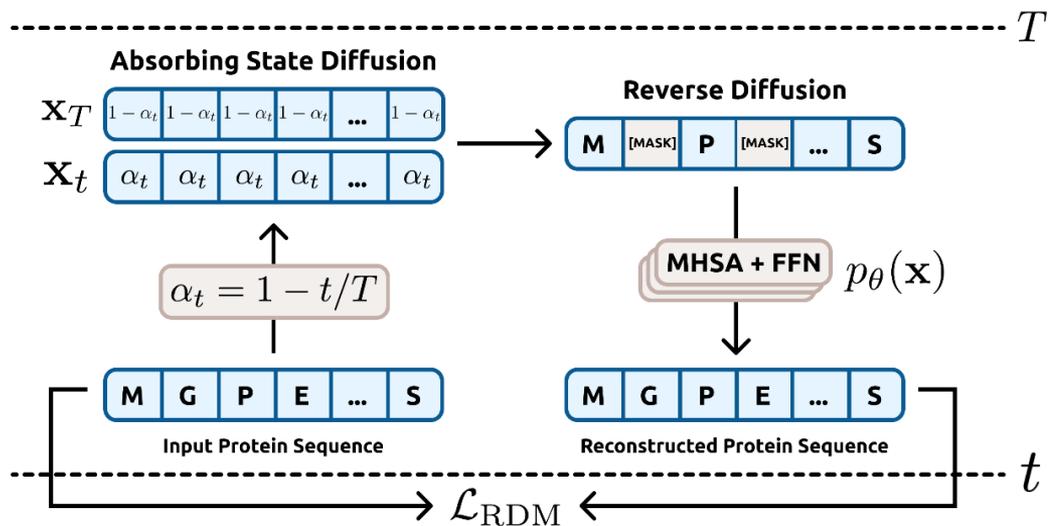


Masked Diffusion Language Model (MDLM)



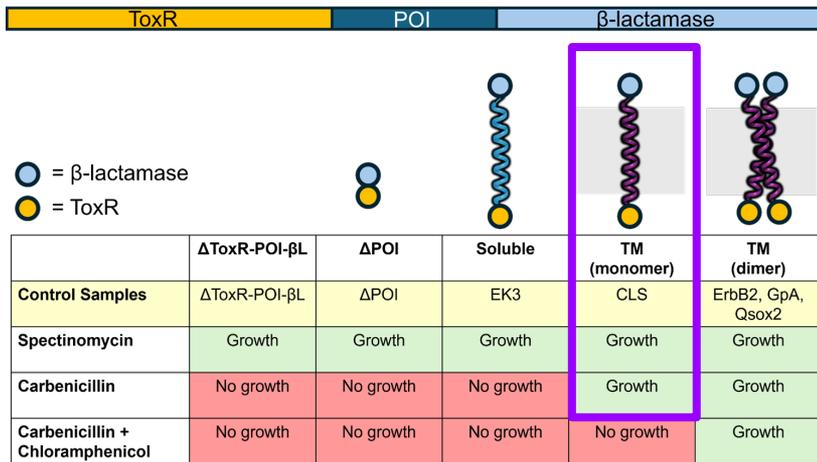


Membrane Diffusion Language Models (MemDLM)



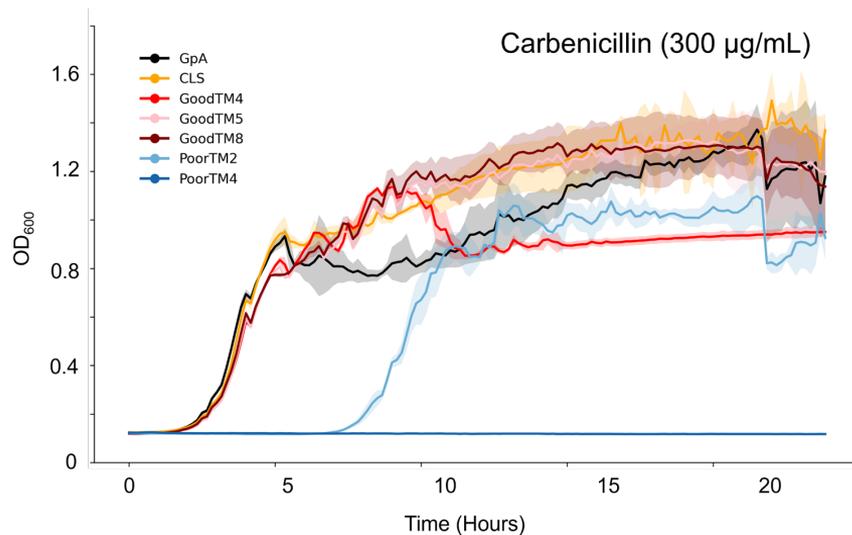
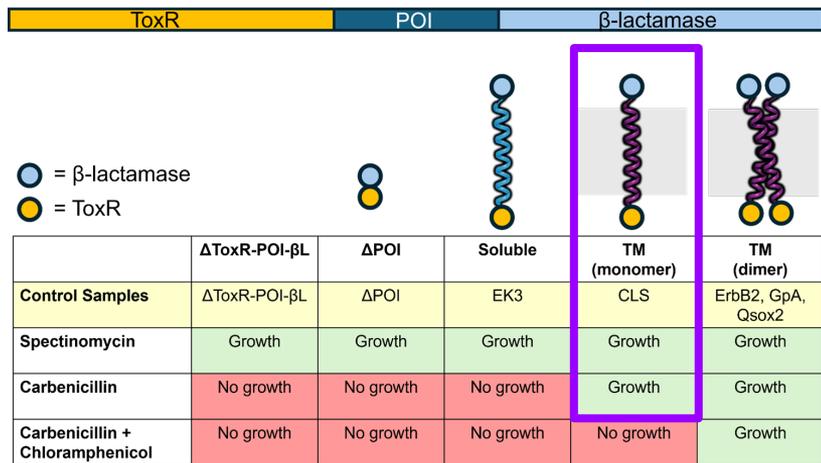


TOXCAT Assay to test valid membrane insertion



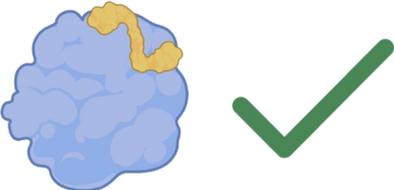


Great! So MDLMs can make valid biological sequences.

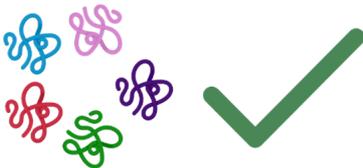


But we need to generate **optimized** sequences.

Binds Target



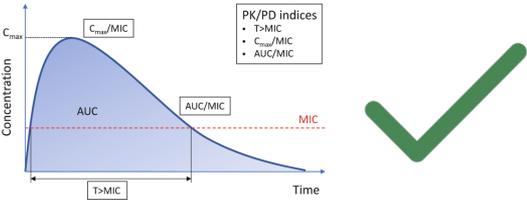
Soluble



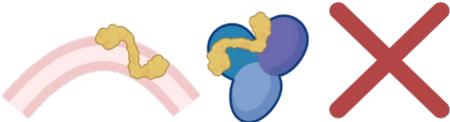
Non-Toxic



Long Half -Life



Non-Fouling

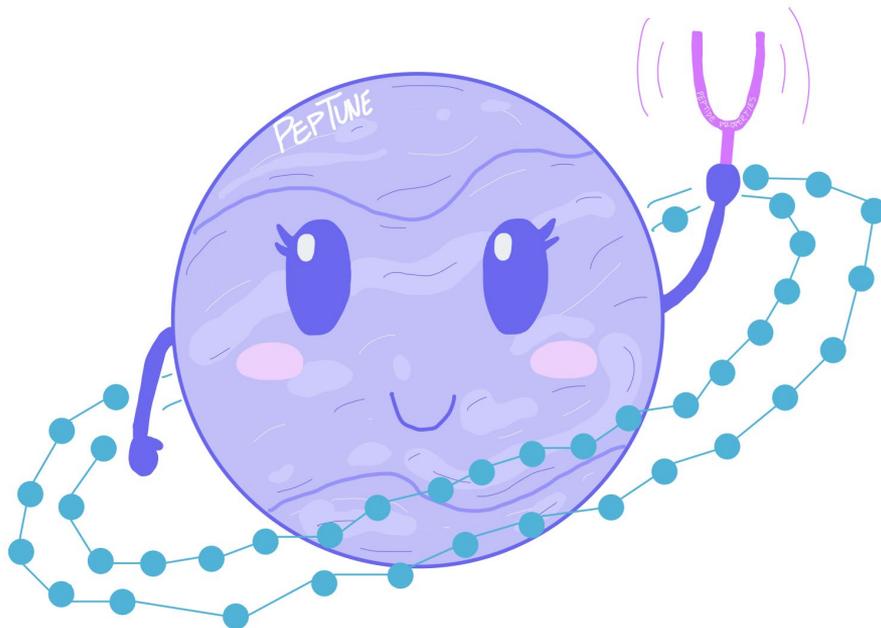


Orally Bioavailable



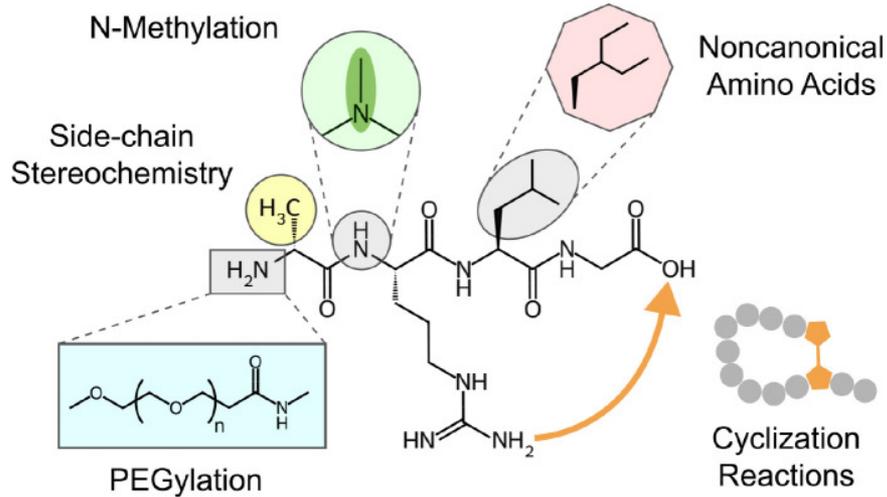


PepTune



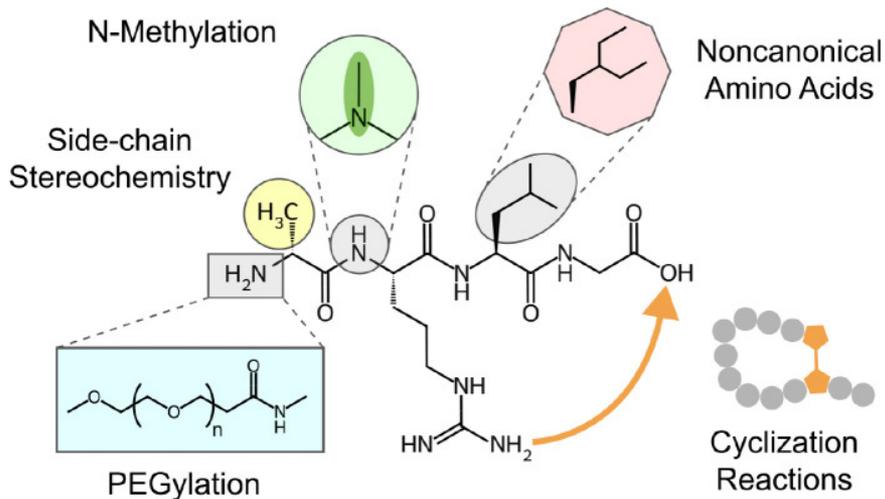


PepTune can model **modified** peptides!

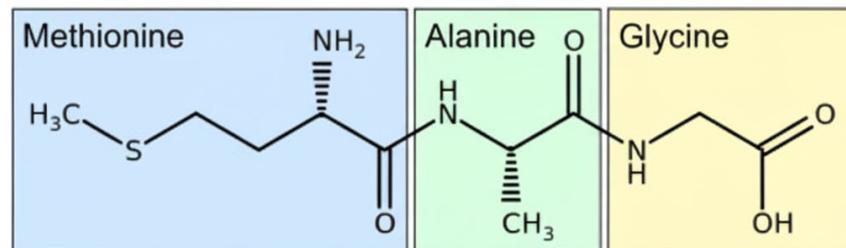




PepTune uses chemical language



Tripeptide Structural Representation



Canonical SMILES string

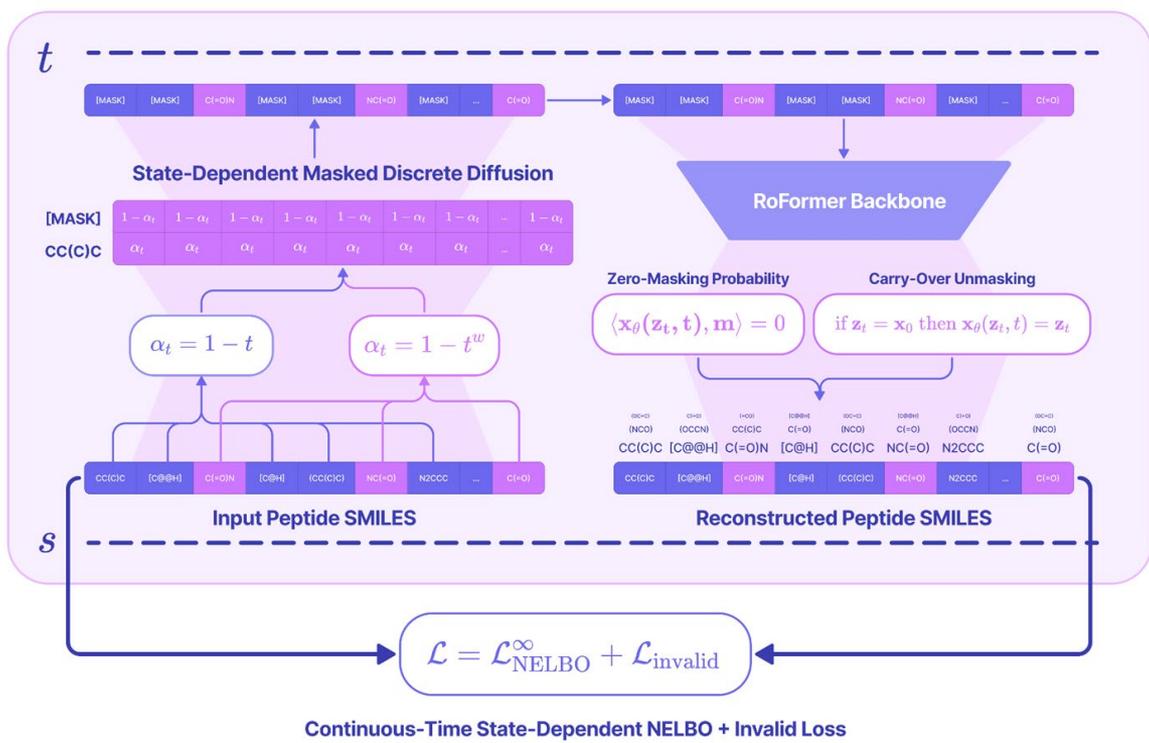
CSCC[C@H](N)C(=O)N[C@@H](C)C(=O)NCC(O)=O

Tokenized string

CSCC [C@H] (N) C(=O) N [C@@H] (C) C(=O) NCC (O) =O

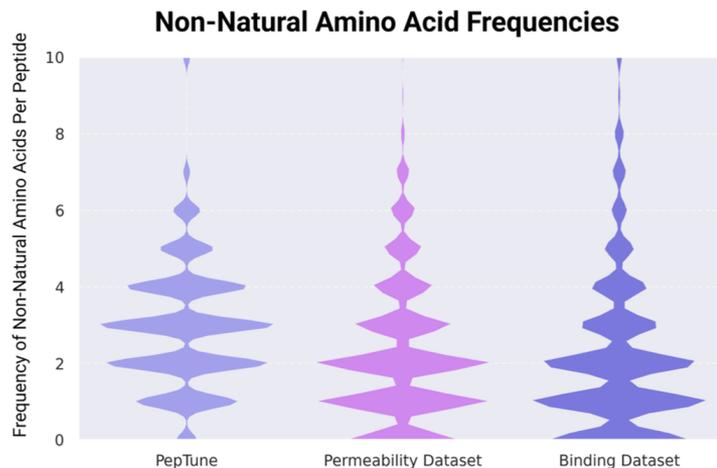


Masked Discrete Diffusion to Generate Peptides





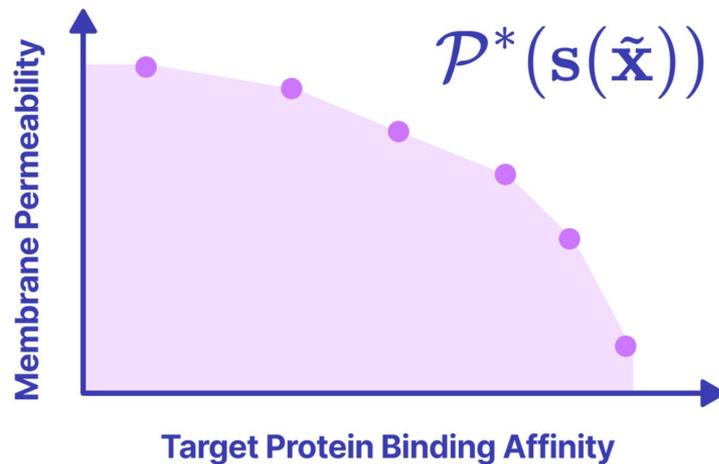
PepMDLM generates diverse new, modified peptides!



	Permeability Data	Binding Data	PepMDLM
Mean nAAs Per Peptide	2.215	2.150	2.940
Cyclic Peptides (%)	0.467	0.027	0.100

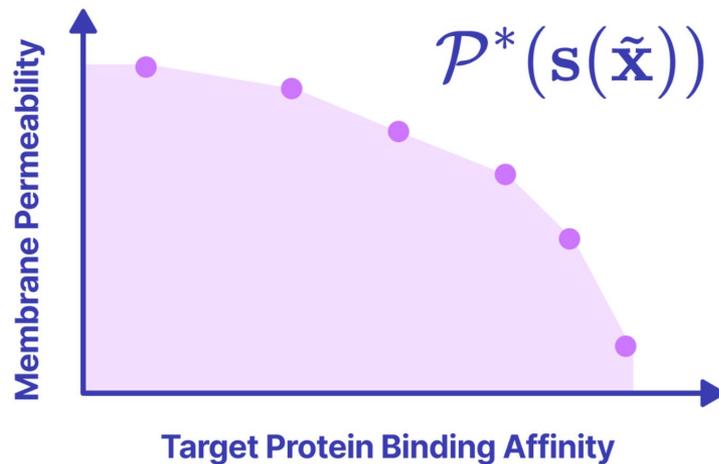


Goal: identify **Pareto optimal** peptide sequences across *all* properties





Goal: identify Pareto optimal peptide sequences across *all* properties





PeptiVerse

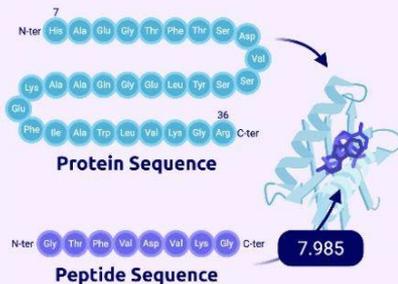


A Unified Platform for

**PEPTIDE THERAPEUTIC
PROPERTY PREDICTION**

Datasets of Experimentally-Validated Peptide Properties

Binding Affinity to Protein Target



Peptide Properties

Peptide			
Hemolysis	+	-	+
Solubility	+	+	-
Non-Fouling	-	+	+
Toxicity	-	+	+
Permeability	-7.25	-7.65	-7.06
Half-Life	5.2 hrs	6.2 hrs	4.9 hrs

PeptiVerse

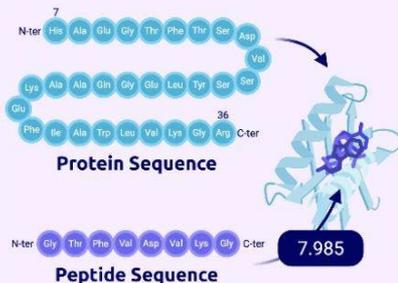


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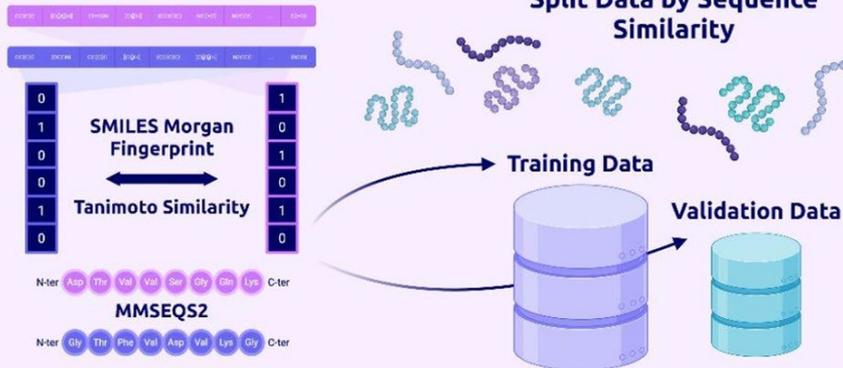
PeptiVerse



A Unified Platform for

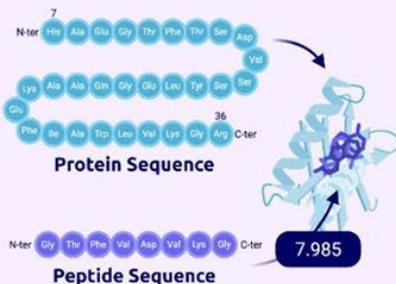
PEPTIDE THERAPEUTIC PROPERTY PREDICTION

Split Data by Sequence Similarity



Datasets of Experimentally-Validated Peptide Properties

Binding Affinity to Protein Target



Peptide Properties

Peptide			
Hemolysis	+	-	+
Solubility	+	+	-
Non-Fouling	-	+	+
Toxicity	-	+	+
Permeability	-7.25	-7.65	-7.06
Half-Life	5.2 hrs	6.2 hrs	4.9 hrs

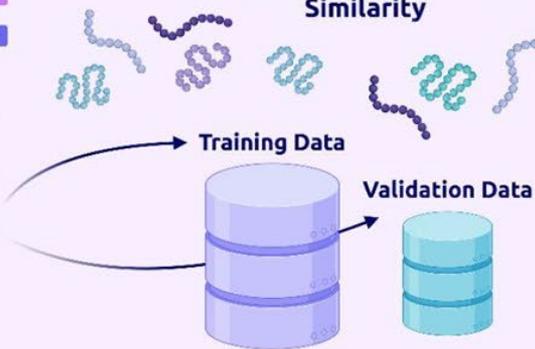
PeptiVerse



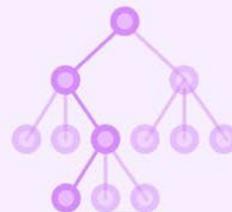
A Unified Platform for
**PEPTIDE THERAPEUTIC
 PROPERTY PREDICTION**



Split Data by Sequence Similarity



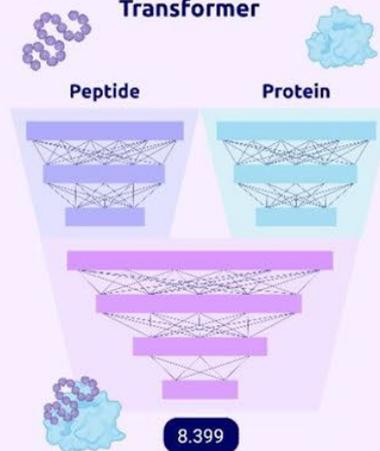
Train Classifier Models with Various Architectures



XGBoost
 Elastic Net
 SVM
 MLP
 CNN
 Transformer

0.637

Cross-Attention Transformer





PeptiVerse: Accurate peptide property predictors

The screenshot shows the PeptiVerse web application interface. At the top, there is a navigation bar with "Spaces", "ChatterjeeLab", and "PeptiVerse" (with a dropdown arrow). To the right of the navigation bar are icons for "App", "Files", "Community", and "Settings". The main header features the "PeptiVerse" logo in a stylized blue font, with the tagline "A Unified Platform for PEPTIDE THERAPEUTIC PROPERTY PREDICTION" below it. A small cartoon rocket character is positioned to the left of the tagline. Below the header, there are four tabs: "Predict" (selected), "Best Models", "Distributions", and "Documentation". The "Predict" tab is active, showing two main sections: "Input" and "Select Properties".

Input Section:

- Input Type:** "Sequence" and "SMILES" (selected).
- Peptide Sequence(s) / SMILES:** A text input field with the placeholder "Enter amino acid sequence(s) or SMILES, one per line".
- Protein Sequence (for binding prediction):** A text input field containing the sequence "MVHLTPPEEKSAVTALWGKVVNDEVGGEALGRLLVVYPWTQRFFESFGD LST".

Select Properties Section:

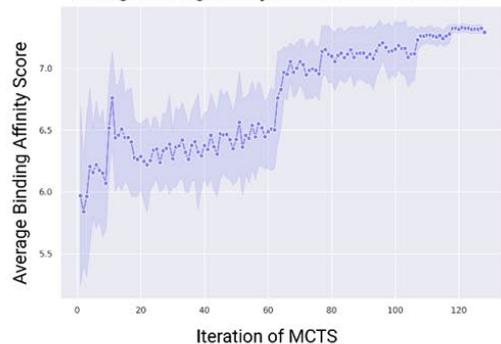
- Physicochemical Properties:** A dropdown menu with a downward arrow. It includes a checked checkbox for "Calculate Basic Properties" and the text "MW, net charge, pI, hydrophobicity (Sequence only)".
- pH for Net Charge:** A slider control. The current value is 7, and the range is from 0 to 14. The text "Physiological pH is ~7.4" is displayed below the slider.
- Prediction Properties:** A dropdown menu with a downward arrow. It includes several options:
 - Solubility ↑ (Not supported)
 - Permeability (Penetration) ↑ (Not supported)
 - Hemolysis ↓
 - Non-Fouling ↑



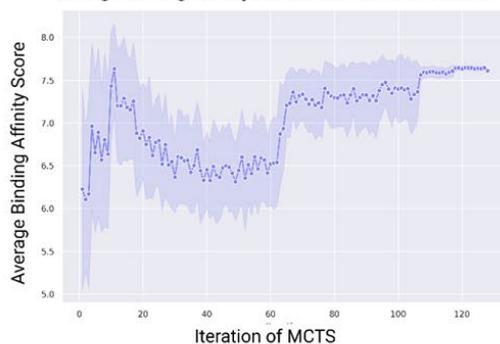


PeptiVerse + PepMDLM

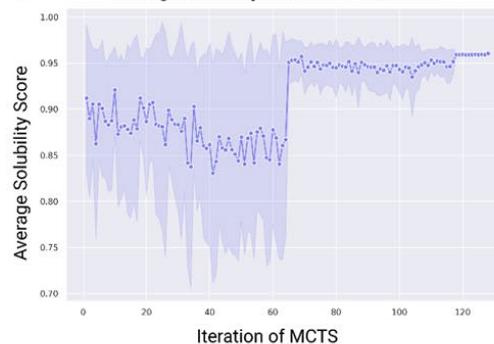
Average Binding Affinity to GLAST Over Iterations



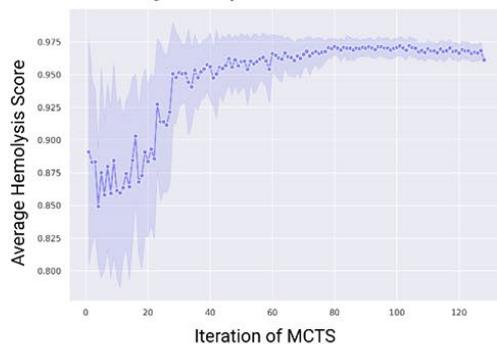
Average Binding Affinity Scores to TfR Over Iterations



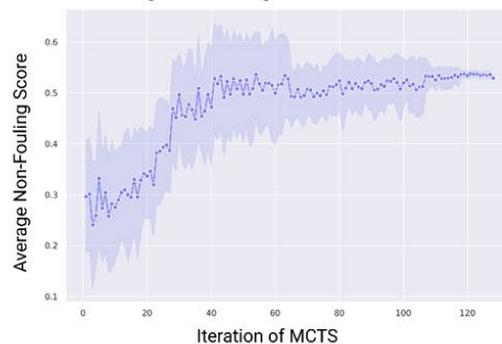
Average Solubility Score Over Iterations



Average Hemolysis Score Over Iterations

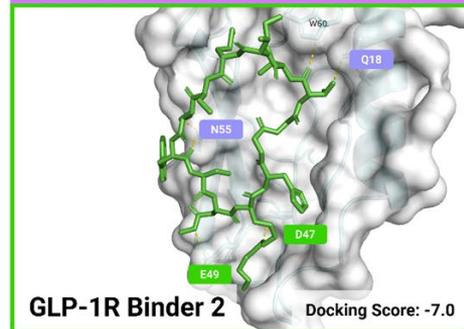
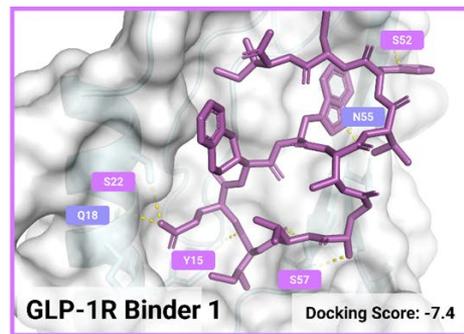
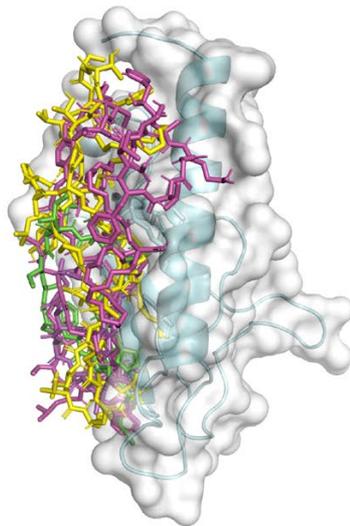
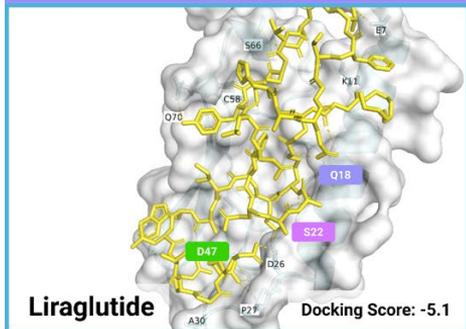
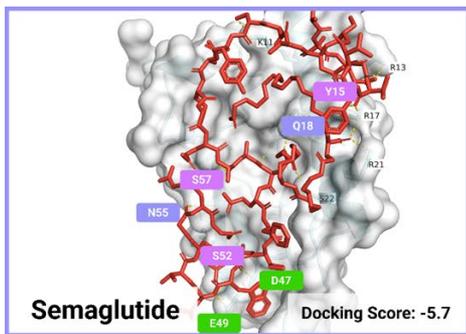


Average Non-Fouling Score Over Iterations



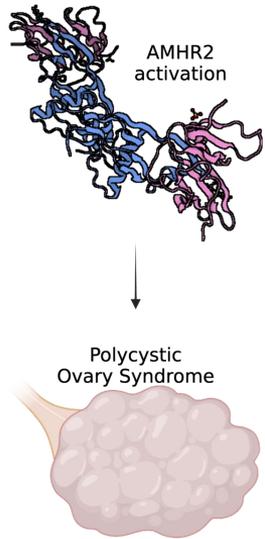


New GLP-1R agonists



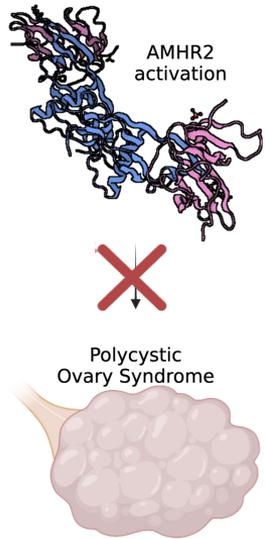


The AMH-AMHR2 axis drives PCOS



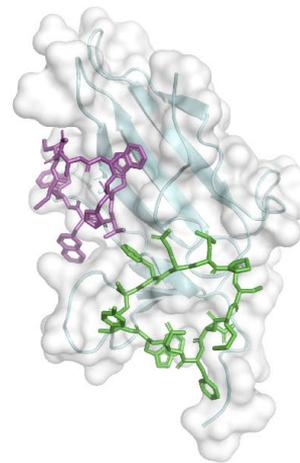
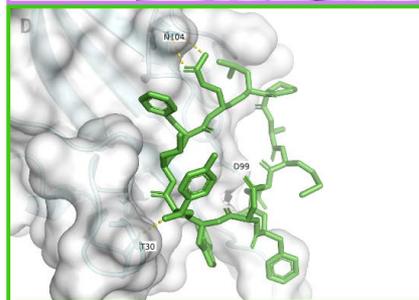
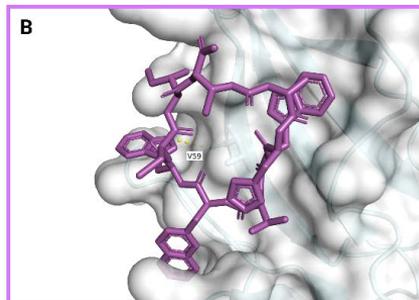
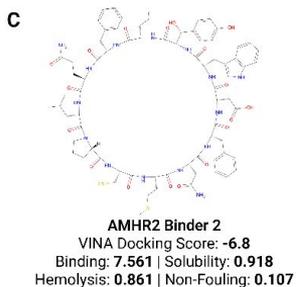
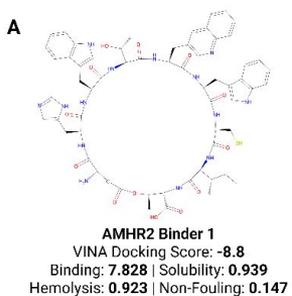
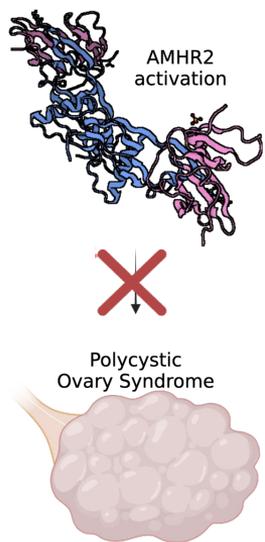


The AMH-AMHR2 axis drives PCOS



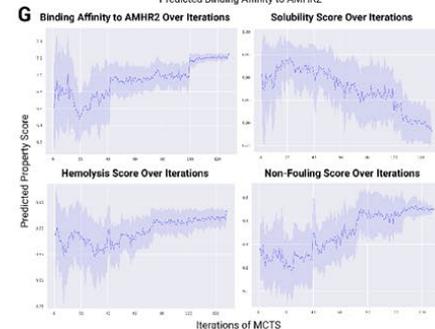
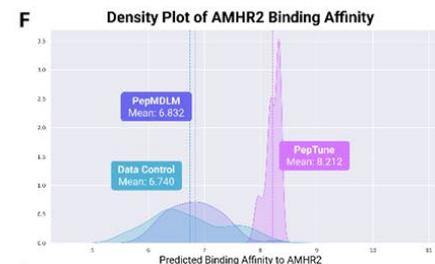
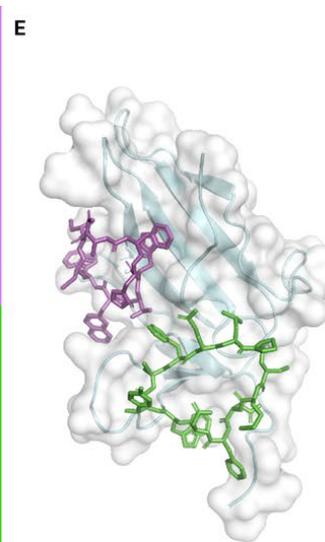
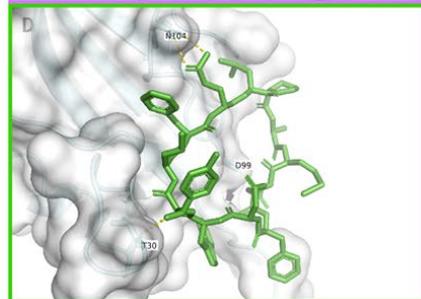
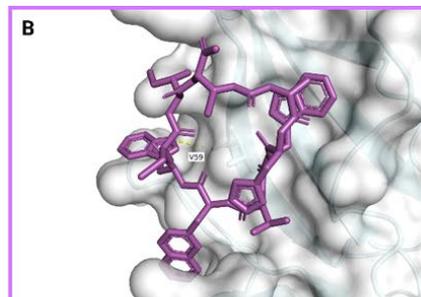
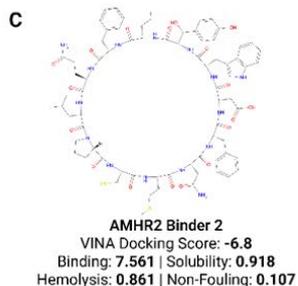
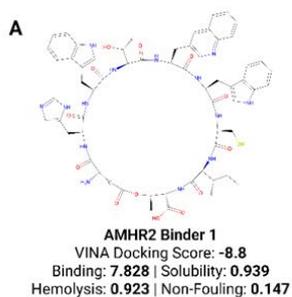
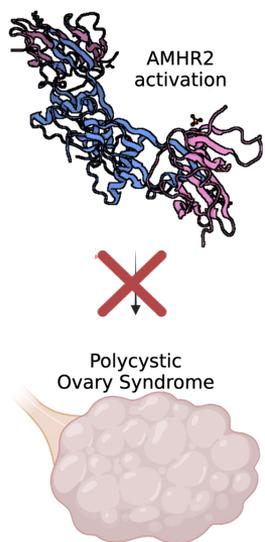


PepTune generates therapeutic antagonists of AMHR2





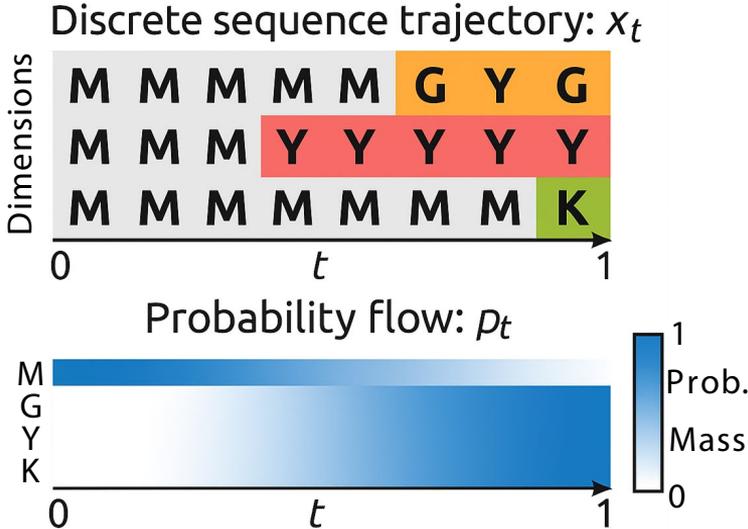
PepTune generates therapeutic antagonists of AMHR2 with ready-to-go, optimal drug-like properties





Discrete Flow Matching (DFM)!

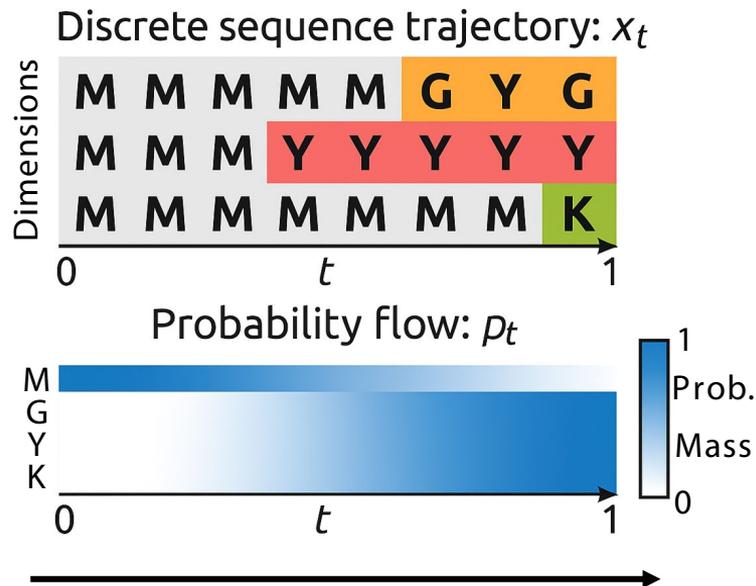
$$\mathbf{x}_{t+1} = \mathbf{x}_t + u_{\theta}(\mathbf{x}_t, t)$$





Discrete Flow Matching (DFM)!

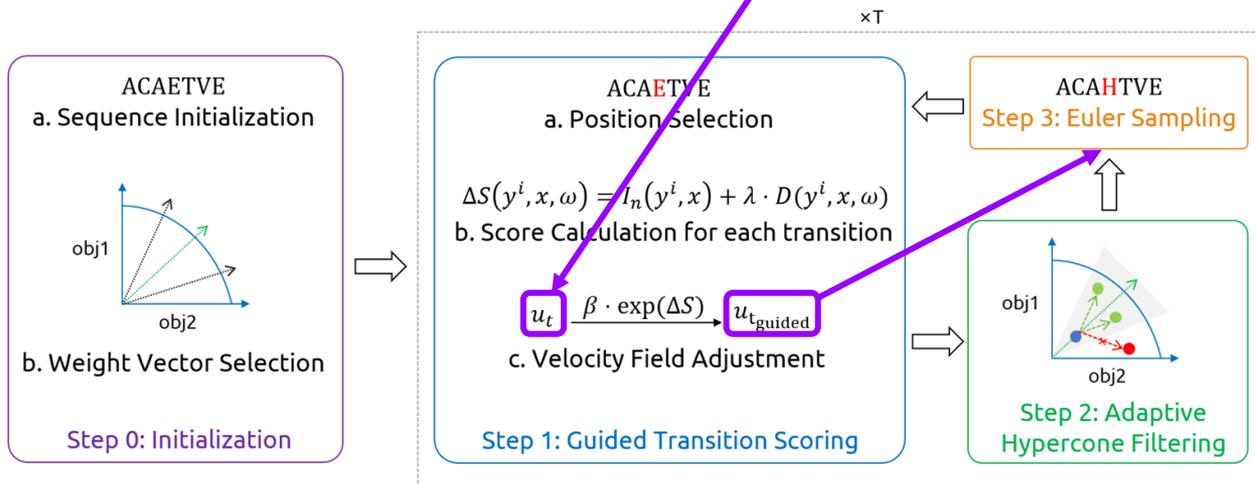
$$\mathbf{x}_{t+1} = \mathbf{x}_t + u_{\theta}(\mathbf{x}_t, t)$$





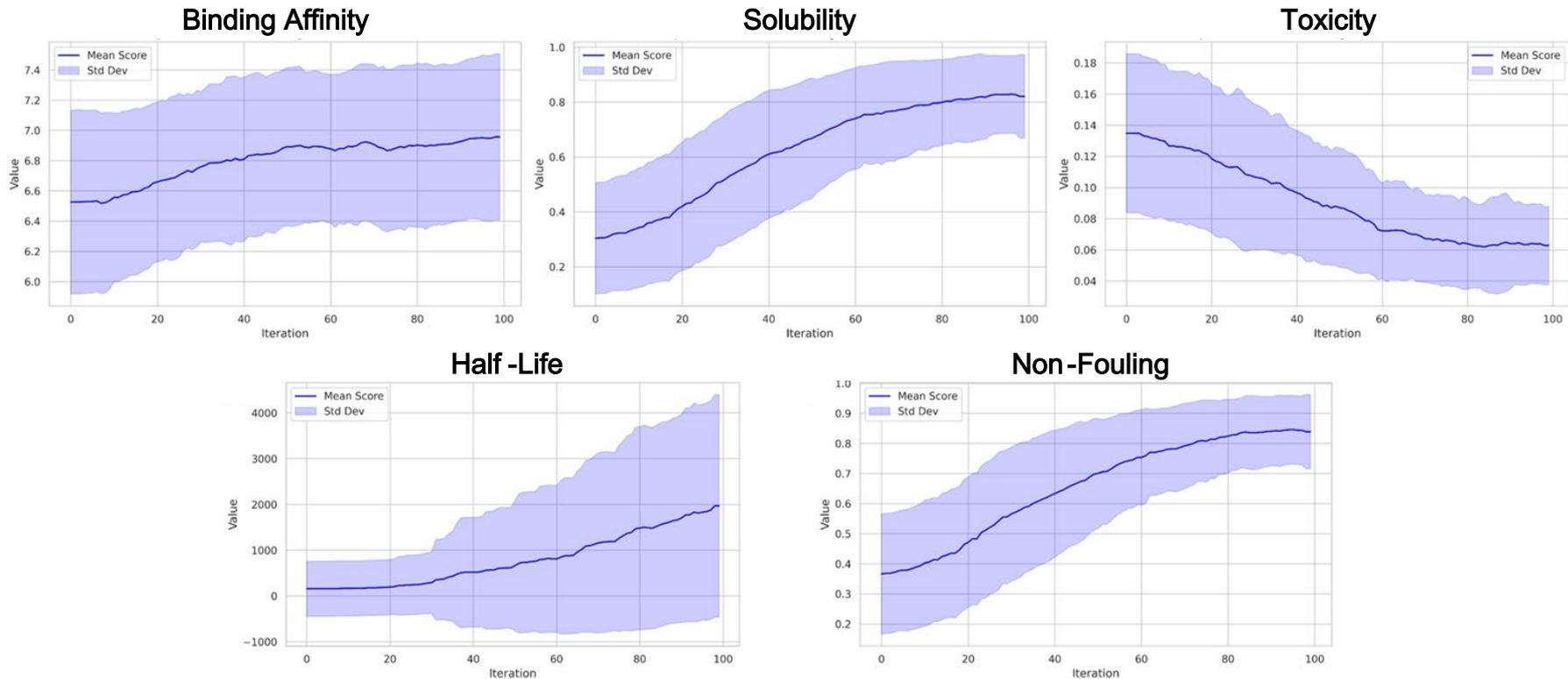
Multi-Objective-Guided Discrete Flow Matching

$$\mathbf{x}_{t+1} = \mathbf{x}_t + u_{\theta}(\mathbf{x}_t, t)$$





MOG-DFM optimizes properties during peptide generation



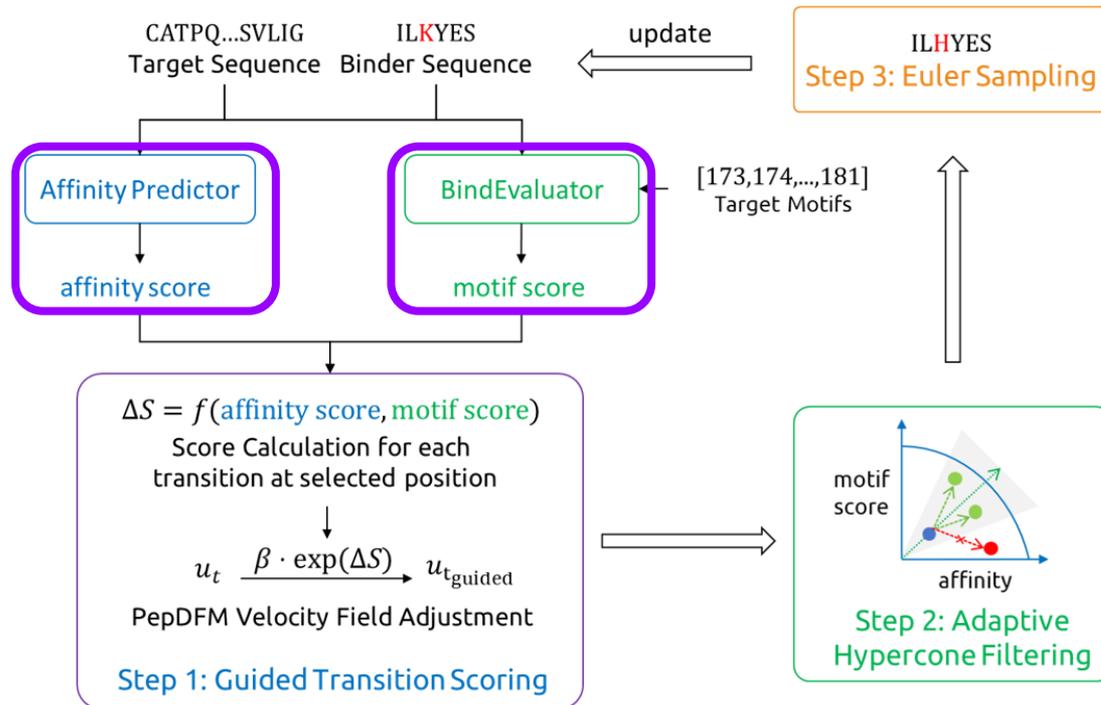


MOG-DFM enables highly specific peptide design with
moPPIt !



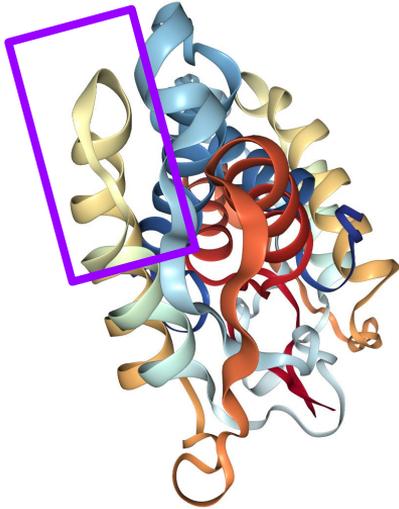


MOG-DFM enables highly specific peptide design with moPPIt!



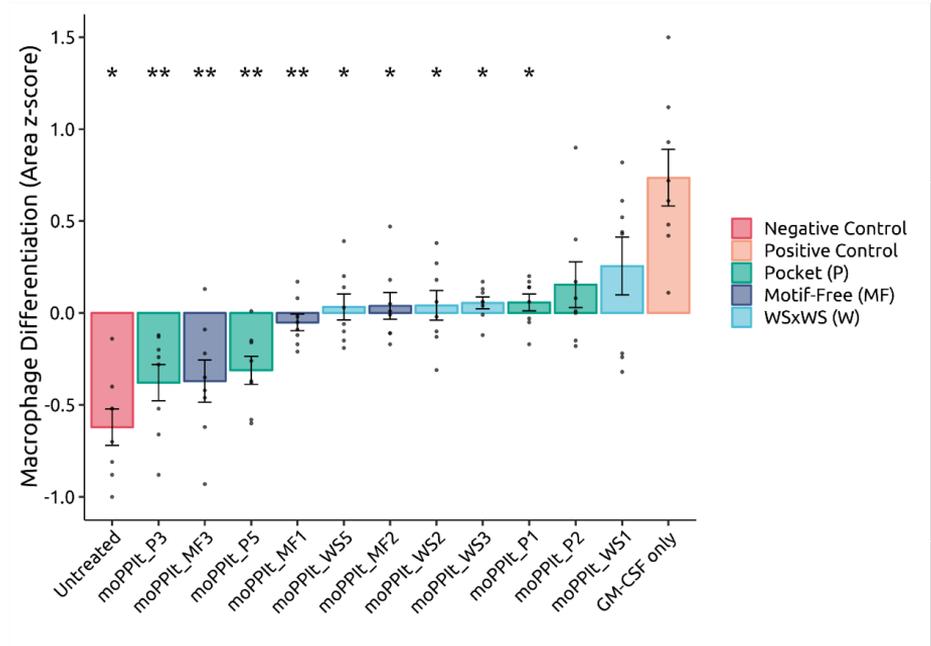
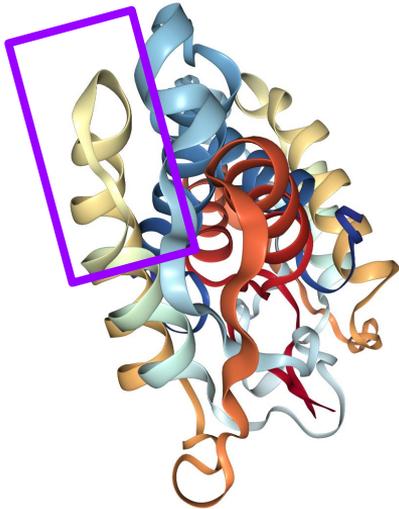


moPPIt-designed peptides bind **receptors** like GM-CSFR

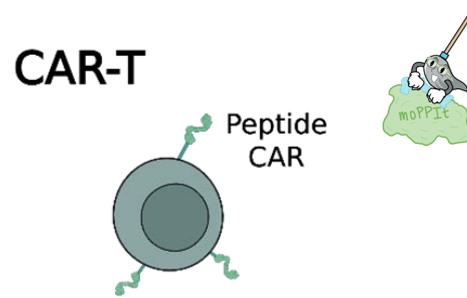




Enhance endometrial receptivity and implantation

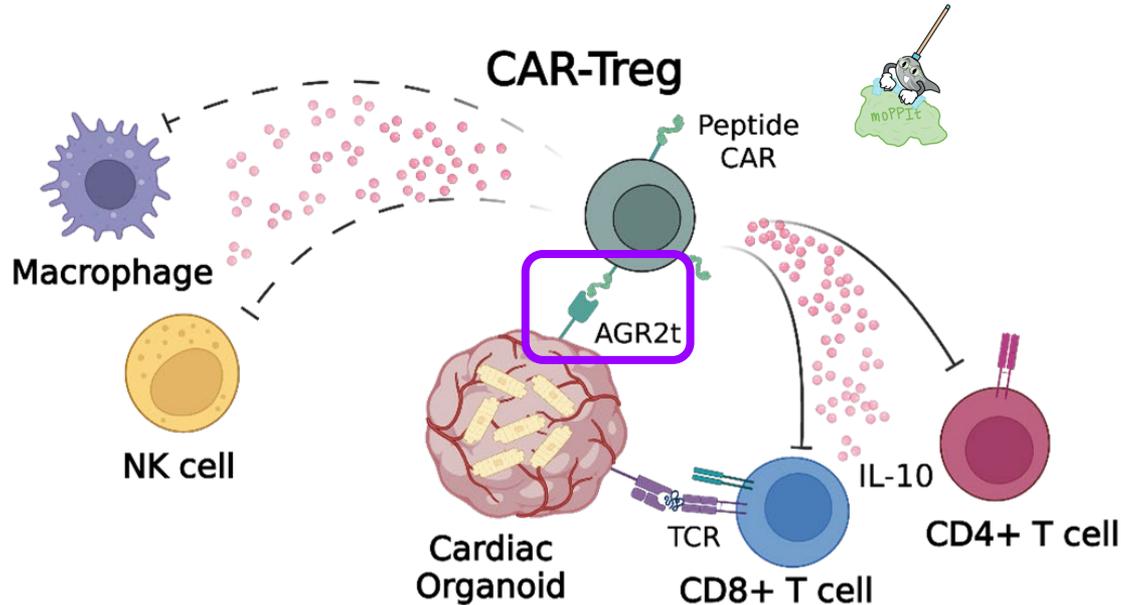


moPPIt-designed peptides as **CAR-T** ligands ?





moPPIt-designed peptides as **CAR-Treg** ligands ?



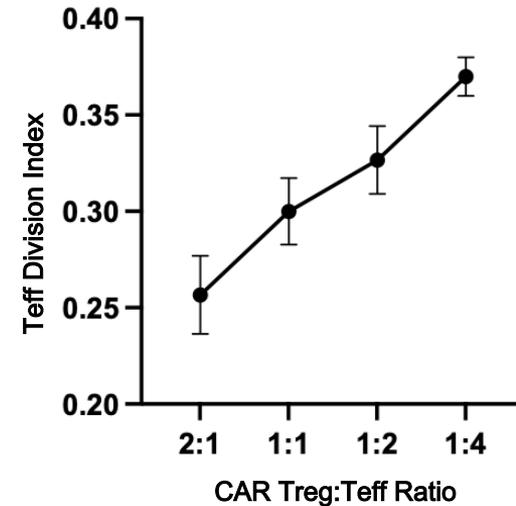
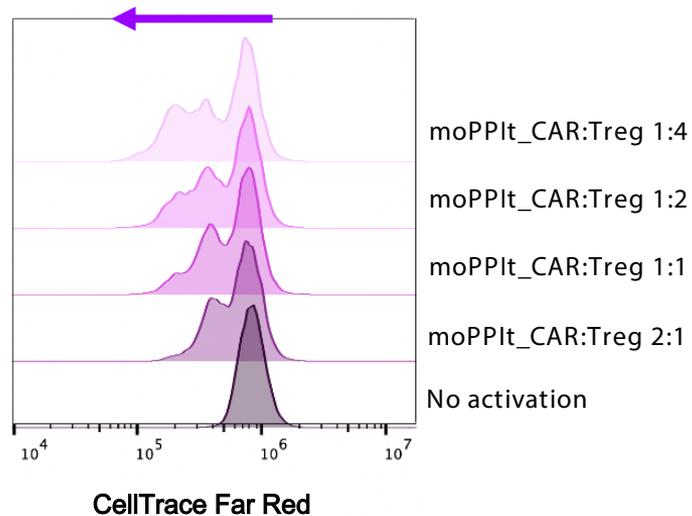


moPPIt-designed peptides → CARs

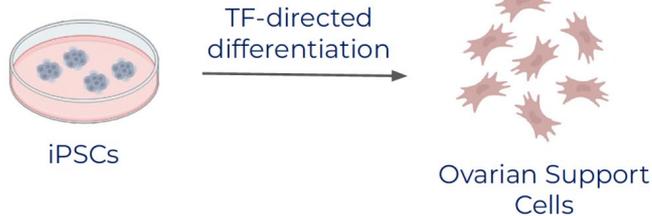




Suppress T cell proliferation !



Next: CAR Tregs to protect transplanted ovarian cells?



Ameno

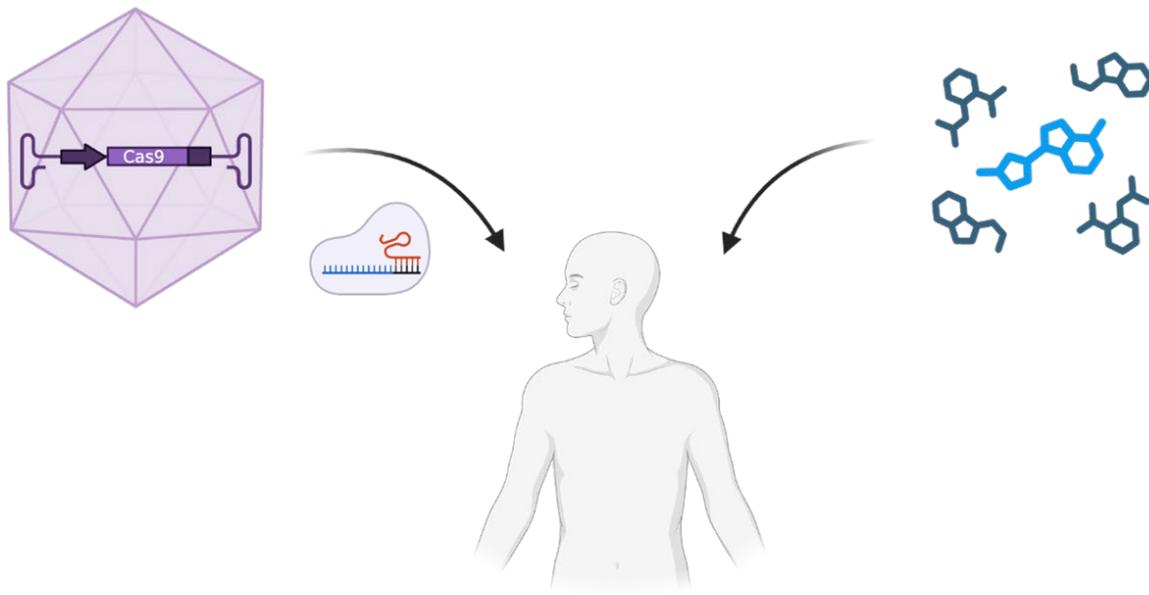
Utilizing our OSCs, we aim to develop a cell therapy that makes the onset of menopause slow and gradual, no longer an abrupt and inevitable milestone but something we can manage and modify with medical intervention.

Goals

- 📌 Integrate dynamically into the body's chemical conversation
- 🧠 Respond to signals from the brain and send hormonal feedback, mimicking the HPO axis
- 🎵 Tune the levels of hormones released to each individual's needs

We can also use DFM to *shrink* molecules!

We can also use DFM to *shrink* molecules!



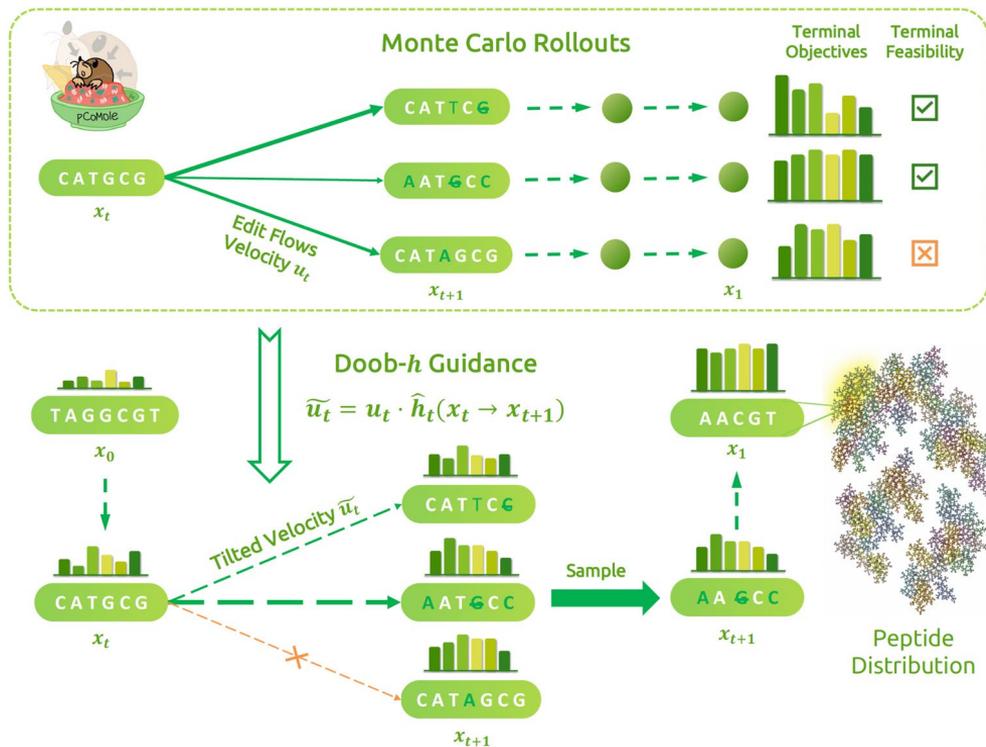


Pareto -Constrained Molecule Editing (pCoMole)





Pareto -Constrained Molecule Editing (pCoMole)





Peptidomimetics with *improved* therapeutic properties

Target	Ligand	Length	Non-Toxicity	Solubility	Permeability	Half-life (h)	Affinity	Motif Score	Specificity
PTH1R	Teriparatide	259	0.8890	0.6905	0.2793	2.8301	0.9148	0.2696	0.9834
	pCoMole	217	0.8729	0.7268	0.2959	2.9488	0.9358	0.3593	0.9706
p53	p28	170	0.8533	0.6737	0.2627	2.2015	0.6369	0.5418	0.9873
	pCoMole	133	0.8076	0.6309	0.3071	3.8517	0.6885	0.5014	0.9649
GLP-1R	Semaglutide	248	0.8872	0.7656	0.2583	1.8571	0.9467	0.4691	0.9827
	pCoMole	186	0.8122	0.7240	0.2985	4.0277	0.9367	0.5212	0.9778
	Liraglutide	237	0.7854	0.8307	0.2609	2.0808	0.8440	—	—
	Exenatide	256	0.8319	0.7547	0.2319	2.0663	0.8949	—	—
	Lixisenatide	286	0.7204	0.8392	0.2215	2.7535	0.8280	—	—
	Tirzepatide	273	0.9491	0.8303	0.2573	1.7149	0.9798	—	—

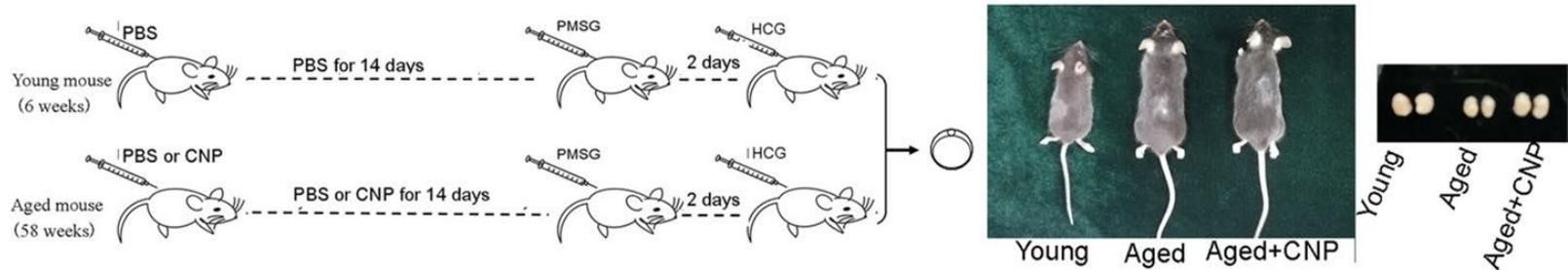
CNP peptide as a fertility treatment?



C-type natriuretic peptide improves maternally aged oocytes quality by inhibiting excessive PINK1/Parkin-mediated mitophagy

Hui Zhang^{1,2†}, Chan Li^{1,2†}, Qingyang Liu^{1,2}, Jingmei Li^{1,2}, Hao Wu^{1,2}, Rui Xu^{1,2}, Yidan Sun^{1,2}, Ming Cheng^{1,2}, Xiaoe Zhao^{1,2}, Menghao Pan^{1,2}, Qiang Wei^{1,2*}, Baohua Ma^{1,2*}

¹College of Veterinary Medicine, Northwest A&F University, Yangling, China; ²Key Laboratory of Animal Biotechnology, Ministry of Agriculture, Yangling, China



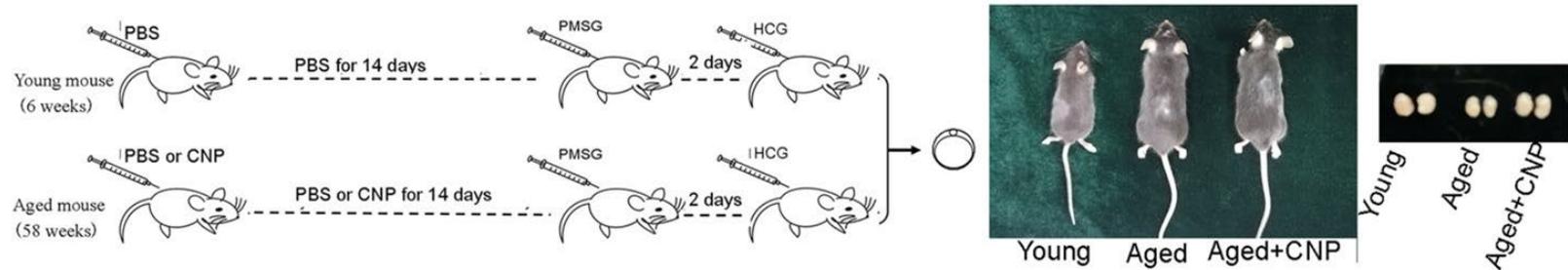
CNP peptides have very low serum half-life/stability (<2 hours)



C-type natriuretic peptide improves maternally aged oocytes quality by inhibiting excessive PINK1/Parkin-mediated mitophagy

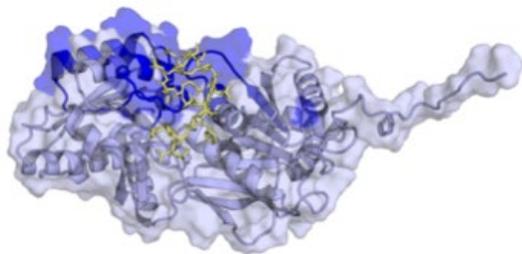
Hui Zhang^{1,2†}, Chan Li^{1,2†}, Qingyang Liu^{1,2}, Jingmei Li^{1,2}, Hao Wu^{1,2}, Rui Xu^{1,2}, Yidan Sun^{1,2}, Ming Cheng^{1,2}, Xiaoe Zhao^{1,2}, Menghao Pan^{1,2}, Qiang Wei^{1,2*}, Baohua Ma^{1,2*}

¹College of Veterinary Medicine, Northwest A&F University, Yangling, China; ²Key Laboratory of Animal Biotechnology, Ministry of Agriculture, Yangling, China





pCoMole-generated CNP analogs with favorable therapeutic profiles

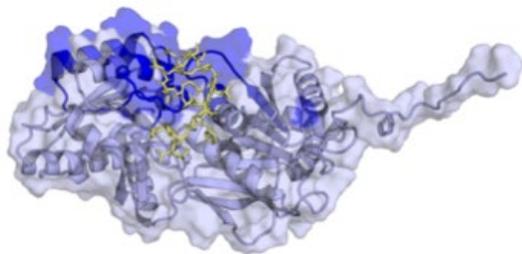


Known CNP

Toxicity: 0.87	Solubility: 0.88
Half-Life: 1.74 h	Permeability: 0.32
Affinity: 0.80	VINA Score: -6.2
Specificity: 0.99	Motif Score: 0.57

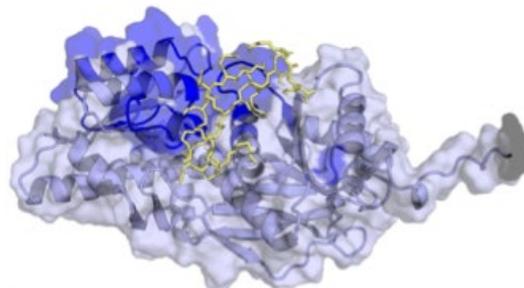


pCoMole-generated CNP analogs with favorable therapeutic profiles



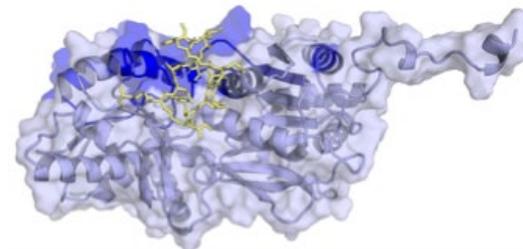
Known CNP

Toxicity: 0.87 Solubility: 0.88
Half-Life: 1.74 h Permeability: 0.32
Affinity: 0.80 VINA Score: -6.2
Specificity: 0.99 Motif Score: 0.57



pCoMole-designed CNP variant 1

Toxicity: 0.84 Solubility: 0.82
Half-Life: 5.17 h Permeability: 0.31
Affinity: 0.81 VINA Score: -5.2
Specificity: 0.98 Motif Score: 0.58



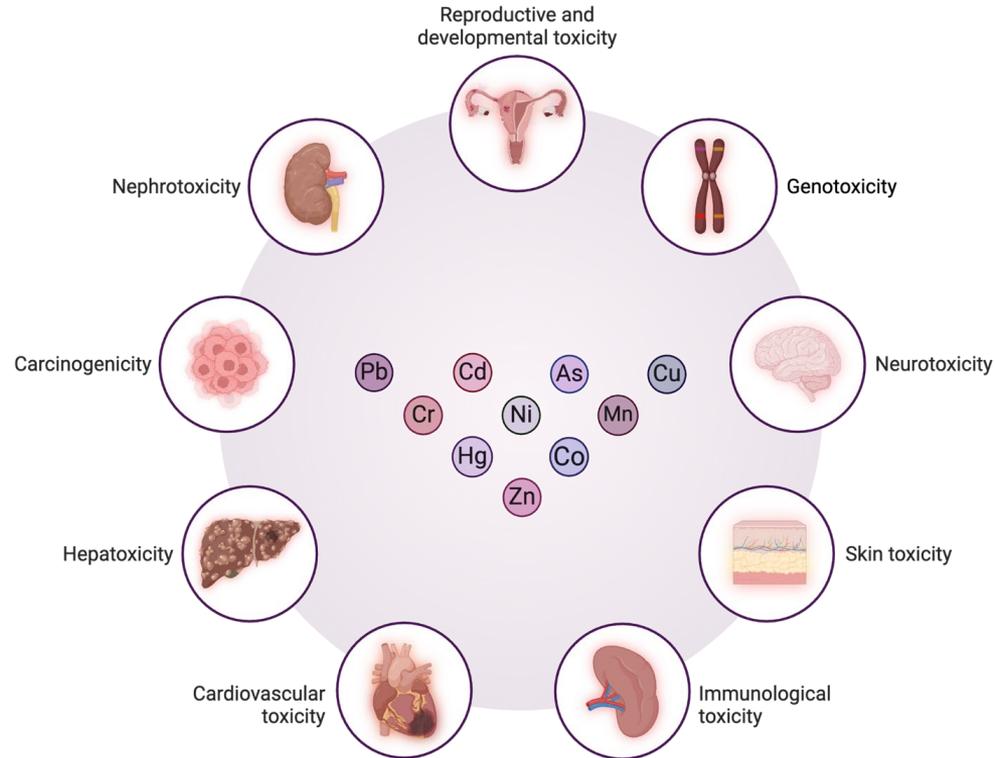
pCoMole-designed CNP variant 2

Toxicity: 0.77 Solubility: 0.89
Half-Life: 3.49 h Permeability: 0.32
Affinity: 0.79 VINA Score: -7.3
Specificity: 1.00 Motif Score: 0.60

Can we specifically bind to non -biological substrates?



Heavy metals pollute our environment

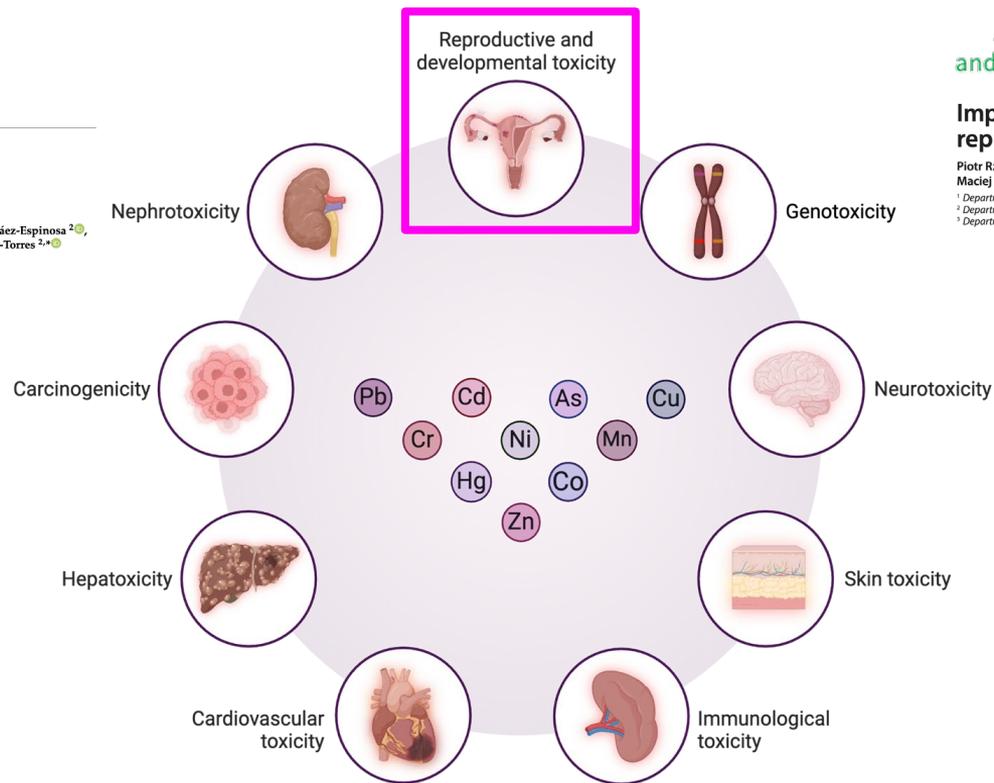


Heavy metals pollute our environment and create reproductive health risks!

 antioxidants

Review
Impact of Heavy Metals on Human Male Fertility—An Overview

Andrea López-Botella ^{1,2,4}, Irene Velasco ^{1,2,4}, Maribel Ación ^{1,3}, Paula Sáez-Espinosa ², José-Luis Todolí-Torró ⁴, Raquel Sánchez-Romero ⁴ and María José Gómez-Torres ^{2,4}



Annals of Agricultural and Environmental Medicine

Impact of heavy metals on the female reproductive system

Piotr Rzymiski¹, Katarzyna Tomczyk², Paweł Rzymiski², Barbara Poniedziałek¹, Tomasz Opala¹, Maciej Wilczak³

¹ Department of Biology and Environmental Protection, Poznań University of Medical Sciences, Poznań, Poland

² Department of Mother's and Child's Health, Poznań University of Medical Sciences, Poznań, Poland

³ Department of Educational Medicine, Poznań University of Medical Sciences, Poznań, Poland



Metalorian = Heavy Metal-Binding Peptides





MetaLATTE = Protein-Metal Multi-Class Classifier

Protein Sequence Input

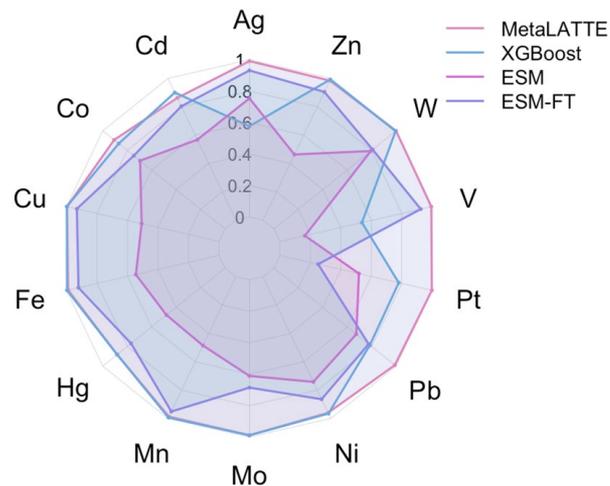
G Q D T D C R E C E S

One-hot Label Input

Cd	Cu	Co	None
[1 , 0 , 0 , . . . 0]			

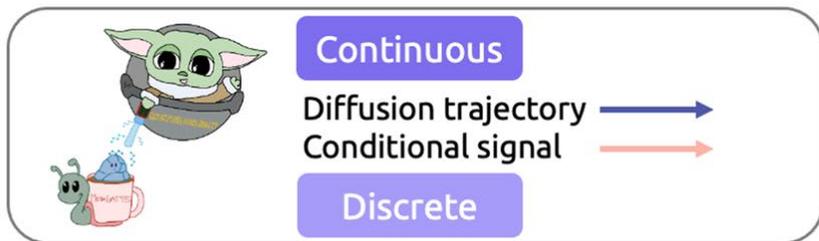
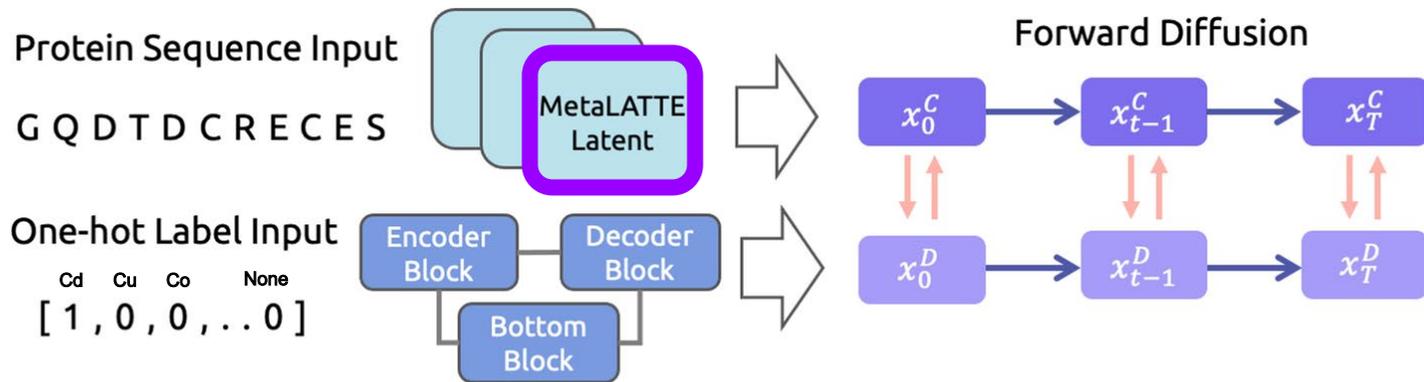


AUCROC Comparison



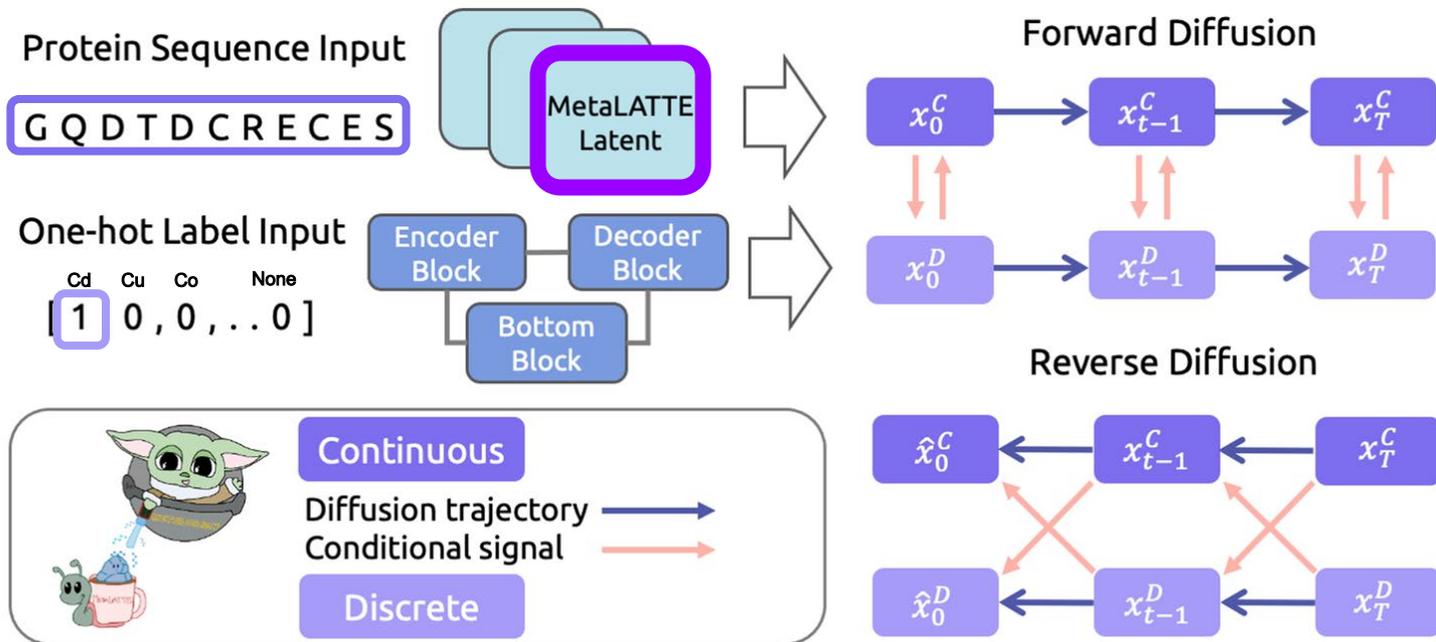


Metalorian = Co-evolving diffusion of peptide binders



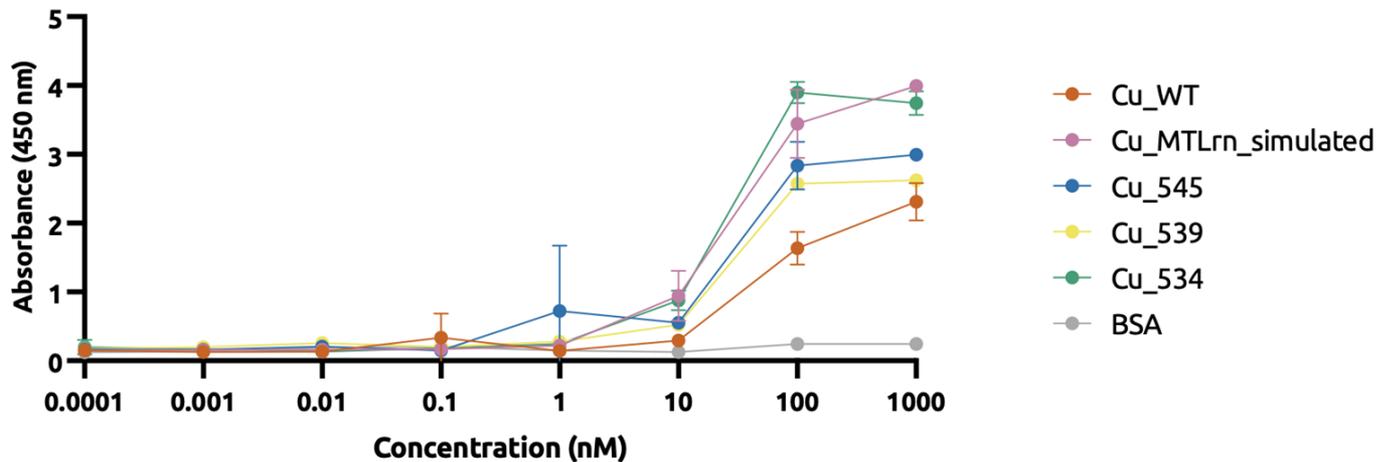


Metalorian = Co-evolving diffusion of peptide binders

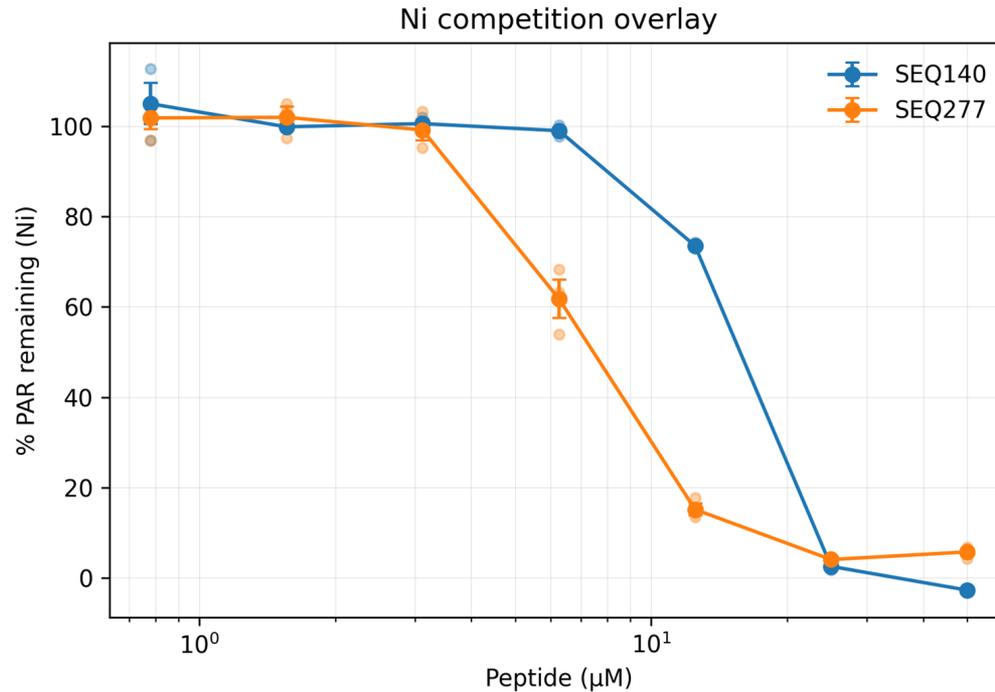




Metalorian peptides **bind Copper !**



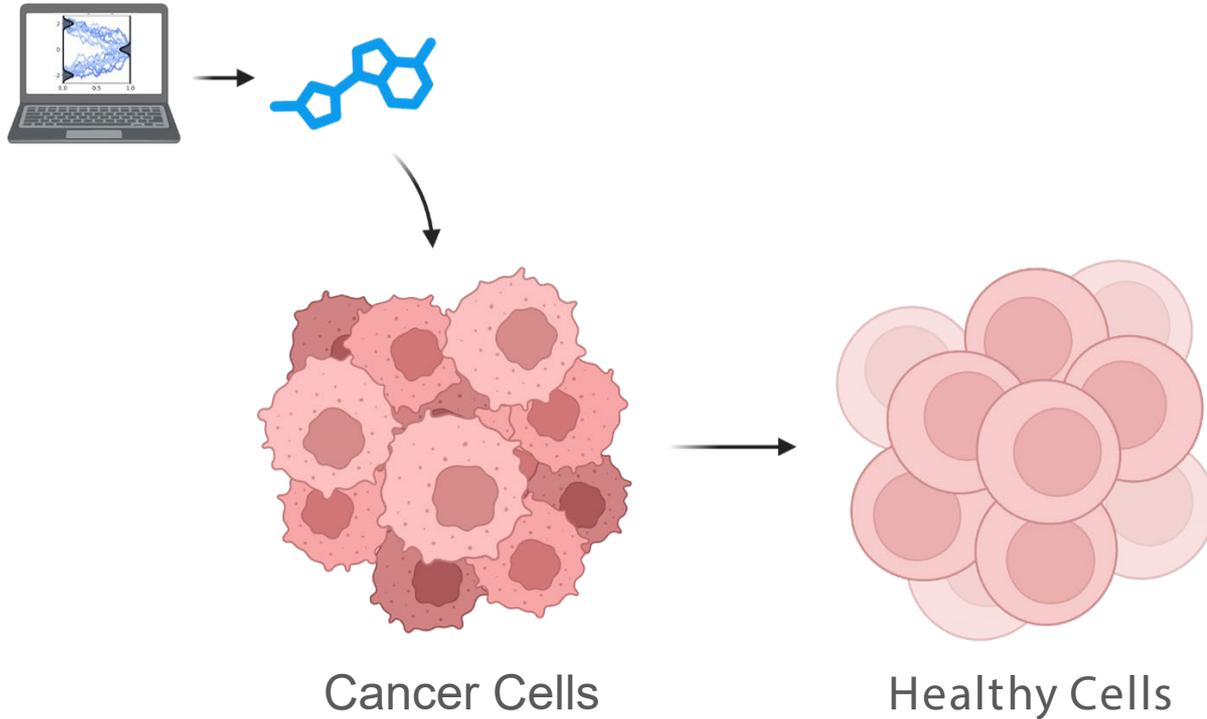
Metalorian peptides can sequester nickel



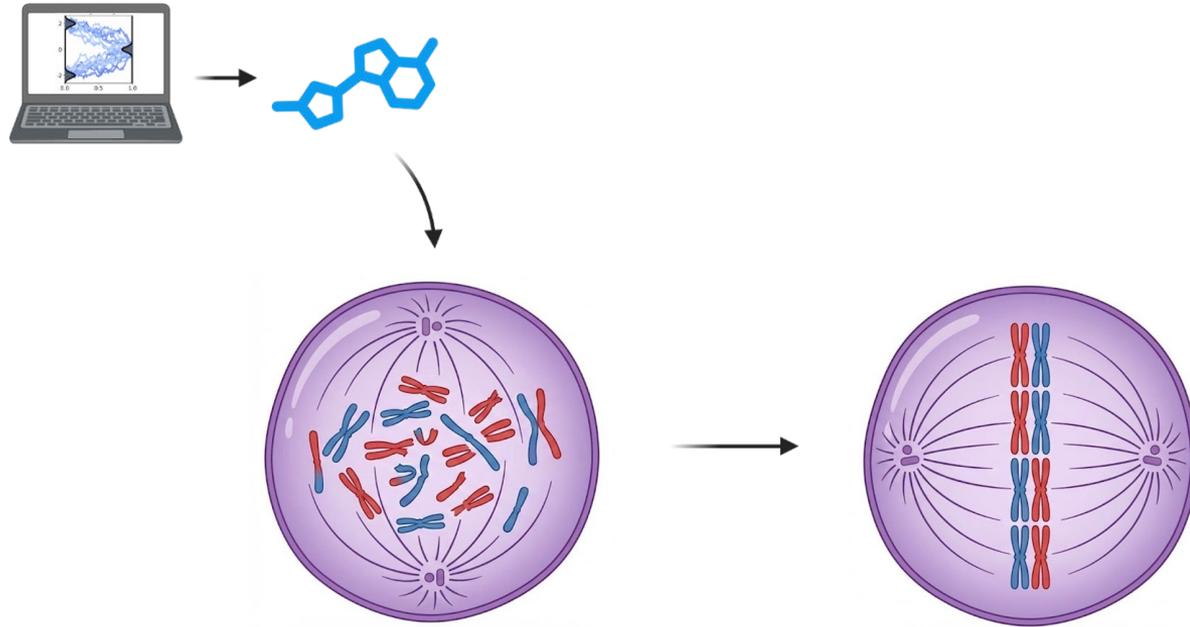
Ultimate goal: Instead of just doing targeted therapeutics...



Ultimate goal: Can we induce desired states?



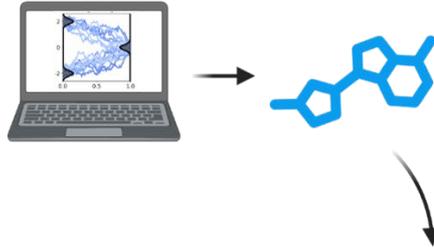
Ultimate goal: Can we induce desired states?



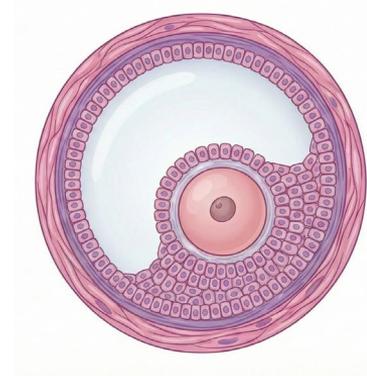
Aneuploid -Prone State

Chromosomally-Stable State

Ultimate goal: Can we induce desired states?

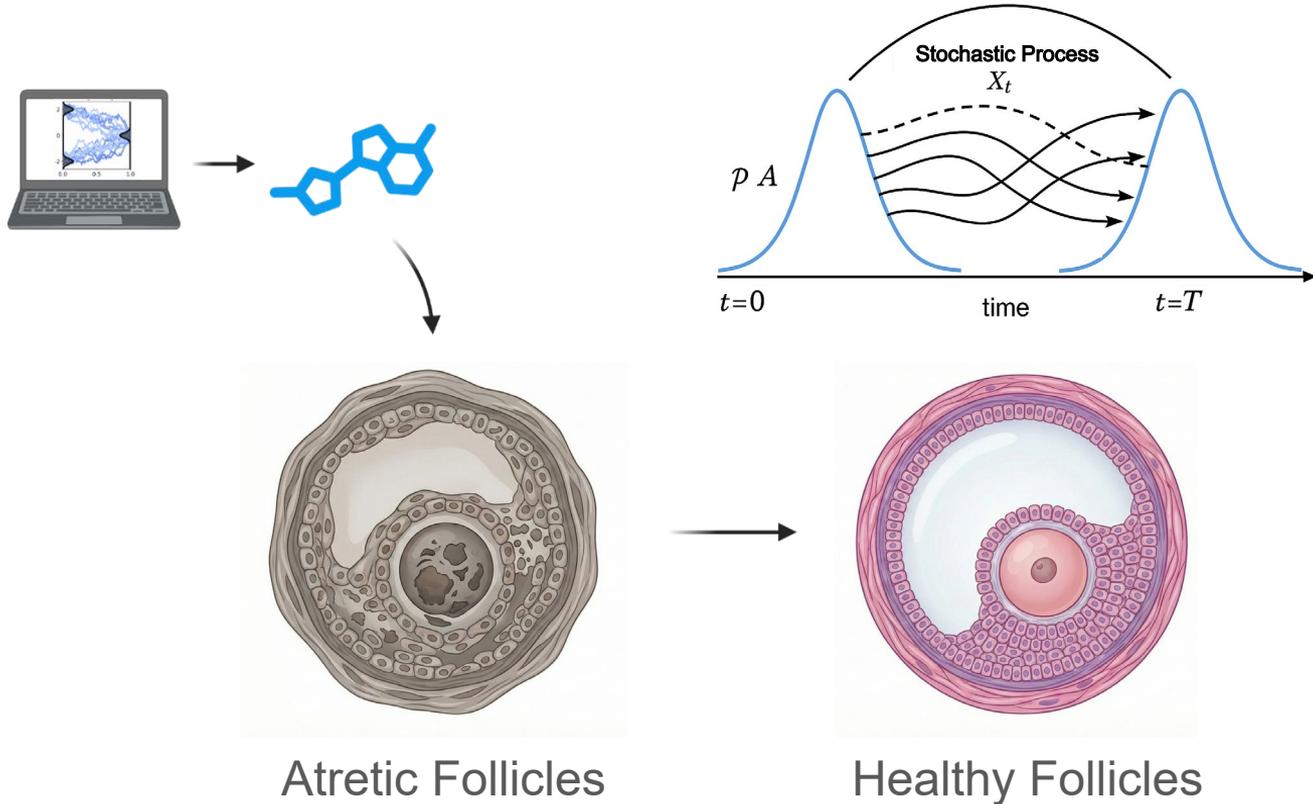


Atretic Follicles

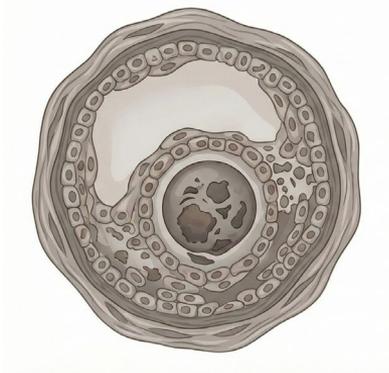
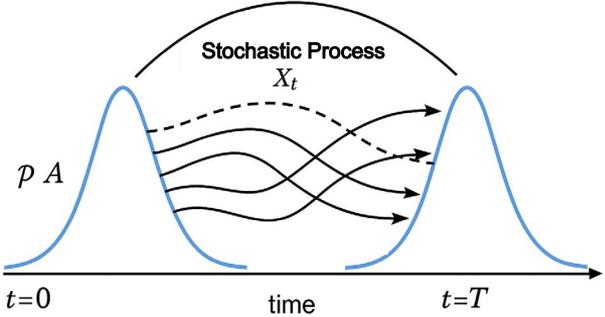
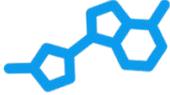
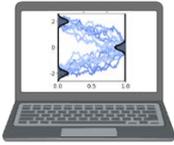


Healthy Follicles

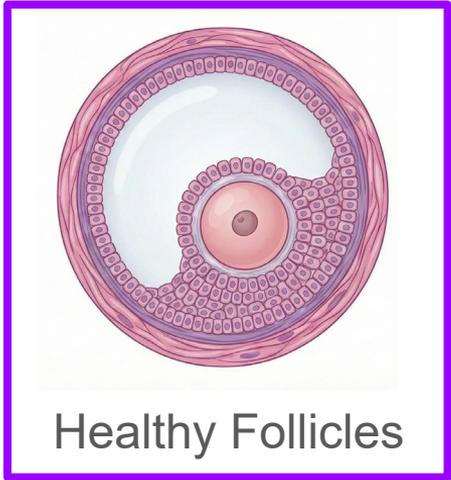
We frame this problem as a **Schrödinger Bridge**



But these end cell states are heterogenous!



Atretic Follicles



Healthy Follicles

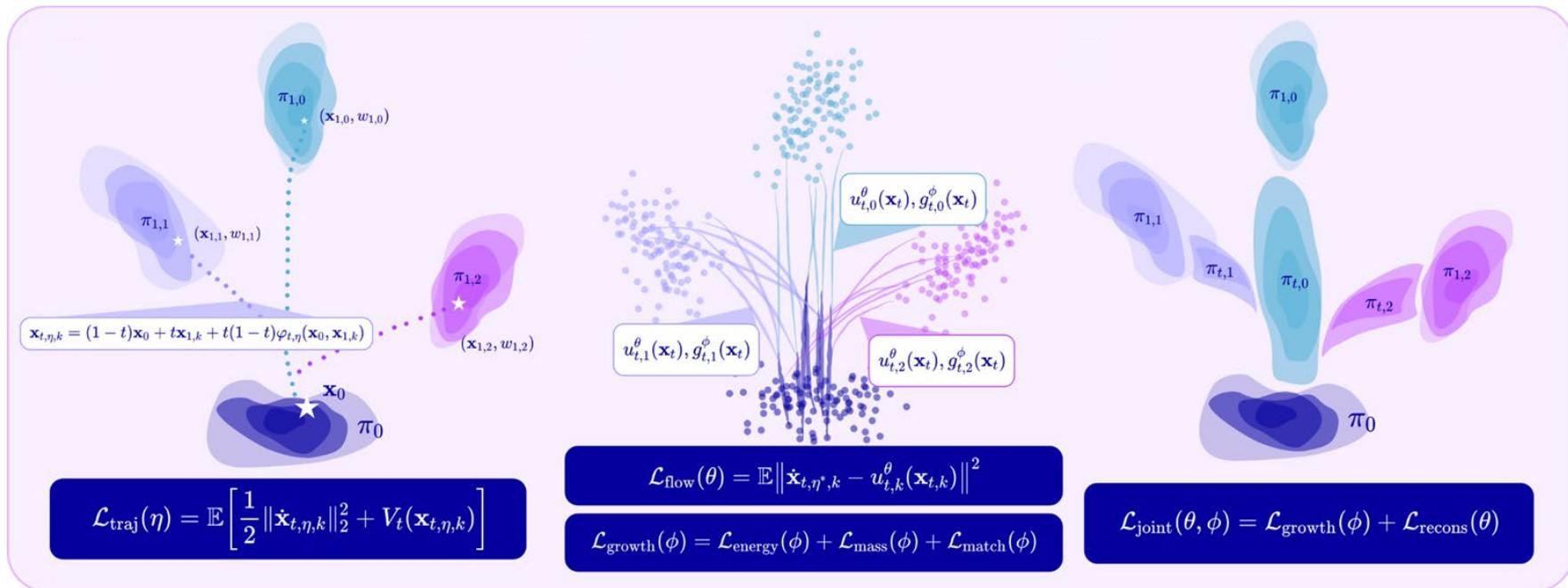


Branched Schrödinger Bridge Matching (**BranchSBM**)



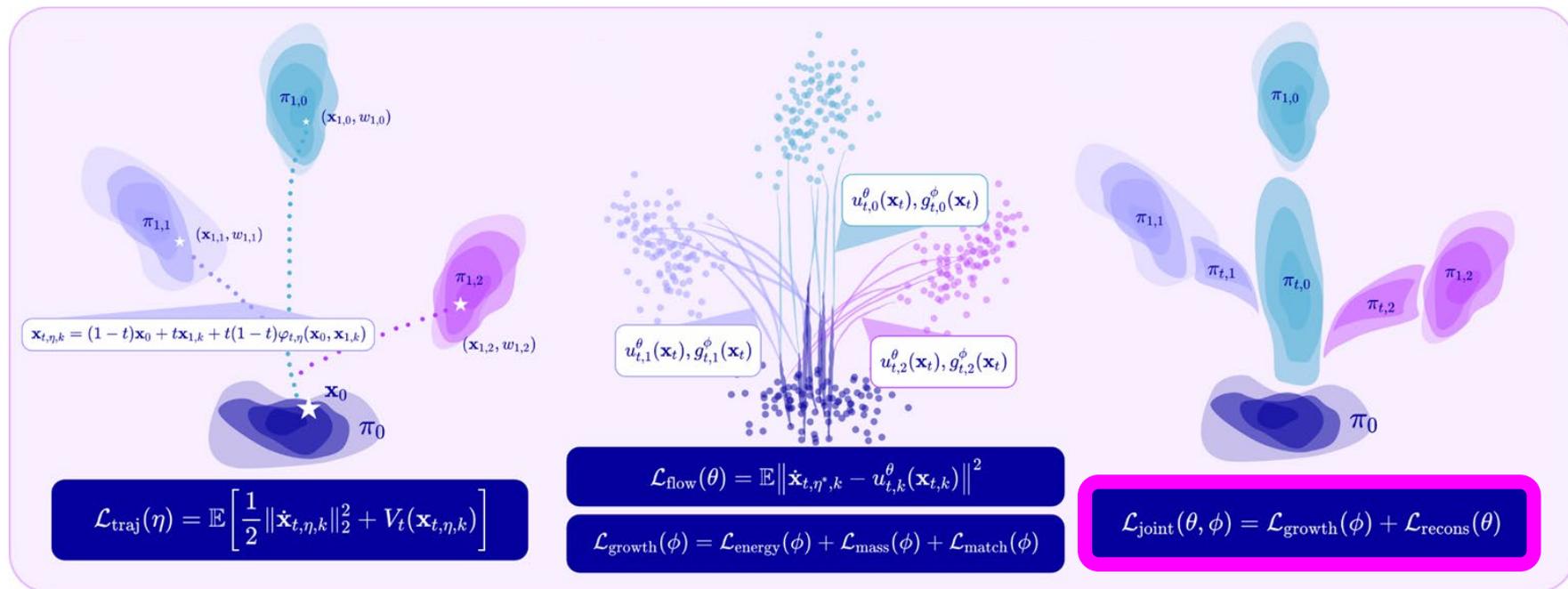


BranchSBM learns multiple coupled bridges



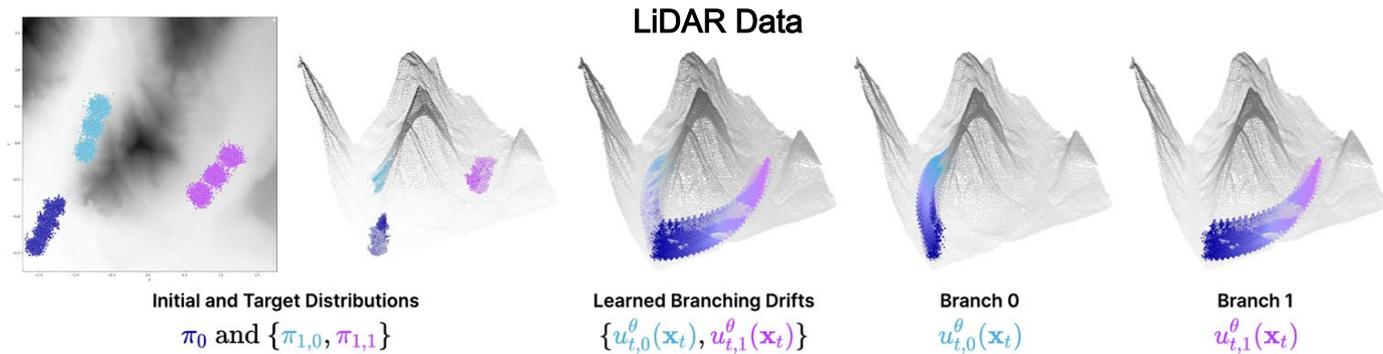


BranchSBM learns multiple coupled bridges, and *jointly* optimizes them



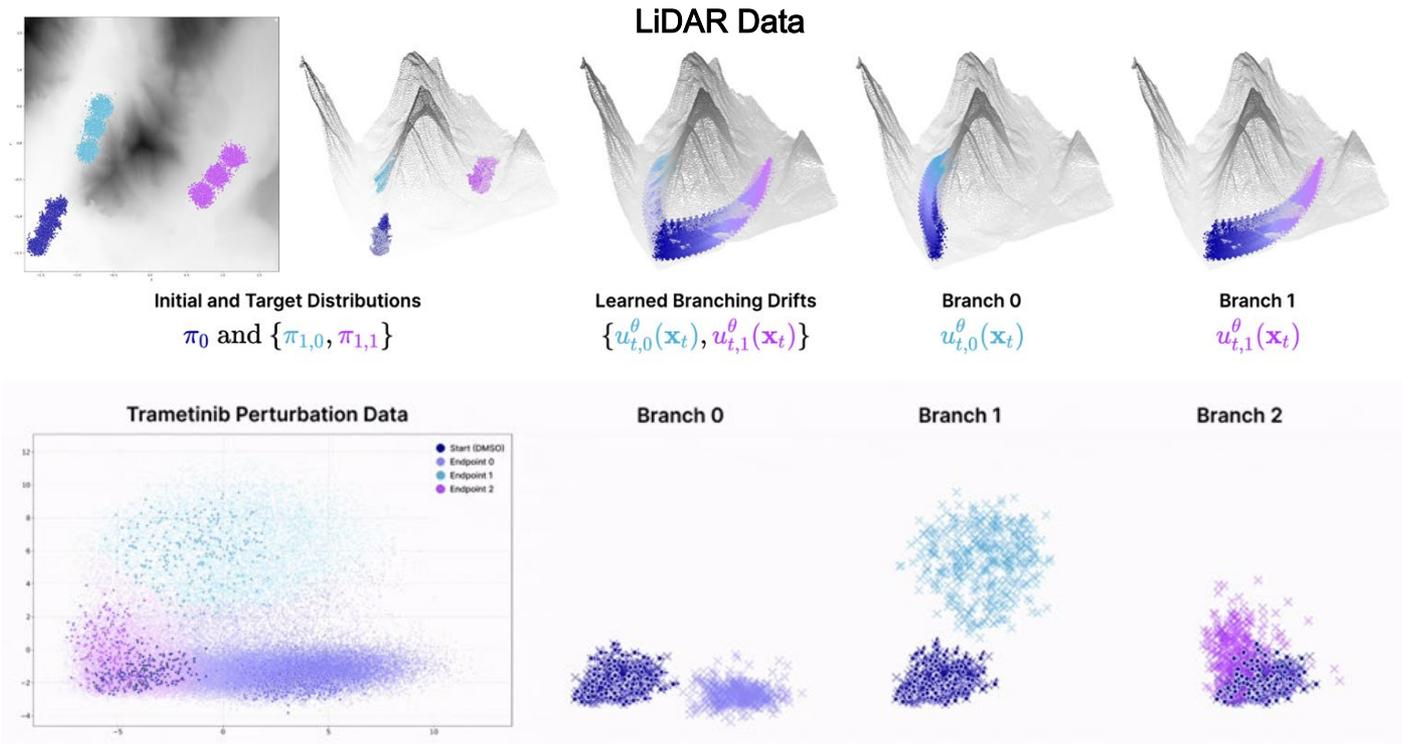


BranchSBM \rightarrow trajectories to heterogeneous terminal states





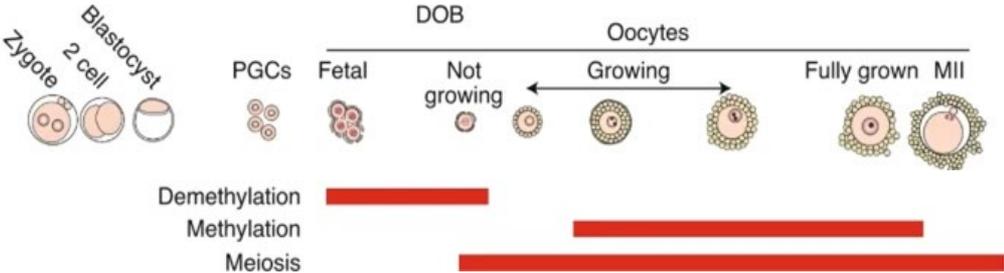
BranchSBM \rightarrow trajectories to heterogeneous terminal states



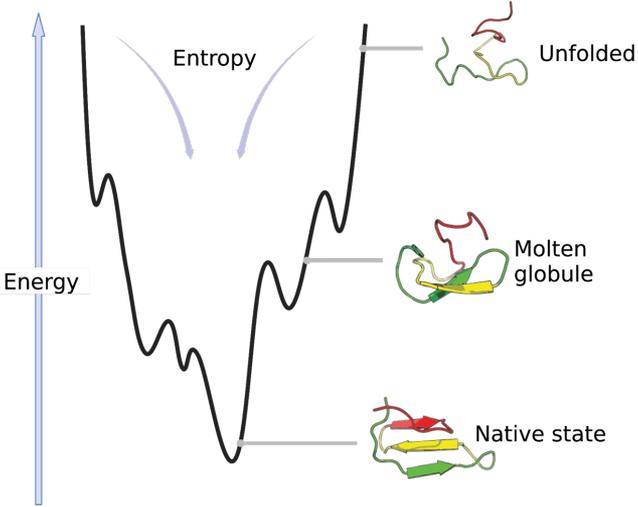
But in biology, transition states **ALSO** matter.

But in biology, transition states **ALSO** matter.

Cell Differentiation

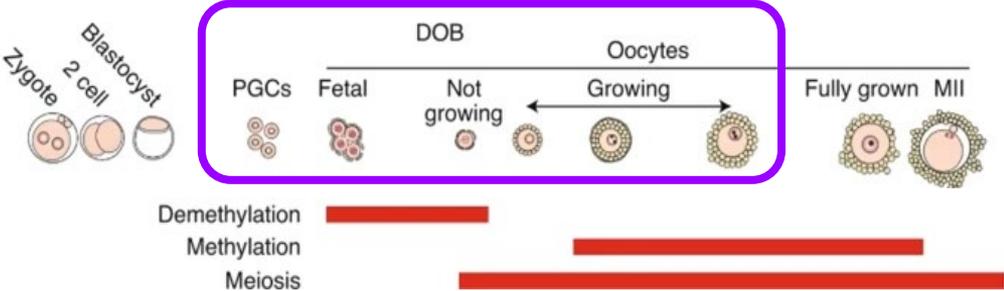


Protein Folding

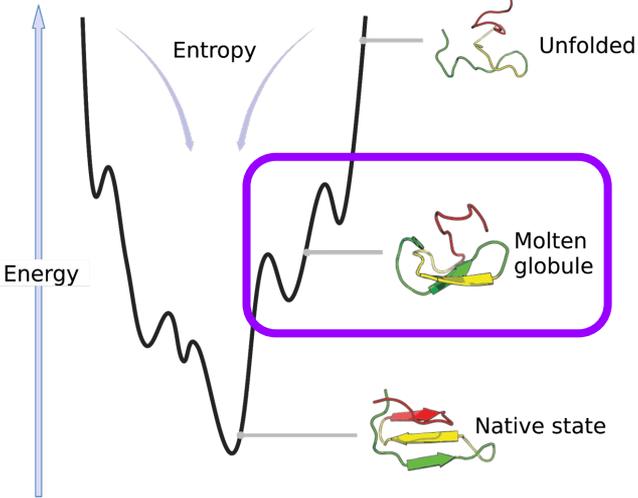


But in biology, transition states **ALSO** matter.

Cell Differentiation

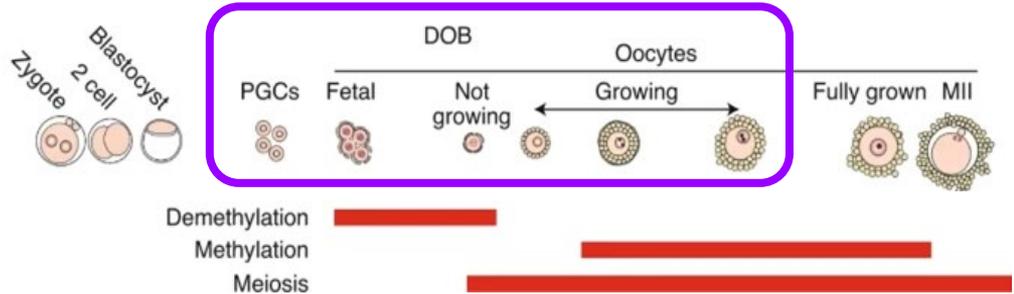


Protein Folding

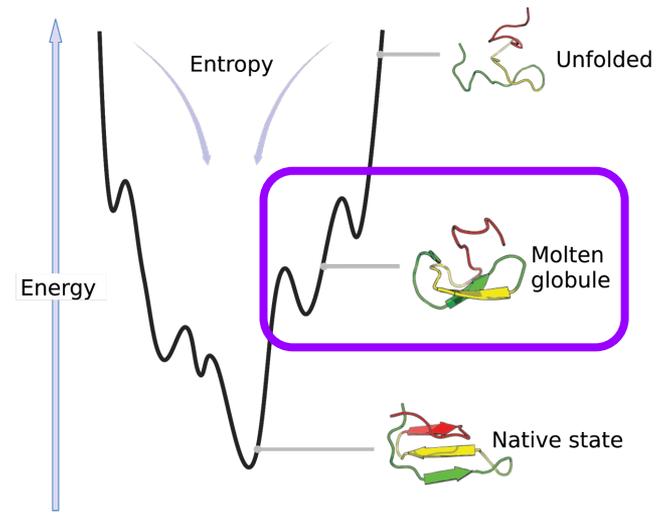


We need to model intermediate dynamics too!

Cell Differentiation

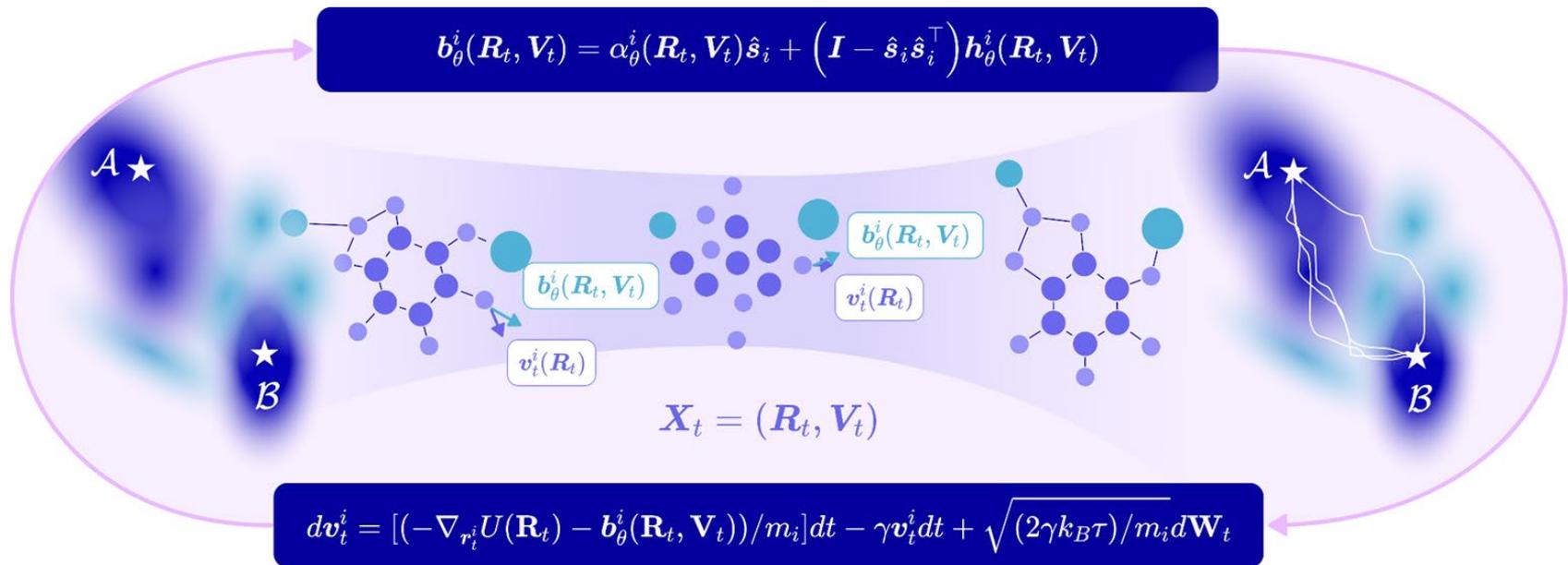


Protein Folding



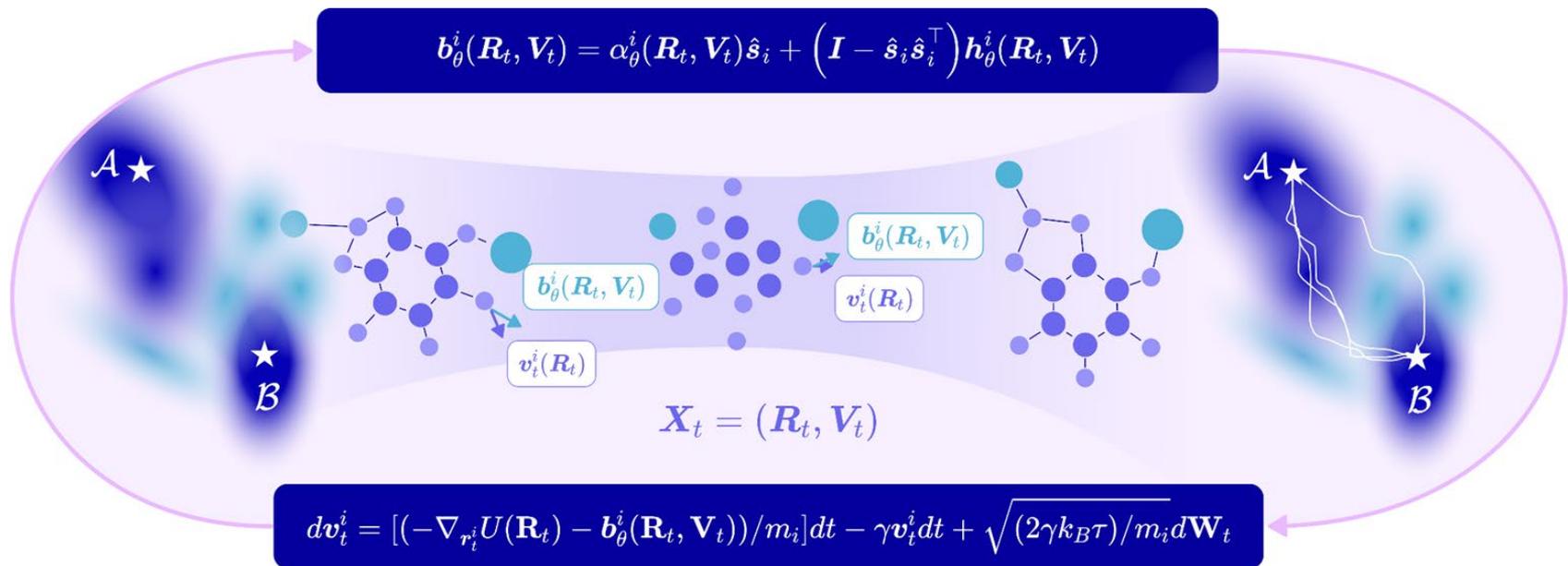


Entangled Schrödinger Bridge Matching



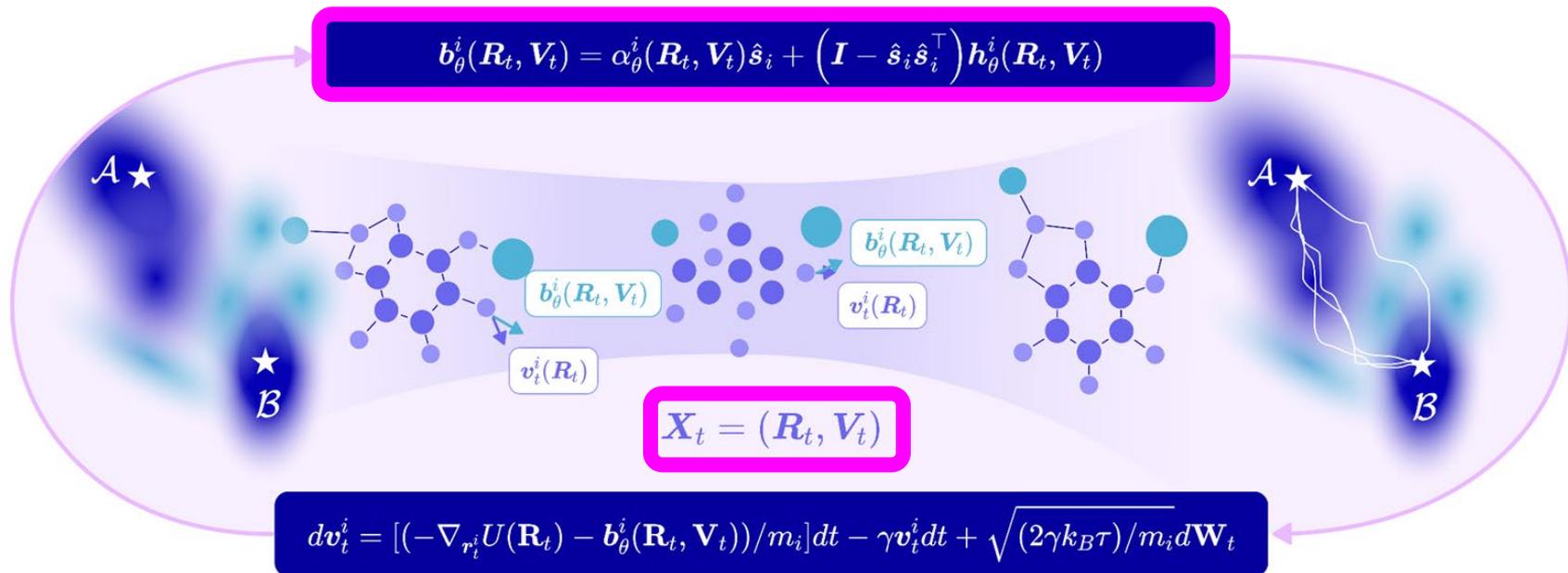


Entangled Schrödinger Bridge Matching (EntangledSBM)



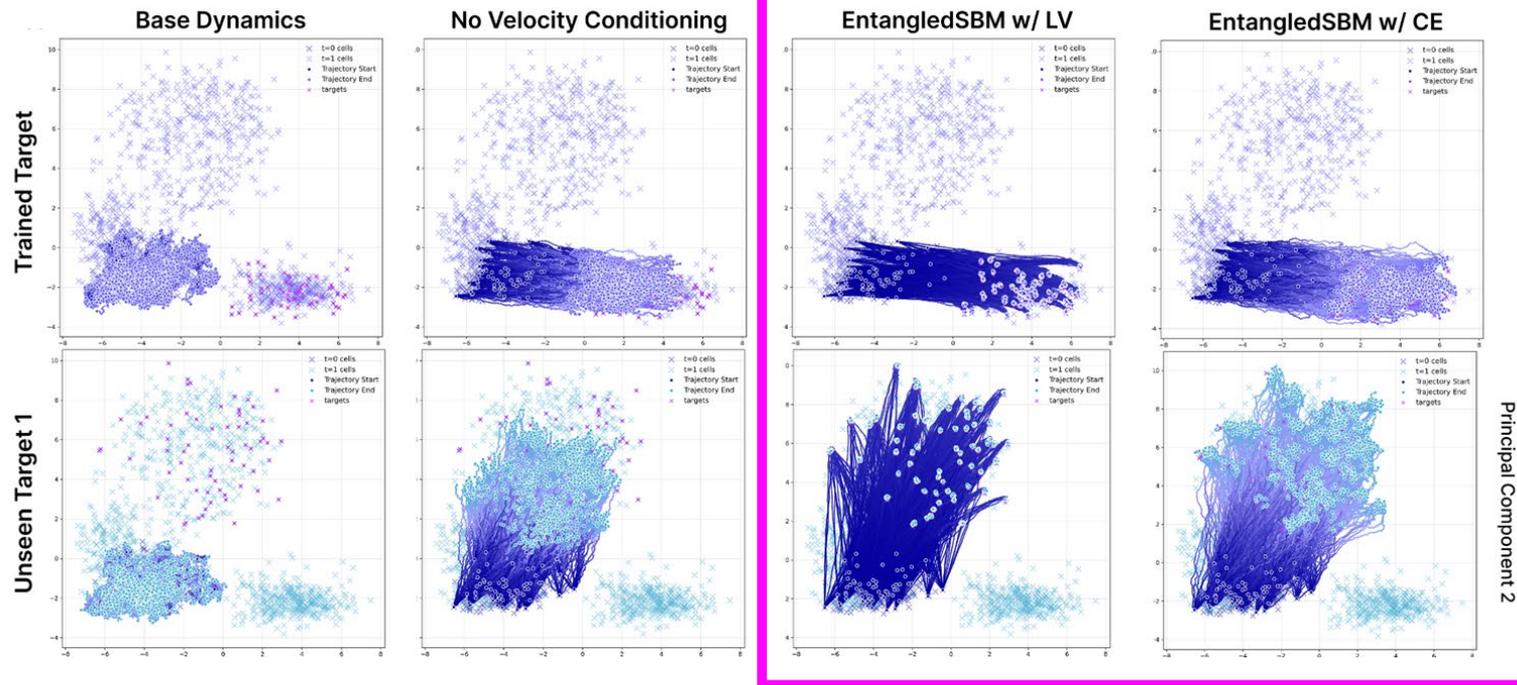


Learn a bias force captures the dynamic interactions between particles





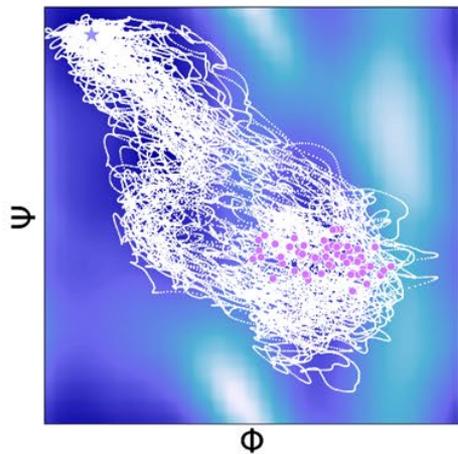
EntangledSBM predicts intermediate cell populations



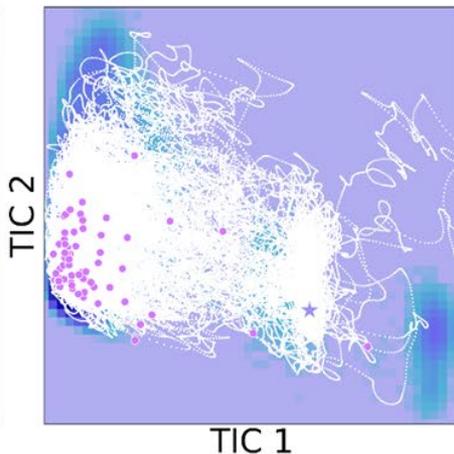


Feasible transition paths for protein folding

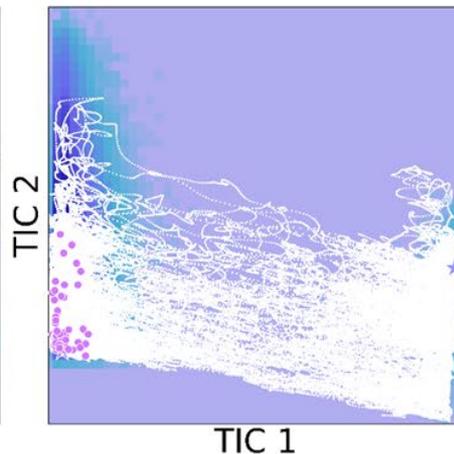
Alanine Dipeptide (22 atoms)



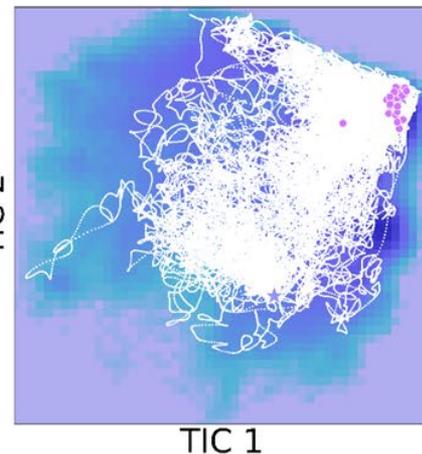
Chignolin (138 atoms)



Trpcage (284 atoms)

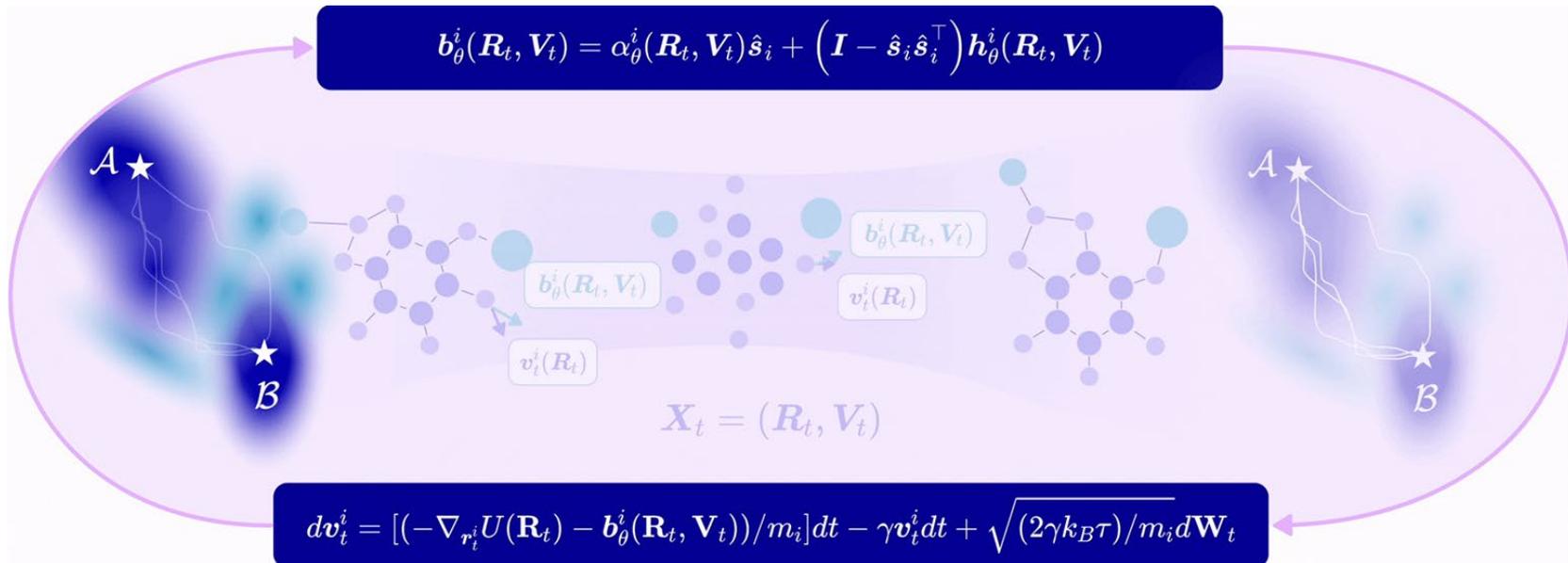


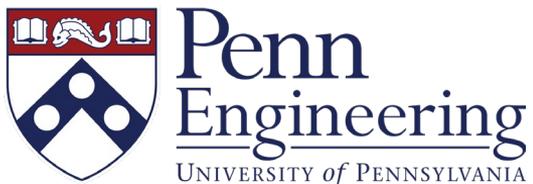
BBA (504 atoms)





A unified model for protein dynamics and cellular trajectories





Thank you to the best students in the world!

Computational Team



Liz Yinuo Sophie Tong



Aastha Rosie Sophia Shrey

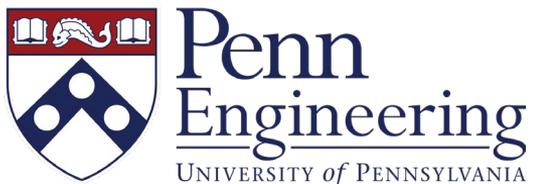
Experimental Team



Lin Zach Lauren Tian Divya



Howard Yesol Sumi Jiale



To our generous funders!

Computational Team



Liz



Yinuo



Sophie



Tong



Aastha



Rosie



Sophia



Shrey



Experimental Team



Lin



Zach



Lauren



Tian



Divya



Howard



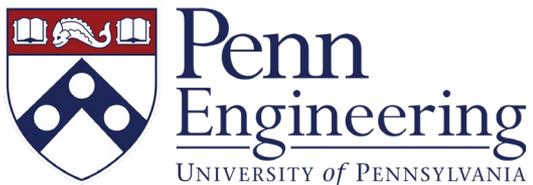
Yesol



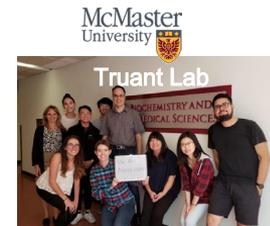
Sumi

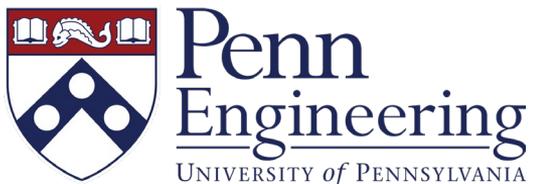


Jiale



To our supportive collaborators!





And to you for listening to our stories!

Computational Team



Liz



Yinuo



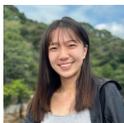
Sophie



Tong



Aastha



Rosie



Sophia



Shrey



Experimental Team



Lin



Zach



Lauren



Tian



Divya



Howard



Yesol



Sumi



Jiale



We hope you will follow our work!

Computational Team



Liz



Yinuo



Sophie



Tong



Aastha



Rosie



Sophia



Shrey



Experimental Team



Lin



Zach



Lauren



Tian



Divya



Howard



Yesol



Sumi



Jiale