

absci.

```
from absci import de_novo_model
model = de_novo_model.load_latest()
antigen = model.load_pdb("7olz.pdb",
chain="A")
antibodies = model.predict(antigen, N=300000)
```

```
from absci_library import codon_optimizer
library
= codon_optimizer.reverse_translate(library)
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```

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from absci import lead_opt_model
lead_optimizer = lead_opt_model.load_latest()
library.naturalness =
lead_optimizer.naturalness(library)
lead_optimizer.optimize(library).to_wet_lab(assay="SPR")
```

# GENERATIVE AI DRUG CREATION



ABSCI R&D DAY 2024

```
from absci import genetic_algorithm; parameters=["maximize|binding_affinity:pH=7.5", "minimize|binding_affinity:pH=6.0",
"maximize|human_naturalness"]; library = genetic_algorithm.multiparametric_optimization(library, parameters, evolutions=100);
library.to_wet_lab(assays=["ACE", "SPR", "Bioassays"])
```

```
from absci_library import codon_optimizer
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# ALEX KHAN

VP, FINANCE & INVESTOR RELATIONS

```
library = []
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```

# Disclaimers




## Forward-Looking Statements

Certain statements in this presentation that are not historical facts are considered forward-looking within the meaning of Section 27A of the Securities Act of 1933, as amended, and Section 21E of the Securities Exchange Act of 1934, as amended, including statements containing the words “will,” “may,” “anticipates,” “plans,” “believes,” “forecast,” “estimates,” “expects,” “predicts,” “advancing,” “aim,” and “intends,” or similar expressions. We intend these forward-looking statements, including statements regarding our strategy, our expectations regarding the clinical, therapeutic and market potential of product candidates discovered and developed through our platform; the potential advantages of our technology and the assets in our internal pipeline; our ability to achieve catalysts in our preclinical and clinical development programs, such as the initiation of IND-enabling studies and Phase 1 clinical development and the receipt of clinical data; the anticipated timing of such events; the expected evolution of our portfolio over time; guidance regarding cash, cash equivalents and our projected cash runway, our future operations, internal research and technological development activities, estimated speed and cost advantages of leveraging our AI drug creation platform; our expectations regarding the status and progress of our existing partnerships and our plans for potential new partnerships; our expected operational efficiencies, research and technology development collaboration efforts, growth plans, prospects, plans and objectives of management, to be covered by the safe harbor provisions for forward-looking statements contained in Section 27A of the Securities Act and Section 21E of the Securities Exchange Act, and we make this statement for purposes of complying with those safe harbor provisions. These forward-looking statements reflect our current views about our plans, intentions, expectations, strategies, and prospects, which are based on the information currently available to us and on assumptions we have made. We can give no assurance that the plans, intentions, expectations, or strategies will be attained or achieved, and, furthermore, actual results may differ materially from those described in the forward-looking statements and will be affected by a variety of risks and factors that are beyond our control, including, without limitation, risks and uncertainties relating to the development of our technology as well as the assets in our internal pipeline, our ability to secure milestone payments and royalties, and our ability to effectively conduct research, drug discovery and development activities with respect to our internal programs and to collaborate with our partners or potential partners with respect to their research, drug discovery and development activities; along with those risks set forth in our most recent periodic report filed with the U.S. Securities and Exchange Commission, as well as discussions of potential risks, uncertainties, and other important factors in our subsequent filings with the U.S. Securities and Exchange Commission. Except as required by law, we assume no obligation to update publicly any forward-looking statements, whether as a result of new information, future events, or otherwise.

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# Agenda

8:00 - 9:00

## Breakfast

### Opening remarks

Prof Sir Mene Pangalos | Board Director and Co-Chair of Scientific Advisory Board, Absci

### Corporate overview

Sean McClain | Founder & CEO, Absci

### Innovation overview

Andreas Busch, PhD | Chief Innovation Officer, Absci

### AI platform updates

Amaro Taylor-Weiner, PhD | Chief AI Officer, Absci

## Break

### Pipeline updates

Christian Stegmann, PhD | SVP Drug Creation, Absci

#### *Guest Presenters:*

Anthony Rossi, MD | MSKCC; Weill Cornell Medical College; Advisor, Absci

Mike Jafar | Advisor, BCG and Advisor, Absci

Luis Diaz, MD | MSKCC; Advisor, Absci

Dennis Slamon, MD, PhD | Chief of Division of Hematology and Oncology, UCLA Medicine

### Partnership and Business updates

Zach Jonasson, PhD | CFO & CBO, Absci

#### *Guest Presenters:*

Karl Ziegelbauer, PhD | Chief Scientific Officer, Almirall

### Closing remarks

Sean McClain | Founder & CEO, Absci

### Q&A

12:15 - 1:00

### Reception

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# SIR MENE PANGALOS, PHD

BOARD DIRECTOR, ABSCI  
CO-CHAIR, ABSCI SCIENTIFIC ADVISORY BOARD

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abs ci



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# SEAN MCCLAIN

FOUNDER & CEO

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absci®

Absci is a data-first  
generative AI Drug  
Creation™ company





absci<sup>®</sup>

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generative AI Drug  
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## PLATFORM

Leading *de novo* antibody design models to unlock novel biology and create differentiated therapeutics

## PARTNERSHIPS

Successful execution on Large Pharma partnerships such as AstraZeneca & Almirall

New Partnerships announced with Memorial Sloan Kettering Cancer Center and Twist Bioscience

## PIPELINE

Rapidly advancing and expanding pipeline of differentiated assets designed using Absci's generative AI platform

- ABS-101: “Best-in-class” potential TL1A antibody entering clinic 1H 2025
- ABS-201: Novel hair re-growth prolactin receptor antibody addressing significant clinical and commercial need
- ABS-301 & ABS-501: showcasing novel and differentiated programs designed using AI



Absci has been  
developing AI  
antibody design  
tools to:

➤ ADDRESS COMPLEX AND PREVIOUSLY  
“HARD TO DRUG” TARGETS

- Bind specific extracellular domains
- Target specific conformations

➤ INTRODUCE PRECISE CONTROL OVER  
ANTIBODY DESIGN:

- “Smart” biologics
- Engineer selectivity, minimizing off target toxicity
- Agonism vs antagonism
- Multidimensionally co-optimized

## Ingredients for Success

### LEADING AI MODELS

Leading de novo AI model for antibody design with proof-points in internal and partnered programs

### COMPUTE AT SCALE

Compute at scale enabled by partnerships with NVIDIA & Oracle

### DATA ADVANTAGE

Proprietary ultra-high throughput data generation in 77,000+ ft<sup>2</sup> lab  
Amassing high quality data at scale since 2020

### DRUG DISCOVERY EXPERTISE

World-Class Discovery Team  
>10 Drugs Approved under current leadership

**Absci's  
leadership in  
AI de novo  
antibody  
design**

# Our AI platforms are enabled by our 6-week 'lab-in-the loop' active learning cycles

## AI PLATFORMS

### DE NOVO ANTIBODY DESIGN

*de novo* design of epitope-specific antibodies against targets without requiring a known binder

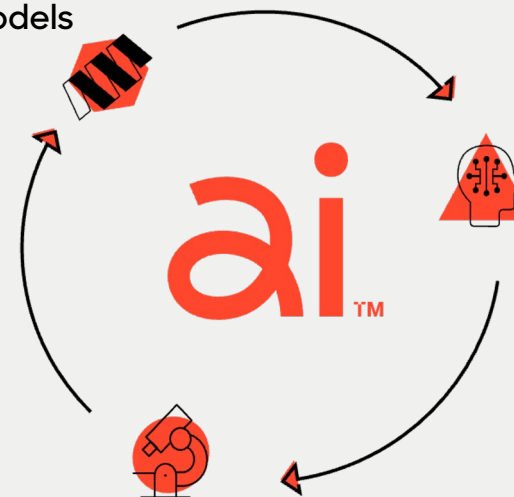
### LEAD OPTIMIZATION

AI guided lead optimization enables tunable pharmacology



## LAB-IN-THE-LOOP

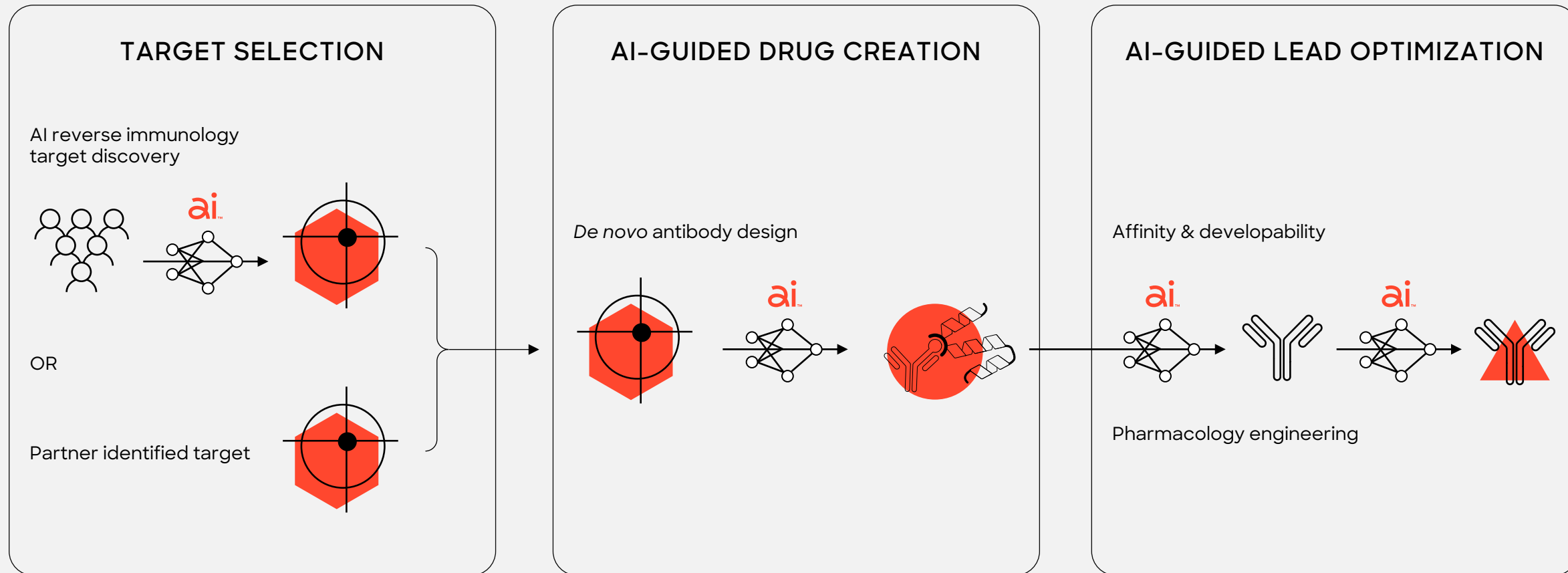
Wet Lab data improve models



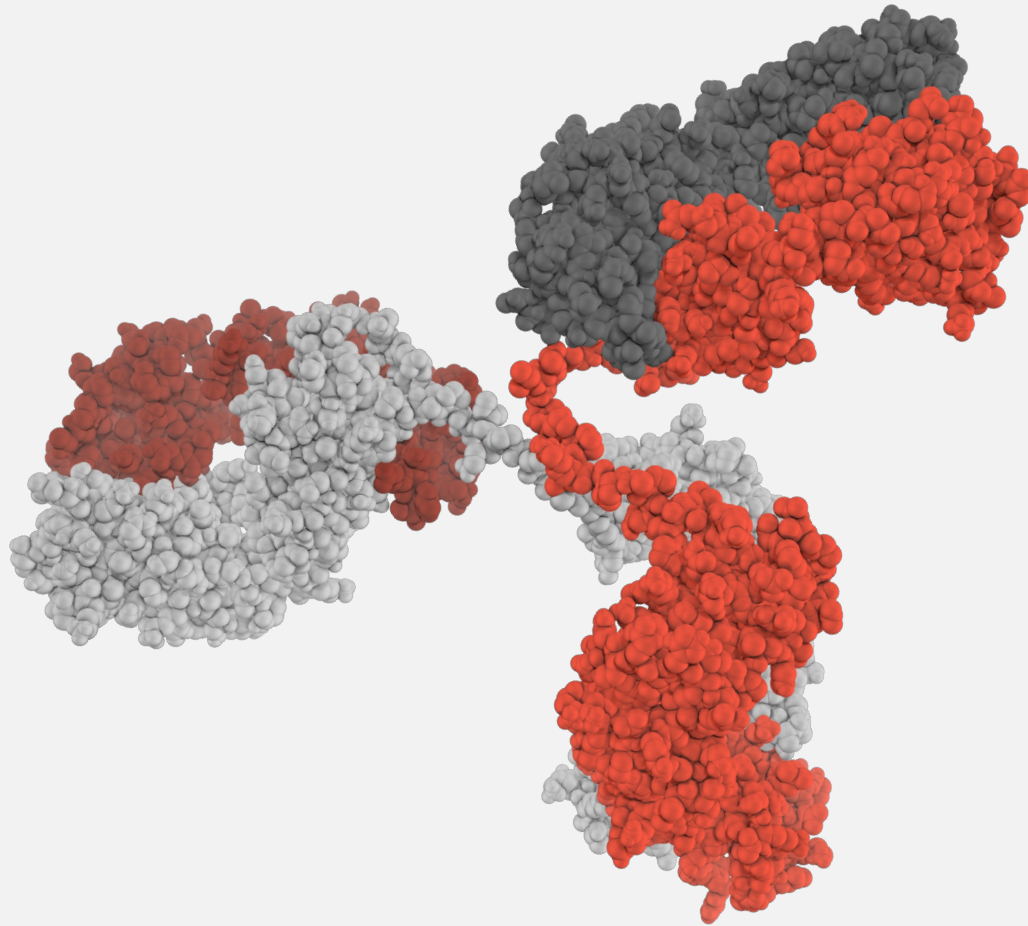
AI Platform designs antibodies

Wet Lab confirms AI-designed antibodies maintain drug-like properties

## Leveraging AI throughout the end-to-end drug discovery process



# We use AI to create novel & differentiated therapeutics



✓ EPI TOPE-SPECIFIC DESIGN +  
EPI TOPE INTERFACE OPTIMIZATION

✓ ENHANCED POTENCY AND MOA

✓ ABILITY TO ADDRESS DIFFICULT  
TARGET CLASSES, E.G. GPCRS

✓ ENABLING FEATURES: MULTI-VALENCY,  
pH-DEPENDENT BINDING

✓ BROAD IP: 100S TO 10,000S OF  
FUNCTIONALLY VALIDATED  
SEQUENCES ENABLED BY  
PROPRIETARY WET-LAB VALIDATION

# Absci partnership ecosystem

## AI Drug Creation™ Partnerships



**23 PARTNERED PROGRAMS**

**4 NAMED INTERNAL PROGRAMS**

**ADDITIONAL PROGRAMS IN EARLY DEVELOPMENT**

## Data & compute collaborations



**SCALING COMPUTE**

**IMPROVING MODELS**

**INCREASING PROGRAMS**

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# OUR PEOPLE

## "Multilingual" team with expertise in AI and drug creation

### LEADERSHIP TEAM



**Sean McClain**  
Founder, CEO & Director



**Andreas Busch, PHD**  
Chief Innovation Officer



**Zach Jonasson, PHD**  
Chief Financial Officer & Chief Business Officer



**Amaro Taylor-Weiner, PHD**  
SVP, Chief AI Officer



**Shelby Walker, JD**  
Chief Legal Officer



**Karin Wierinck**  
Chief People Officer



**Christian Stegmann, PHD**  
SVP, Drug Creation



**Christine Lemke, DVM**  
SVP, Portfolio & Growth Strategy



**Penelope**  
Chief Morale Officer

### BOARD OF DIRECTORS



**Frans Van Houten**  
Chairman of the Board  
Former CEO, Royal Phillips



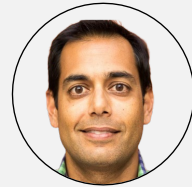
**Sean McClain**  
Founder, CEO & Board Director



**Sir Mene Pangalos, PHD**  
Former EVP R&D  
AstraZeneca



**Karen McGinnis, CPA**  
Former Chief Accounting Officer,  
Illumina



**Amrit Nagpal**  
Managing Director,  
Redmile Group

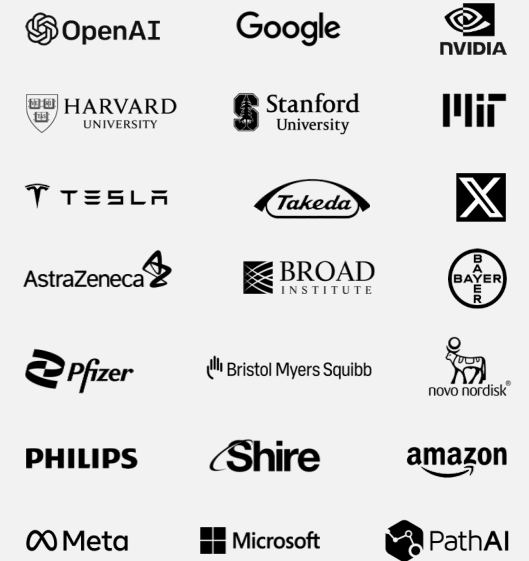


**Dan Rabinovitsj**  
VP Hardware  
Engineering, Meta



**Joseph Sirosh, PHD**  
Former CTO, Compass  
VP, Amazon & Microsoft

### EXPERTISE & BACKGROUND FROM



### SCIENTIFIC ADVISORY BOARD



**Sir Mene Pangalos, PHD**  
Co-Chair SAB  
Former EVP R&D  
AstraZeneca



**Andreas Busch, PHD**  
Co-Chair SAB  
Chief Innovation Officer



**Ian McInnes, PHD**  
Vice Principal and  
Head of College  
University of Glasgow



**Luis Diaz, MD**  
Head, Division of Solid  
Tumor Oncology  
Memorial Sloan  
Kettering Cancer  
Center



**John Wherry, PHD**  
Director, Institute for  
Immunology & Immune  
Health, University of  
Pennsylvania



**Victor Greiff, PHD**  
Associate Professor  
University of Oslo



**Hubert Truebel, MD, PHD, MBA**  
Chief Medical Officer  
AiCuris



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# ANDREAS BUSCH, PHD

## CHIEF INNOVATION OFFICER

```
library = []
for antibody in antibodies:
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    library.append((antibody, naturalness))
```

# Demonstration of Absci's progress in Drug Creation, AI platforms and internal portfolio



| On track to deliver **ABS-101** (TL1A) IND and FiH in 1H 2025

| Delivered Development candidate **ABS-201** (PRLR, best-in-class) with blockbuster potential in androgenic alopecia (and upside in endometriosis)

| Progress of first-in-class **ABS-301** with target validation. Expect full data package including efficacy read outs and potential candidate nomination in 1H 2025

## | **Expansion of internal portfolio:**

| 1 best-in-class lead in 2024

| 3 first-in-class or best-in-class targets identified in 2024



| *De novo* AI generated antibodies for **disease-relevant target epitopes and difficult targets**

| Antibodies designed for AstraZeneca target with no-known binder using *de novo* AI model

| First Hits identified for challenging Almirall target

| *de novo* AI design of antibodies to difficult target with no-known binder in partnership with BMGF and Caltech

| Successful expansion of **multi-dimensional AI-guided lead optimization**

```
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```

# AMARO TAYLOR-WEINER, PHD

CHIEF AI OFFICER

```
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Absci has been  
developing AI  
antibody design  
tools to:

➤ ADDRESS COMPLEX AND PREVIOUSLY  
“HARD TO DRUG” TARGETS

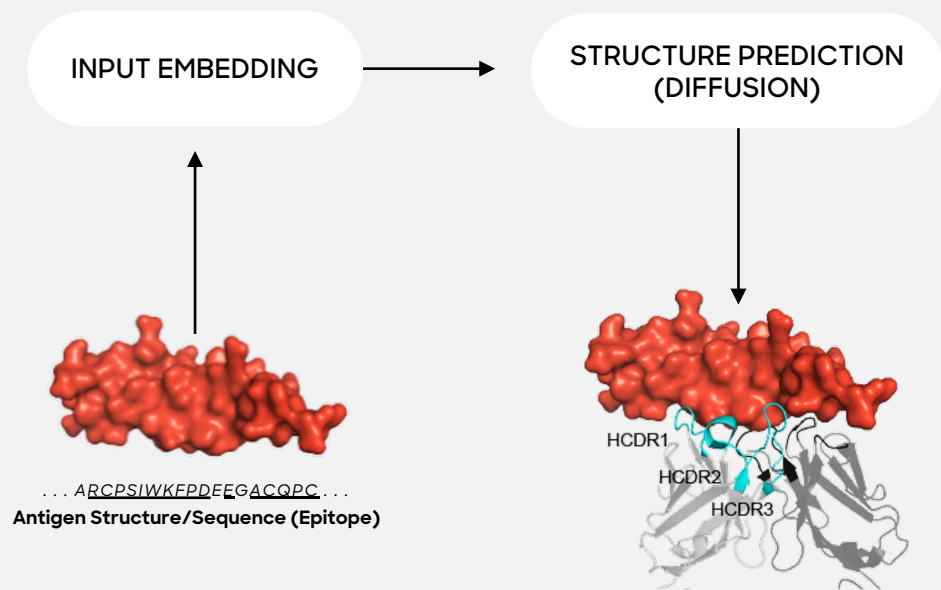
- Bind specific extracellular domains
- Target specific conformations

➤ INTRODUCE PRECISE CONTROL OVER  
ANTIBODY DESIGN:

- “Smart” biologics
- Engineer selectivity, minimizing off target toxicity
- Agonism vs antagonism
- Multidimensionally co-optimized

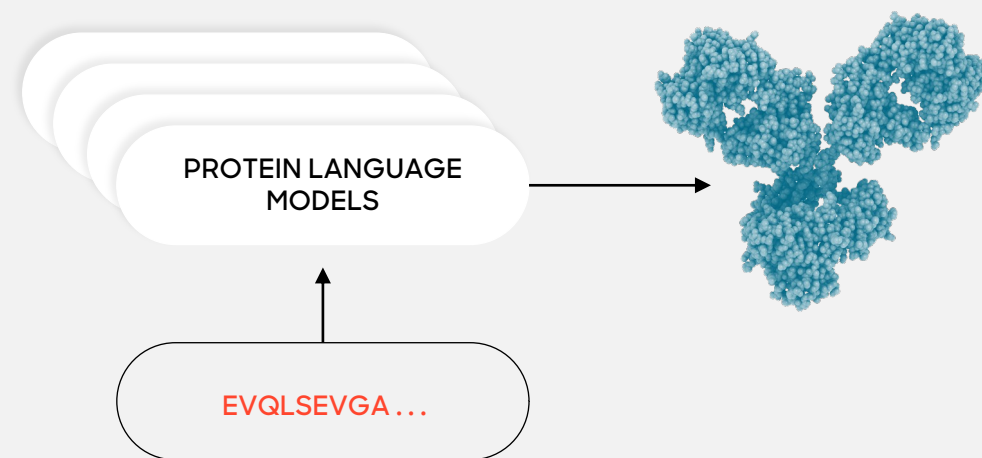
# Leadership in AI *de novo* design of antibody-based therapeutics

## DE NOVO ANTIBODY DESIGN



- › *de novo* antibody design model creates epitope-specific binders given a target structure
- › Designed in framework of choice or multiple frameworks

## AI LEAD OPTIMIZATION



- › Co-optimization enables improvement of antibody attributes while maintaining developability
- › Precise engineering of molecule pharmacology

# Our AI platforms are enabled by our 6-week 'lab-in-the loop' active learning cycles

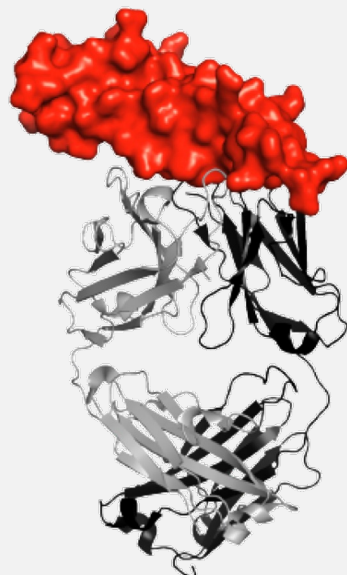
## AI PLATFORMS

### DE NOVO ANTIBODY DESIGN

*de novo* design of epitope-specific antibodies against targets without requiring a known binder

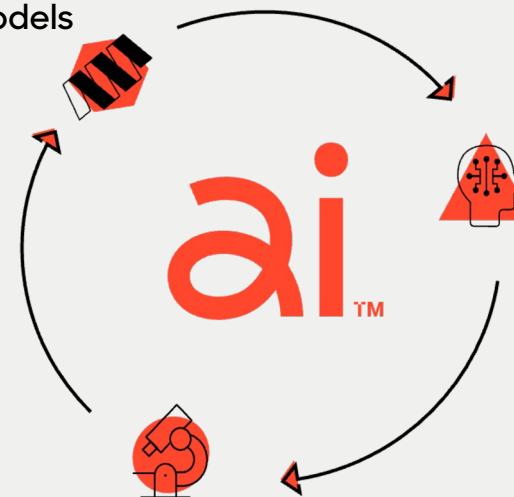
### LEAD OPTIMIZATION

AI guided lead optimization enables tunable pharmacology



## LAB-IN-THE-LOOP

Wet Lab data improve models




AI Platform designs antibodies

Wet Lab confirms AI-designed antibodies maintain drug-like properties

# DE NOVO ANTIBODY DESIGN

## AbsciDesign comprises two categories of AI models for *de novo* antibody design

**AbsciGen:**  
antibody<->antigen complex structure and sequence design



*Design 1*  
HCDR1: GFNIKDTY  
HCDR2: IYPTNGYT  
HCDR3: SRWGGDGFYAMDY

LCDR1: QDVNTA  
LCDR2: SAS  
LCDR3: QQHYTTPPT

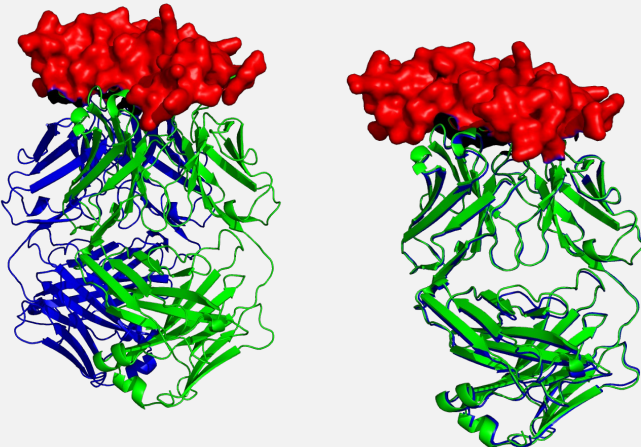
▪  
▪  
▪

*Design N*  
HCDR1: GFNIKDTW  
HCDR2: IYPSNGYT  
HCDR3: ARWGGDGFYAMDY

LCDR1: QDVNTA  
LCDR2: SAS  
LCDR3: QQHYTTPPT

The image shows a 3D model of an antibody structure. The antigen is represented as a red, textured surface. The antibody is shown as a grey ribbon structure, with the heavy chain (HCDR1-3) and light chain (LCDR1-3) regions clearly visible. The antibody is bound to the antigen.

**AbsciBind:**  
antibody design scoring and filtering



Antigen  
AbsciGen  
AbsciBind

AbsciBind  
Low Rank

RMSD = 5.3 Å  
Confidence = 0.64

AbsciBind  
High Rank

RMSD = 2.3 Å  
Confidence = 0.95

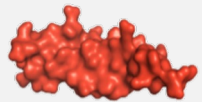
The image shows two 3D models of antibody structures. The antigen is represented as a red, textured surface. The antibody is shown as a ribbon structure, with the heavy chain (HCDR1-3) and light chain (LCDR1-3) regions clearly visible. The antibody is bound to the antigen. The legend indicates that the antigen is red, AbsciGen is green, and AbsciBind is blue. The Low Rank model has an RMSD of 5.3 Å and a Confidence of 0.64. The High Rank model has an RMSD of 2.3 Å and a Confidence of 0.95.

# The AbsciDesign AI platform delivers *de novo* antibodies via an end-to-end design-validation workflow

**STEP 1.**  
Define  
design parameters

**STEP 2.**  
Fine-tune and deploy AbsciGen and AbsciBind to generate hundreds of thousands of variants and filter to a subset that are likely binders

**STEP 3.**  
Wet lab screening and  
model performance validation

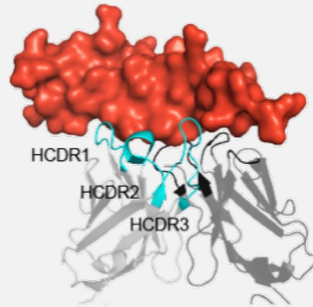


...ARCPSIWKFPDEEGACQPC...

Antigen Structure/Sequence (**Epitope**)

HFWR1: EVQLE.....GSLSCAAS    LFW1: DIQMT...RVTITCRAS  
HFWR2: IHWVR.....LEWVAR    LFW2: VAWYQ...KLLIY  
HFWR3: RYRF.....SLEDTAVYYC    LFW3: FLLQPE....DFATYYC  
HFWR4: WGQGLVTVSS    LFW4: FGQGTKVEIK

Heavy/Light Framework Sequences



**Design 1**  
HCDR1: GFNIKDTY  
HCDR2: IYPTNGYT  
HCDR3: SRWGGDGFYAMDY  
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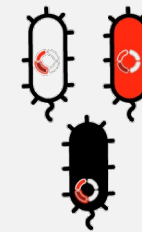
⋮

**Design N**  
HCDR1: GFNIKDTW  
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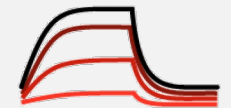
Heavy/Light CDR Sequences



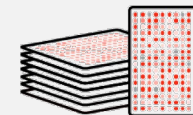
Cloning



Surface Plasmon Resonance



Expression



Sequencing





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# CASE STUDY

Collaboration with Professors Steve Mayo and Pamela Bjorkman at Caltech funded by the Bill and Melinda Gates Foundation

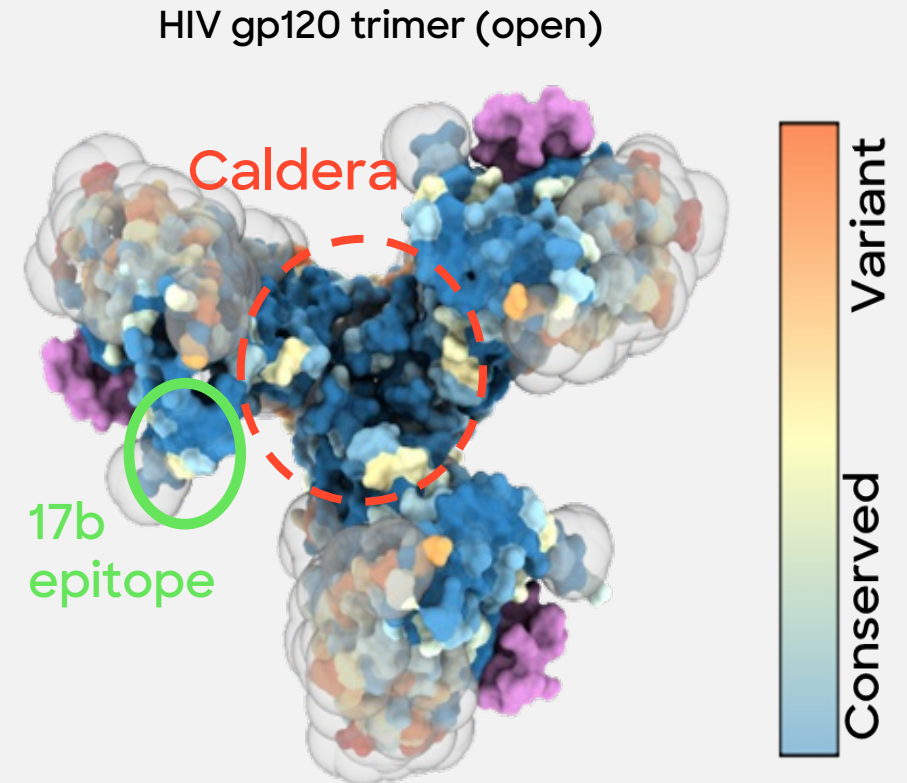
Caltech absci. BILL & MELINDA GATES foundation

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say="SPR")
```

## CASE STUDY - HIV: DE NOVO DESIGN

### de novo design antibody that binds to the highly conserved caldera region of HIV gp120

- No natural or synthetic antibody for HIV exists today because immune system cannot derive an antibody that is universally neutralizing against HIV
- Design challenge: create universally neutralizing HIV antibody by binding unique and conserved epitope within “caldera” of open conformation of gp120 to prevent HIV from entering host cells
- Numerous attempts to target this epitope have failed—previous efforts have identified antibodies, but none bind the “caldera” and none are universally neutralizing.



## CASE STUDY – HIV DE NOVO DESIGN

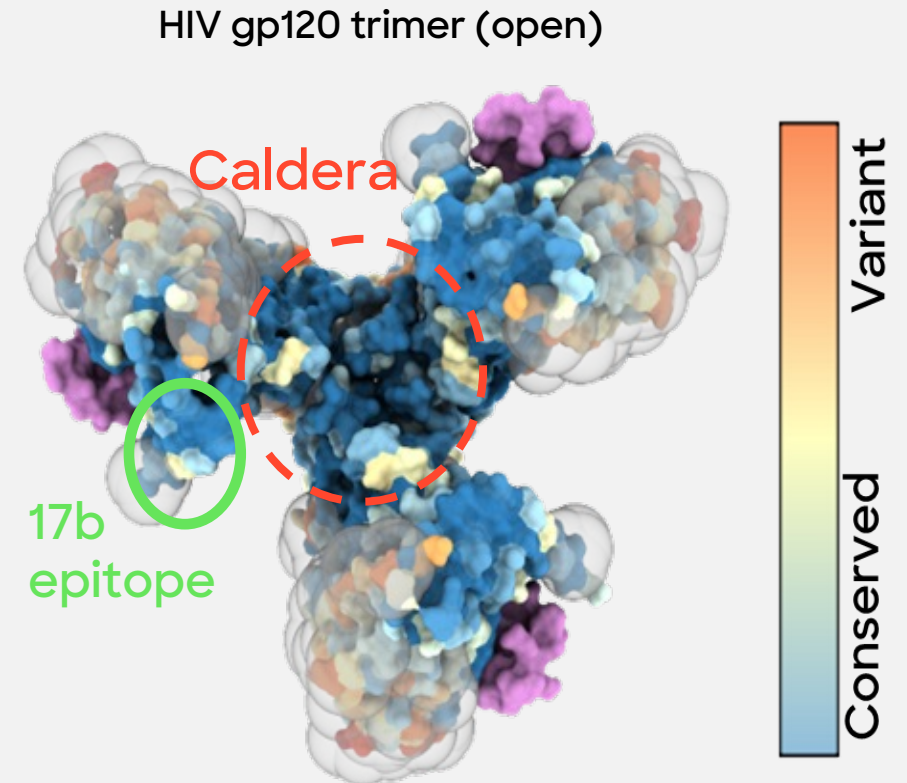
### Inputs and AI *de novo* design for HIV-Caldera

#### Model inputs:

1. Antigen structure
2. Framework of 17b
3. Epitope selected conserved across HIV strains (Clades A, B, and C)

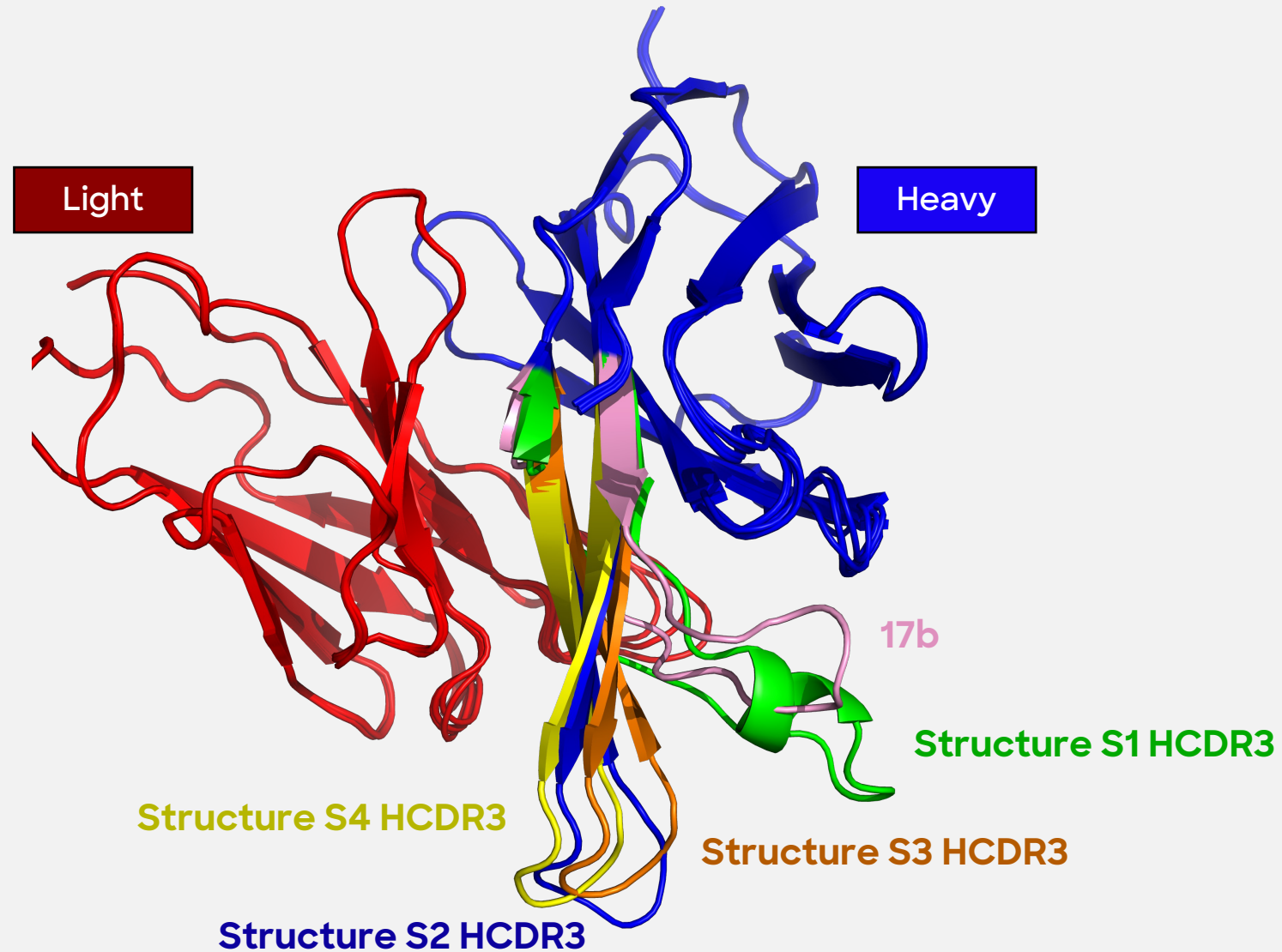
#### Design of CDRs:

- › Condition the model to design long HCDR3 to reach into open caldera region (>20 residues)
- › Designed HCDR2 and LCDR3 to bind to HIV surface



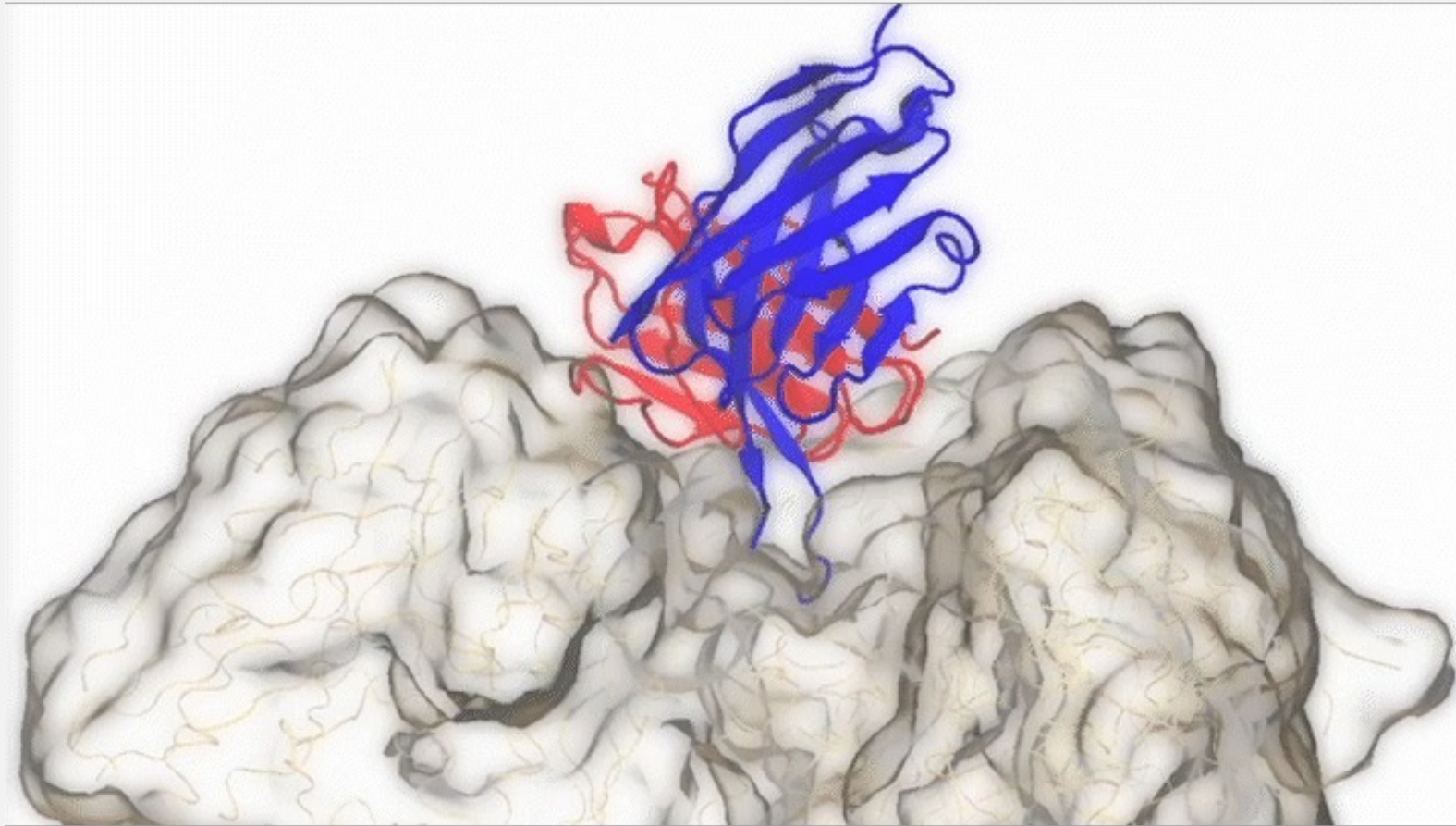
## CASE STUDY – HIV DE NOVO DESIGN

4 best structures selected from 10,000+ structures generated by *de novo* model



## CASE STUDY – HIV DE NOVO DESIGN

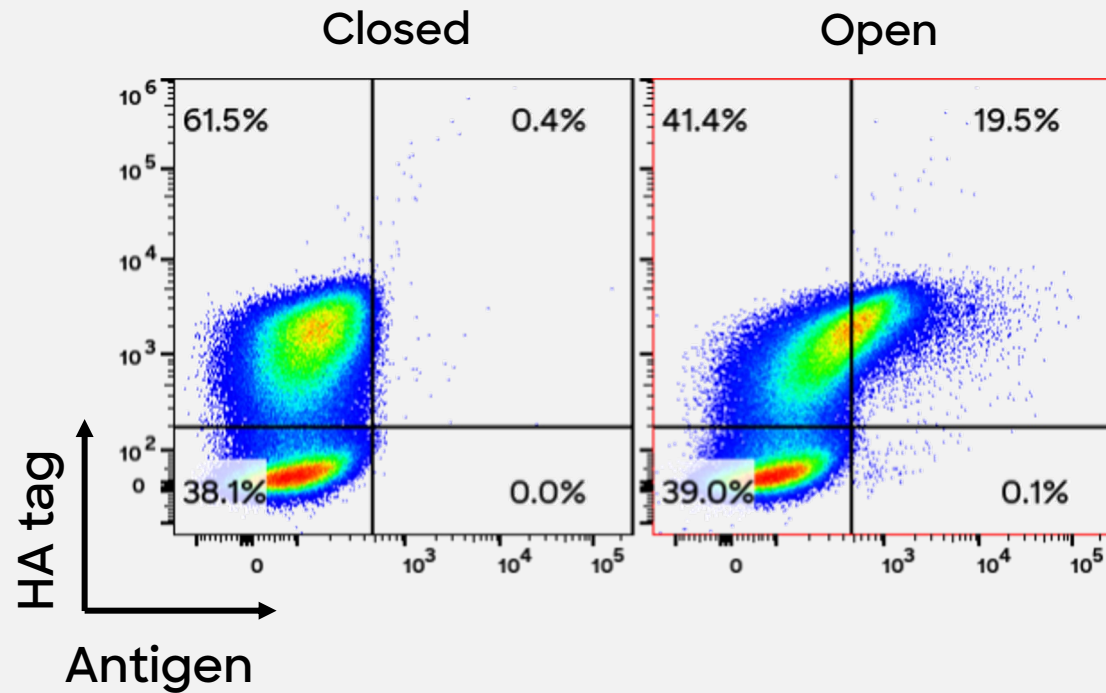
Applied molecular dynamics simulation to evaluate *de novo* designed antibodies



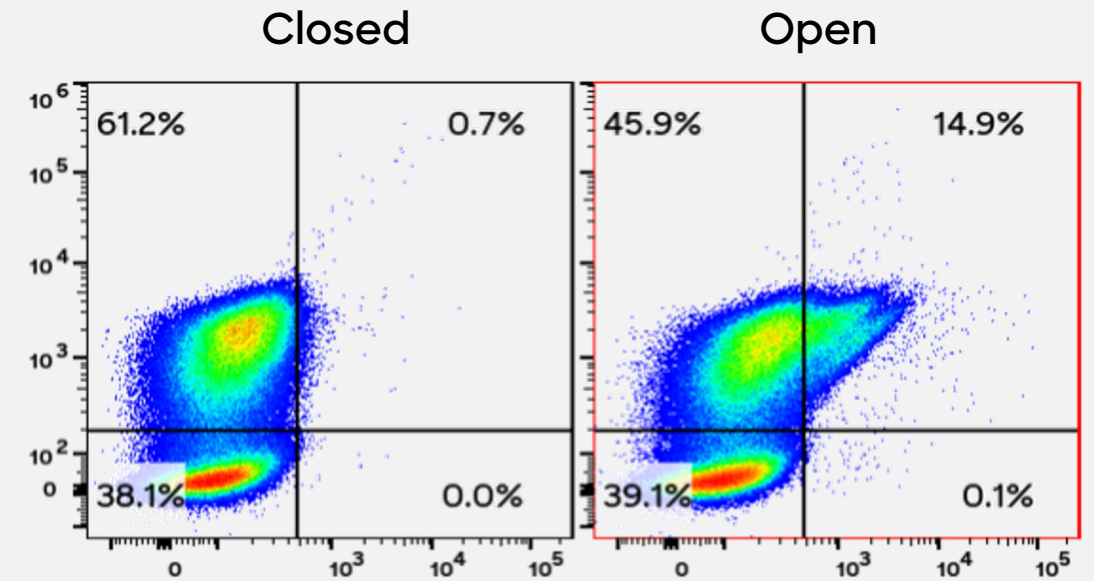
## CASE STUDY – HIV DE NOVO DESIGN

Enriched *de novo* library binds open, not closed, gp120 trimer conformation in YSD

### Clade A gp120 trimer



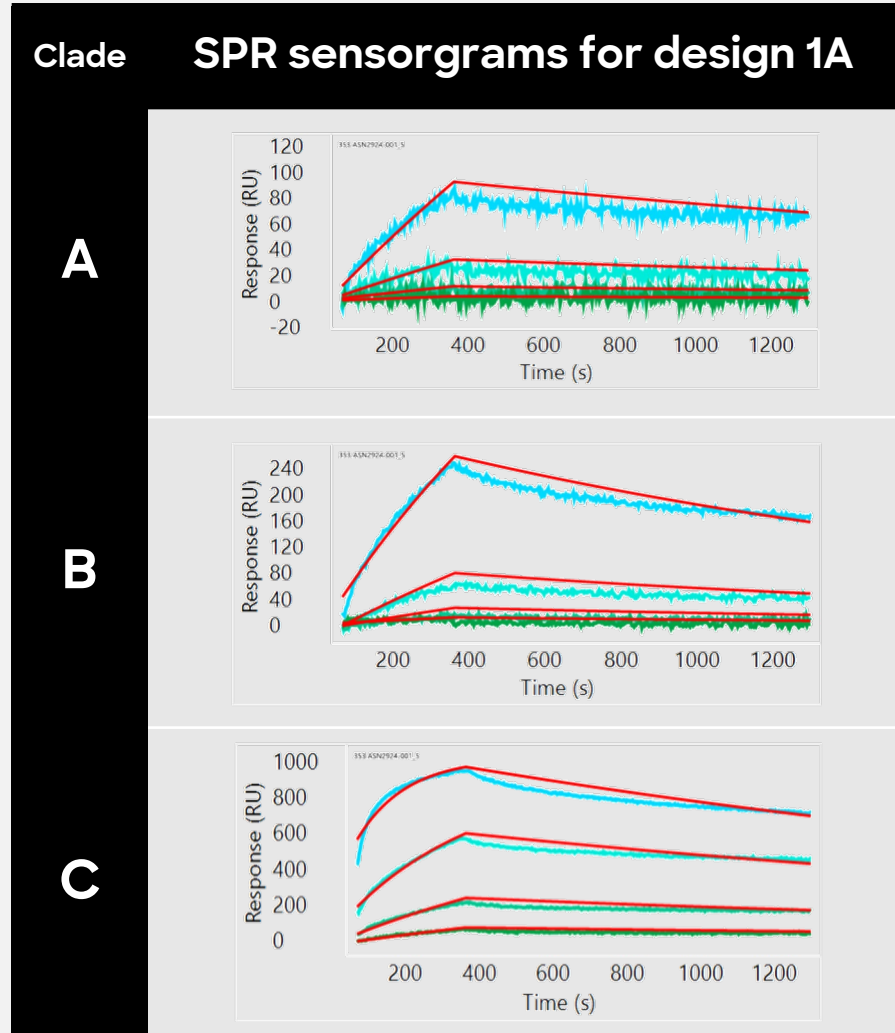
### Clade B gp120 trimer



## CASE STUDY – HIV DE NOVO DESIGN

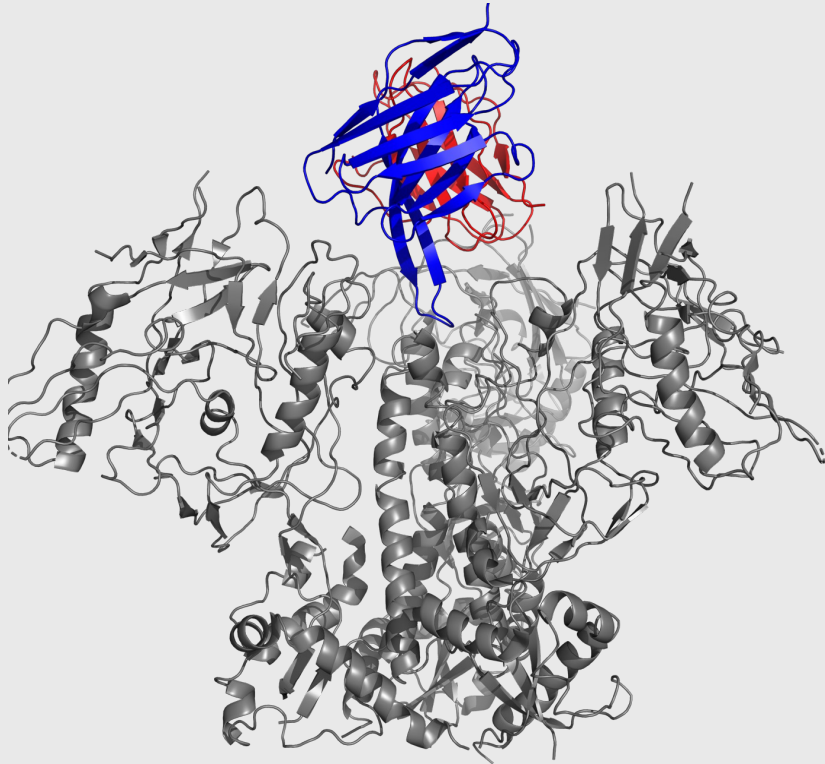
# HIV-Caldera: Preliminary SPR data demonstrate binding of *de novo* designs to open conformations of all 3 clades

- Of enriched designs, five (5) unique HCDR3s were frequently represented
  - Four (4) unique HCDR3s came from Structure S2
  - One (1) unique HCDR3 came from Structure S3
- In SPR, these designs bound across clades A, B, and C to the open conformation suggesting caldera binding



## CASE STUDY - HIV DE NOVO DESIGN TAKEAWAYS

### HIV-Caldera: demonstrating *AI de novo* design for challenging target



#### SUMMARY

- *de novo* design model created a novel and diverse antibody which binds multiple clades of HIV indicating successful targeting of the caldera epitope
- Screening cascade enabled selection of differentially binding variants

#### NEXT STEPS

- Binders from this study will be selected for affinity maturation
- Structure of *de novo* binder and epitope specificity will be experimentally solved to confirm fidelity with designed structure and targeted epitope



```
from absci_library import codon_optimizer
library
= codon_optimizer.reverse_translate(library)
library.to_csv("covid-antibody-designs.csv")
library.to_wet_lab(assay="ACE")
```

# CASE STUDY

AI Optimization for pH sensitivity

```
from absci import lead_opt_model
lead_optimizer = lead_opt_model.load_latest_model()
library.naturalness =
lead_optimizer.naturalness(library)
lead_optimizer.optimize(library).to_wet_lab(assay="SPR")
```

## CASE STUDY – AI LEAD OPTIMIZATION for pH SENSITIVITY

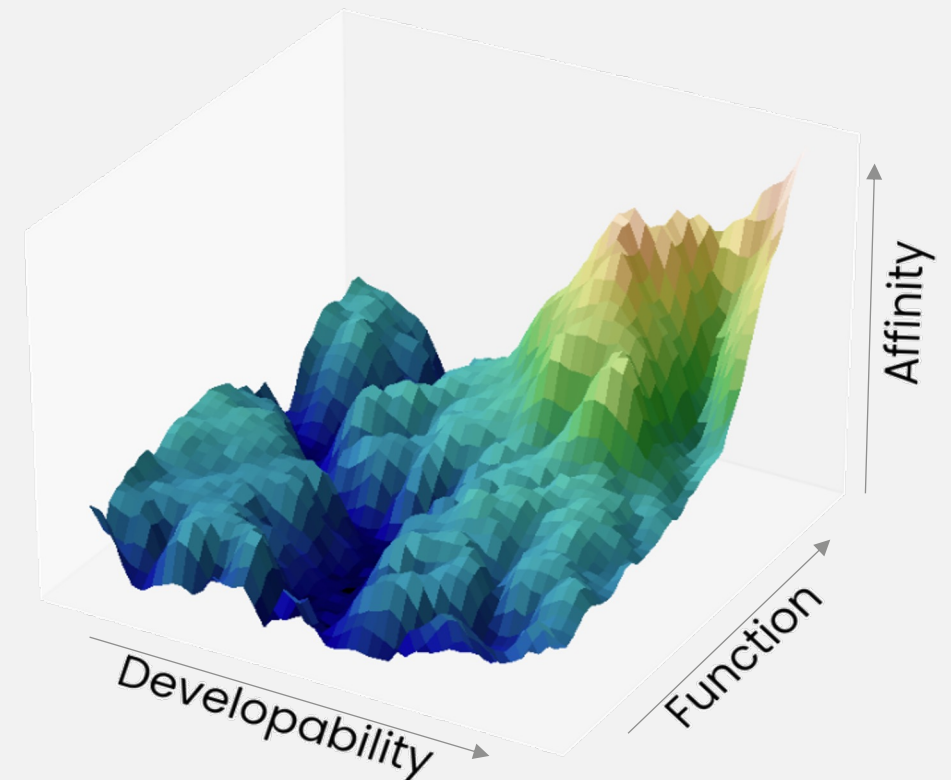
### AI lead optimization platform for ‘smart biologics’

#### THE CHALLENGE:

The diversity of antibodies is vast, making it impossible for traditional methods to explore effectively.

#### ABSCI SOLUTION:

Our AI can search a space of  $\sim 10^{19}$ , a million times larger than traditional methods, identifying functional, developable antibodies in one step.

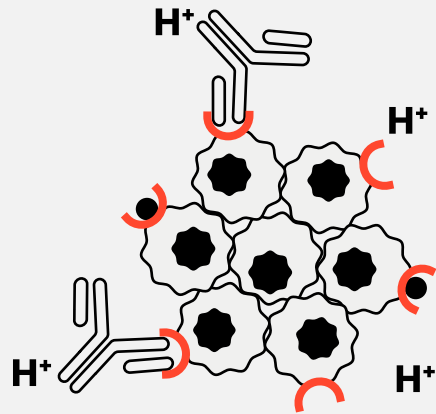


## CASE STUDY - AI LEAD OPTIMIZATION for pH SENSITIVITY

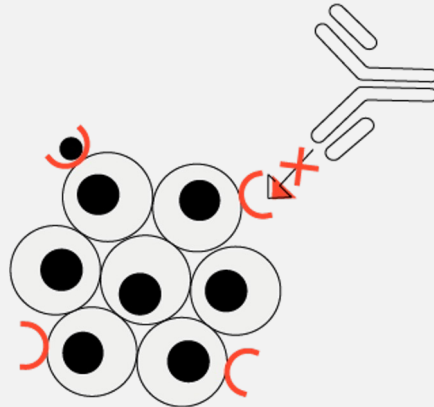
### pH sensitivity may reduce toxicity and/or improve efficacy of therapeutic mAbs

#### TUMOR SPECIFICITY IMPROVES EFFICACY AND REDUCES "ON-TARGET OFF-TUMOR" TOXICITIES

Binding occurs in the acidic pH of the tumor microenvironment



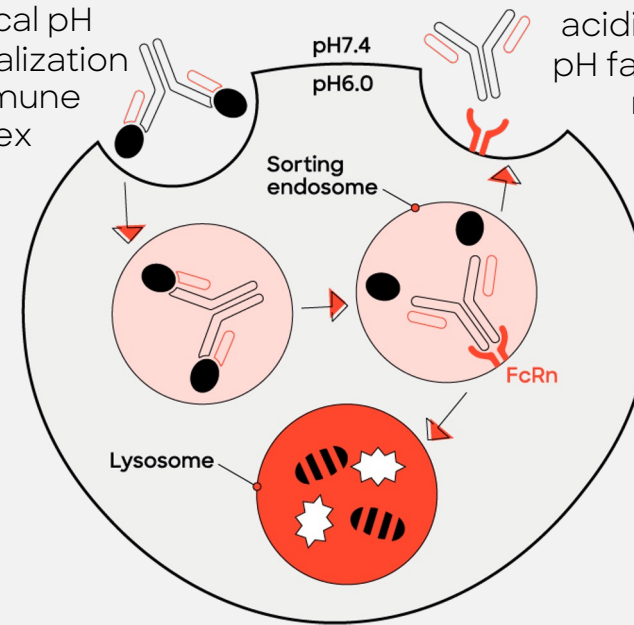
No binding occurs in the neutral pH surrounding healthy cells



#### DISSOCIATION IN THE ENDOSOME DRIVES ANTIBODY RECYCLING AND EFFICIENT CLEARANCE OF SOLUBLE TARGETS

Binding at physiological pH drives internalization of the immune complex

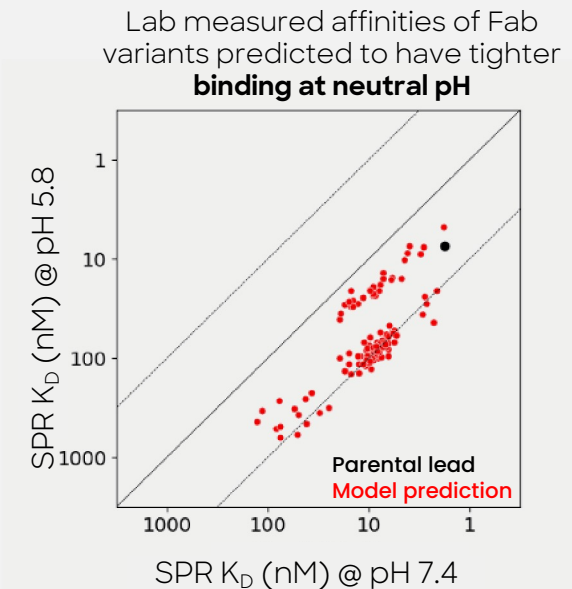
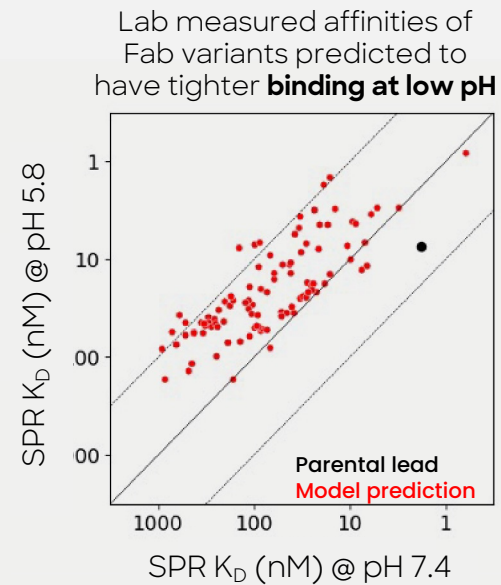
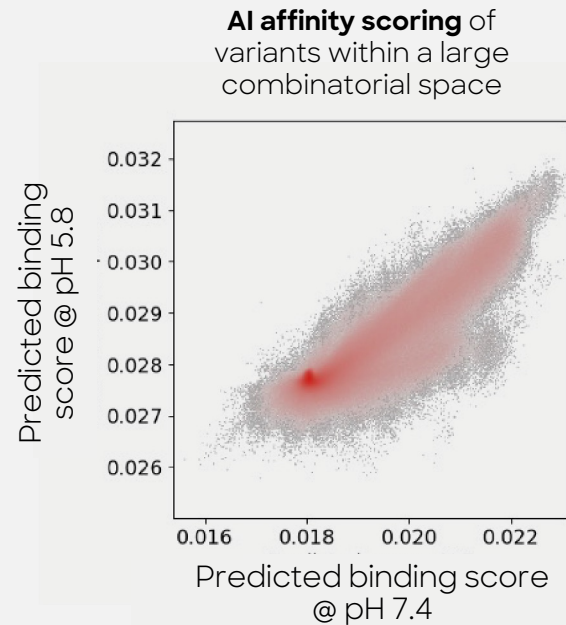
Dissociation at acidic endosomal pH favors antibody recycling



## CASE STUDY - AI LEAD OPTIMIZATION for pH SENSITIVITY

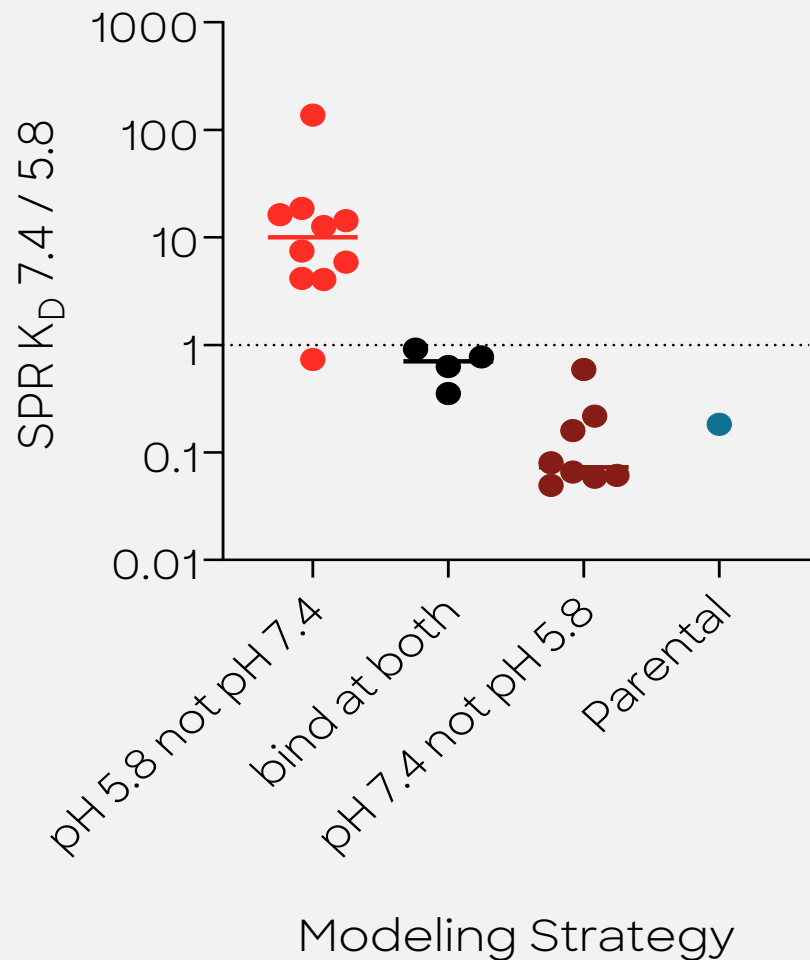
### Models identify pH sensitive Fab variants from the same lead for either indication

1. Library for model training sampled 60 positions on heavy chain framework and CDRs with up to 7 substitutions biased for ionizable residues (H, K, R, D, E)
2. Library screened for antigen binding at pH 7.4 and pH 5.8
3. Model trained and used to generate antibodies with tuned pH dependency



## CASE STUDY - AI LEAD OPTIMIZATION for pH SENSITIVITY

### Hits reformatted as mAbs show desired binding profiles



- AI optimized leads achieves variants with pH sensitive binding up to 100x differential
- pH-sensitive leads had no liabilities for stability, aggregation and polyreactivity<sup>1</sup>
- Model proposed mutations use all 6 ionizing residues in heavy chain CDRs and framework region
- Sequences were proposed from a  $>10^{13}$  combinatorial space

## Summarized platform case studies

### DE NOVO DESIGN

- › *de novo* design model created molecule binds multiple clades of HIV suggesting successful targeting of the caldera epitope
- › Represents second disclosed target success for our *de novo* platform in the 2<sup>nd</sup> half of this year

**Absci's *de novo design* platform can successfully address difficult to drug target epitopes**

### AI OPTIMIZATION

- › Models identify unseen variants with 10x-20x pH sensitivity in both directions, and up to 100x differential compared to parental molecule after only one round
- › Designed leads had no liabilities indicating the ability to successfully search a fitness landscape

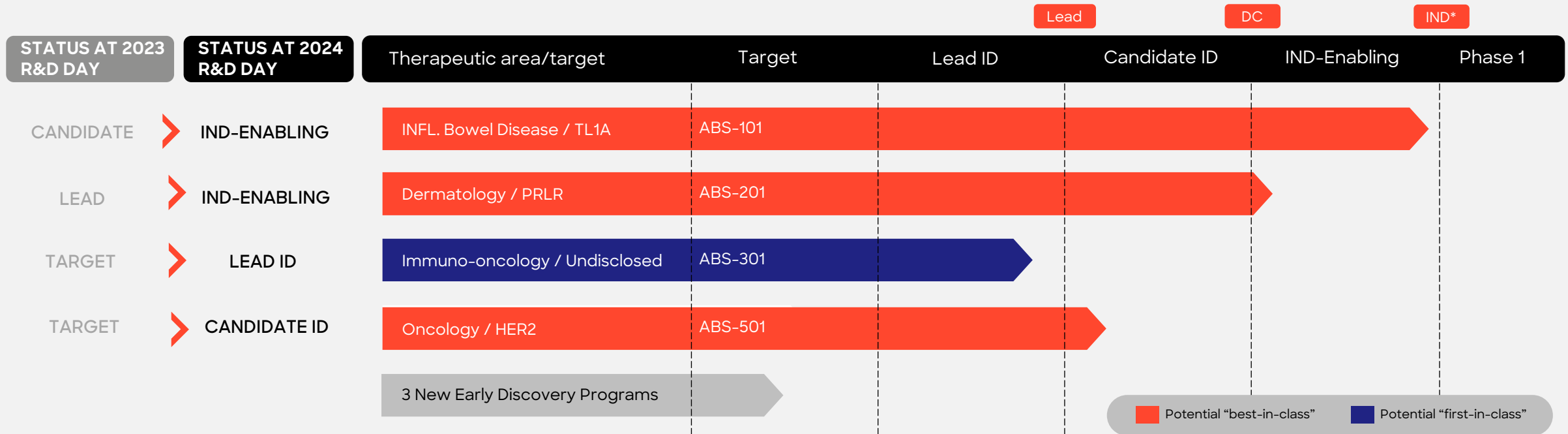
**Absci's lead optimization platform enables molecules with differentiated pharmacology**

**BREAK**



# AI PIPELINE

## Advancing and expanding our pipeline of novel & differentiated assets designed using AI



\*or equivalent ex-US filing

Partnered Programs {





## INTERNAL PIPELINE

# Absci's progress in Drug Creation

### > Continued advancement of lead assets

#### ABS-101

Continued progress of TL1A asset with FiH in 1H 2025

- New preclinical data support potentially superior immunogenicity profile
- Phase 1 Interim data readout in 2H 2025

#### ABS-201

Development Candidate for PRLR (prolactin receptor) nominated early December 2024

IND-enabling activities initiated

### > Discovery of next assets

#### ABS-301

Progress of first-in-class asset with target validation and initial preclinical efficacy readouts in 1H 2025

#### NEW: ABS-501

Nomination of a potential best-in-class HER2 asset

```
from absci_library import codon_optimizer
library = codon_optimizer.reverse_translate(library)
library.to_csv("covid-antibody-designs.csv")
library.to_wet_lab(assay="ACE")
```

```
from absci import de_novo_model
model = de_novo_model.load_latest()
antigen = model.load_pdb("7olz.pdb", chain="A")
antibodies = model.predict(antigen, N=300000)
```

# CHRISTIAN STEGMANN, PHD

SVP, DRUG CREATION

```
library = []
for antibody in antibodies:
    naturalness = model.naturalness(antibody.sequence)
    library.append((antibody, naturalness))
```

## ABS-101 TL1A

# Potential best-in-class TL1A mAb designed using generative AI



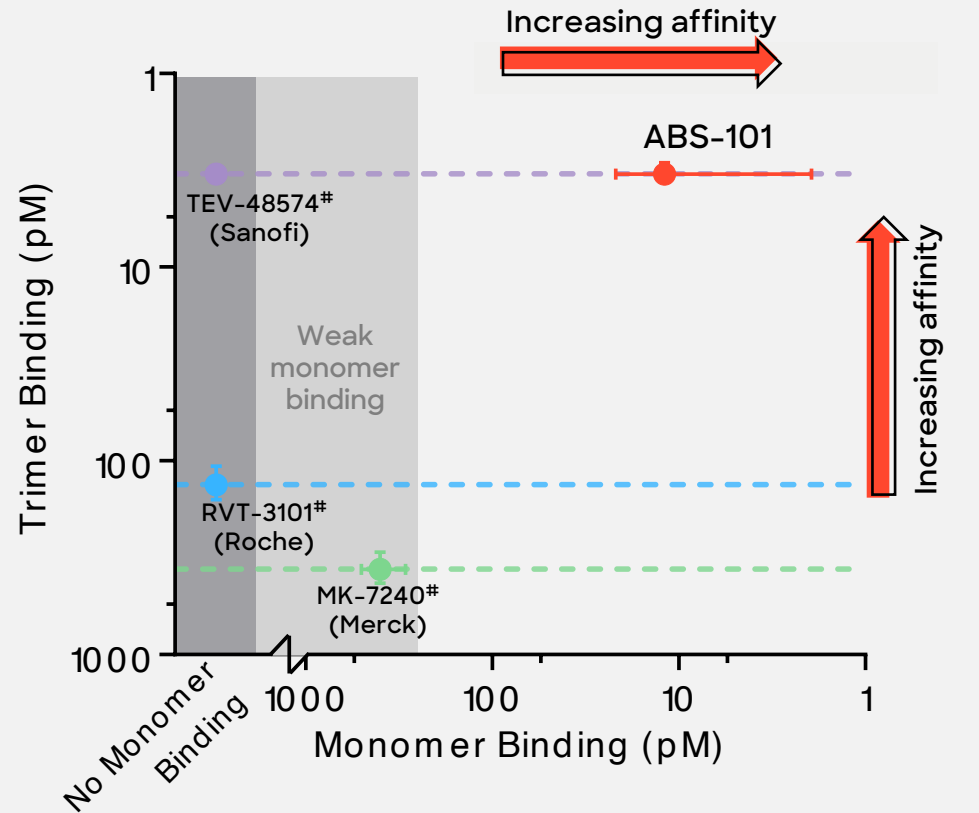
ABS-101 designed to achieve **competitive** therapeutic properties and potential for clinical **differentiation**

- Higher affinity and potency
- Bind monomer and trimer TL1A
- High bioavailability
- Expected low immunogenicity
- Favorable developability
- High convenience based on half-life extension and sub-Q dosing

# ABS-101, TL1A

## AI platform designed advanced leads with high affinity and superior potency

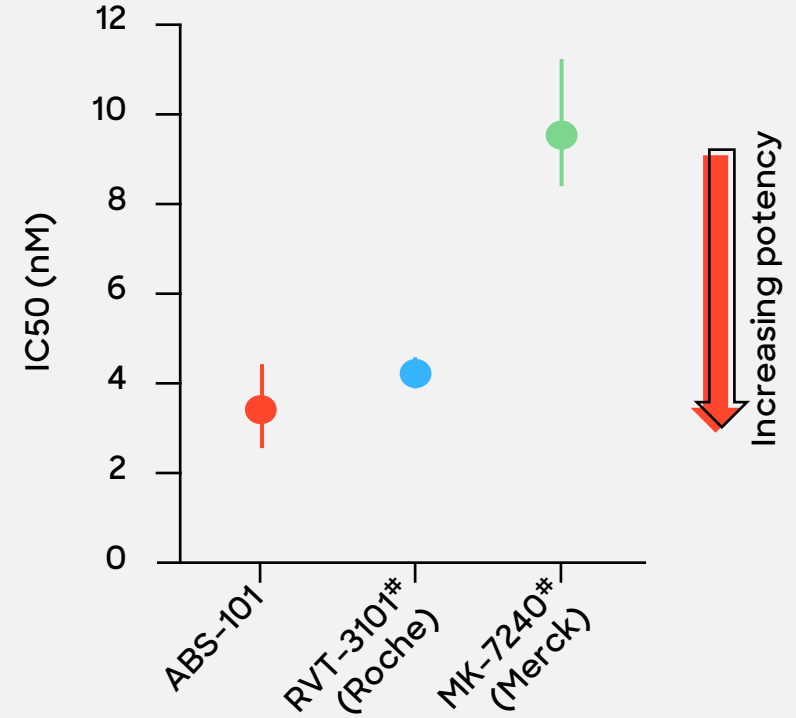
### HIGH AFFINITY MABS WITH BINDING TO BOTH THE TL1A MONOMER AND TRIMER



AFFINITY BY BIOLAYER INTERFEROMETRY (BLI)

### AI-OPTIMIZED LOW pM AFFINITY TRANSLATES TO SUPERIOR OR EQUIVALENT POTENCY

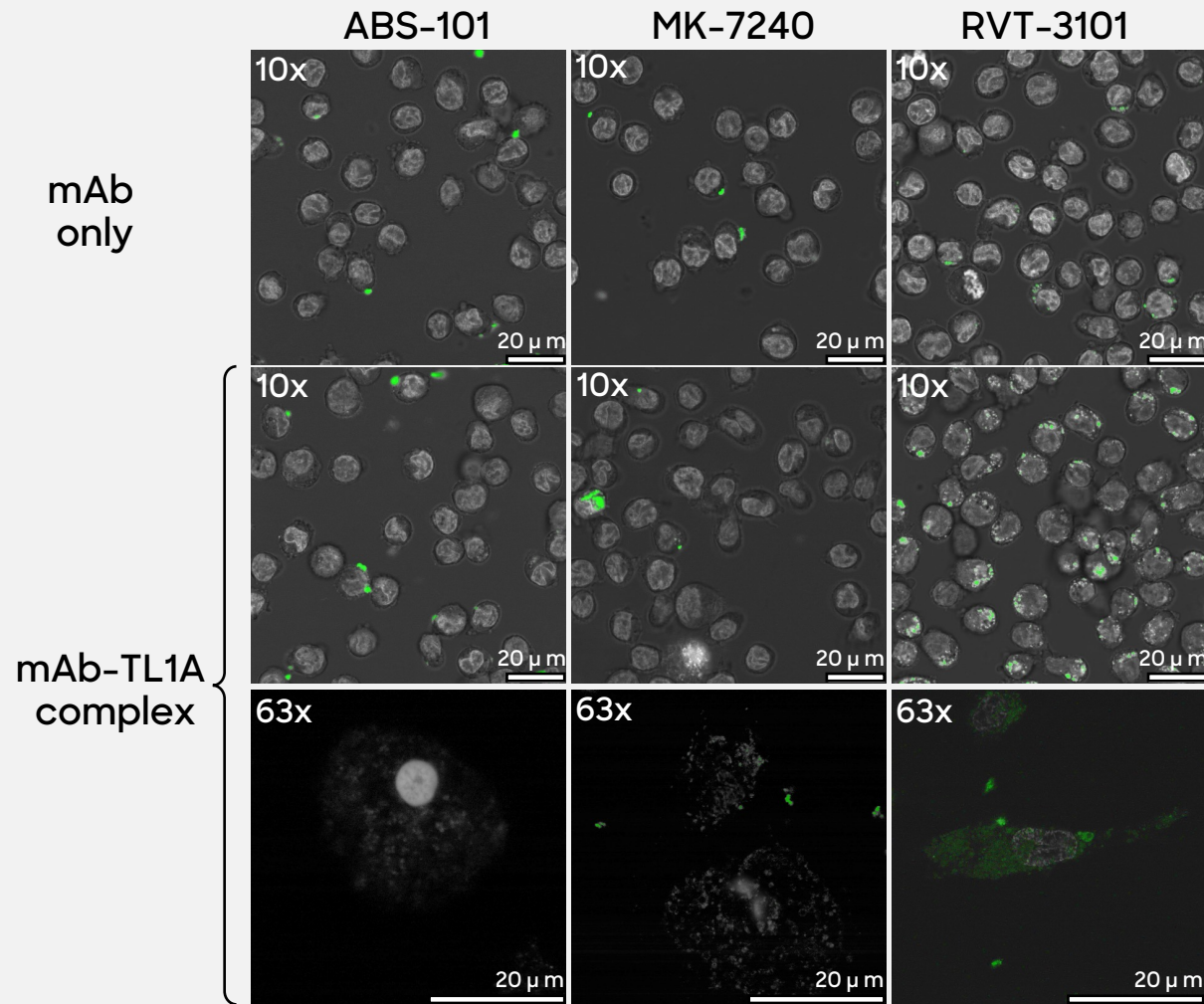
#### APOPTOSIS INHIBITION ASSAY IN TF-1 CELLS



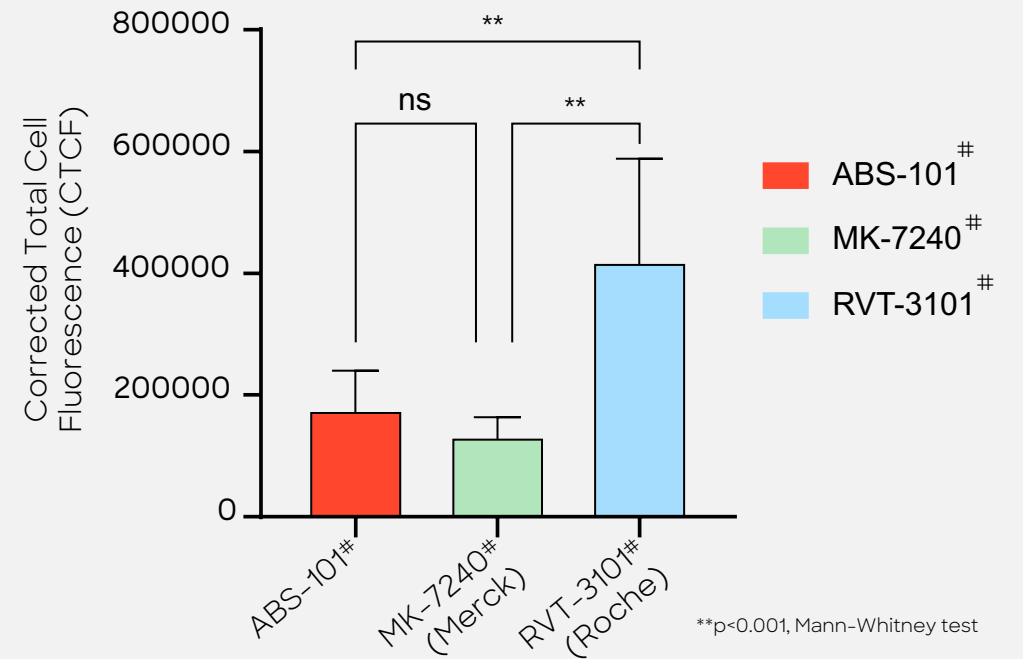
#Estimated performance of a putative clinical competitor molecule

# ABS-101, TL1A

## ABS-101 and MK-7240 show reduced TL1A complex internalization than RVT-3101



### MAB:TL1 COMPLEX INTERNALIZATION IN THP-1 CELLS



➤ Internalization of mAb:TL1A complexes potentially contributes to immune activation and formation of ADA

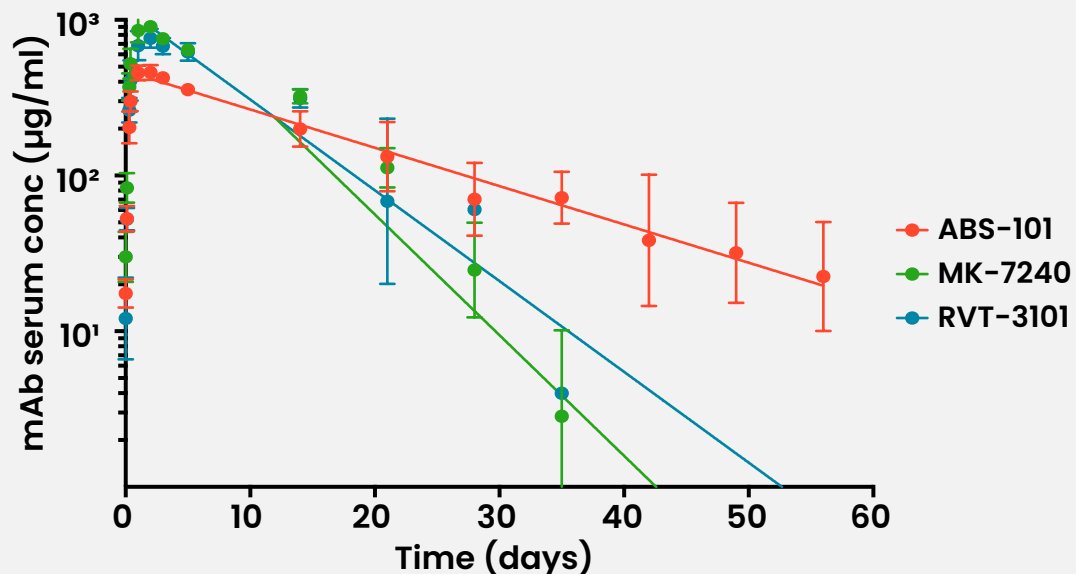
Reference, doi: 10.1053/j.gastro.2019.08.009

<sup>#</sup>Estimated performance of a putative clinical competitor molecule

## ABS-101, TL1A

# Latest Non-Human Primates & CMC data confirm compelling ABS-101 competitive profile

### 2-3X LONGER HALF-LIFE IN NHPs COMPARED TO CLINICAL COMPETITORS



### NHP-PK & PRELIMINARY 13-WEEK NHP GLP-TOX

- › 2-3x extended half-life in NHPs over clinical competitors to support Q8W-Q12W dosing interval
- › ABS-101 shows enhanced biodistribution in NHPs, compared to antibodies in clinical development based on in silico modelling
- › High subcutaneous bioavailability in NHPs at ~80%
- › Preliminary 13-week GLP-tox shows no treatment-related adverse findings during in-life phase and necropsy; histopathology pending

### CMC - HIGH CONCENTRATION FORMULATION

- › Optimal developability profile allowed successful development of high-concentration formulation at 200mg/mL suitable for subcutaneous injection

## ABS-101, TL1A

### AI-designed for potentially optimal therapeutic profile

ATTRIBUTE	ABS-101	MK-7240 (MERCCK, PROMETHEUS)	RVT-3101 (ROCHE, ROIVANT)	TEV-48574 (SANOFI, TEVA)
High affinity/potency	++	-	+	+
Trimer TL1A binding	++	+	+	++
Monomer TL1A binding	++	+	-	-
Low Immunogenicity potential	+	+	-	NA
Bioavailability/ Biodistribution	++	+	-	NA
Sub-Q injection	+	+	+	-
Q8W to once quarterly dosing	++	-	-	--

# Continued progress with FiH expected in 2025

## ● 1Q 2024

### AI-designed Development Candidate

- ✓ High affinity
- ✓ High potency
- ✓ Long half-life
- ✓ Favorable manufacturability



## ● INITIATED FEB 2024

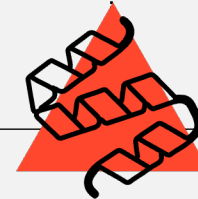
### IND-enabling studies to evaluate:

- ✓ GMP manufacture of sub-Q formulation at high concentration
- ✓ Favorable PK and long half-life
- ✓ High Bioavailability in NHPs
  - Low ADA
- ✓ 13-week GLP tox: No treatment-related adverse findings during in-life phase and necropsy observed. Histopathology pending



## ● 1H 2025

Phase 1 double-blind, placebo-controlled trial initiation



## ● 2H 2025

Phase 1 interim data readout





## INTERNAL PIPELINE

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#### NEW: ABS-501

Nomination of a potential best-in-class HER2 asset

## WHY ARE WE EXCITED ABOUT ABS-201?

# Rationale for developing a PRLR (prolactin receptor) antibody in androgenic alopecia

### Clinical and commercial unmet need

- Significant unmet clinical need for androgenic alopecia
- Large market: 80M patients in US, which is a highly motivated patient population



### Scientific rationale

- Highly validated target (efficacy & safety) for treatment of androgenic alopecia
- Supportive pharmacological profile of ABS-201



### Development path

- Straightforward clinical development path with option for early PoC
- Low competition, potentially first to US market

› Anthony Rossi, MD

› Mike Jafar

› Christian Stegmann, PhD

```
from absci_library import codon_optimizer
library
= codon_optimizer.reverse_translate(library)
library.to_csv("covid-antibody-designs.csv")
library.to_wet_lab(assay="ACE")
```

# UNMET CLINICAL NEED FOR ALOPECIA

ANTHONY ROSSI MD, FAAD, FACMS  
ATTENDING DERMATOLOGIST - MEMORIAL SLOAN KETTERING CANCER CENTER  
PROFESSOR OF DERMATOLOGY - WEILL CORNELL MEDICAL COLLEGE

```
from absci import lead_opt_model
lead_optimizer = lead_opt_model.load_late
library.naturalness =
lead_optimizer.naturalness(library)
lead_optimizer.optimize(library).to_wet_l
say="SPR")
```



## ANTHONY ROSSI, MD, FAAD, FACMS

ATTENDING DERMATOLOGIST - MEMORIAL SLOAN KETTERING CANCER CENTER

PROFESSOR OF DERMATOLOGY - WEILL CORNELL MEDICAL COLLEGE

**Professional Expertise:** Double board-certified dermatologist and micrographic surgeon specializing in cutaneous oncology, aesthetic dermatology, and hair loss.

**Research Contributions:** Conducted 15+ clinical trials, including collaborations on novel drug therapies with major pharmaceutical companies. Pioneered restorative oncodermatology to address skin effects of cancer treatments; led advancements in treating aging and hair loss related to cancer therapy.

### Editorial & Leadership Roles:

- Assistant Editor: *Journal of the American Academy of Dermatology*, *Dermatologic Surgery*, *Lasers in Surgery and Medicine*.
- Leadership: Executive Board, American Society for Dermatologic Surgery (ASDS); Nominating Committee, American Academy of Dermatology (AAD).
- Member: ASLMS, ACMS.

### Education & Training:

- BS, New York University.
- MD, Weill Cornell Medical College.
- Residency: Dermatology, St. Luke's Roosevelt Hospital.
- Fellowship: MSKCC and Weill Cornell Medical College.

# Experience and feedback from alopecia patients

## GROWING # OF PATIENTS ARE ACTIVELY SEEKING SOLUTIONS FOR HAIR LOSS

Patients cut across demographics and includes young women and men

Patients often self-diagnose and have tried various treatment options with limited success

Patients want FDA approved therapeutics that are safe and efficacious

Female patients in particular lack FDA approved treatment

Hair loss has significant psychosocial impact on quality of life



# Male and female androgenetic alopecia comprises the vast majority of alopecia patients



## MALE ANDROGENETIC ALOPECIA

Male androgenetic alopecia, also known as male pattern hair loss is a common type of hair loss that affects men and characterized by frontotemporal and vertex thinning.



## FEMALE ANDROGENETIC ALOPECIA

Female androgenetic alopecia, also known as female pattern hair loss is a common form of hair loss that affects adult women. It's characterized by a gradual thinning of hair at the top of the head and widening of the part, while the hairline usually remains intact

**80 - 90 MILLION AMERICANS LIVE WITH ANDROGENETIC ALOPECIA**

**AFFECTS OVER 50% OF MEN BY AGE 50**

**AFFECTS OVER 40% OF WOMEN BY AGE 50**

Alopecia incidence is anticipated to rise with increasing uptake of GLP-1s. Jastreboff A, et al. (2022) report approximately 5% of patients taking Tirzepatide experienced alopecia as an adverse event

See Jastreboff A, et al. Tirzepatide Once Weekly for the Treatment of Obesity. N Engl J Med 2022;387:205-216.

## SIGNIFICANT CLINICAL UNMET NEED

# Limited treatment options exist for male androgenetic alopecia



**AFFECTS ~50M MEN  
IN THE UNITED STATES**

### LIMITED TREATMENT OPTIONS DUE TO LACK OF EFFICACY AND SIDE EFFECTS

- **Minoxidil - FDA Approved (Topical)**
  - Requires continuous use - compliance is an issue
  - Efficacy varies, and is driven by SULT1A1 activity in an individual, with higher activity is associated with a better response to minoxidil
  - Side effects: temporary hair shedding, scalp irritation, and changes in hair texture and sometime palpitations or hypotension
- **Finasteride - FDA Approved (Oral)**
  - Requires lifetime use and daily compliance is an issue
  - Serious sexual side effects have been reported - even described as permanent - Finasteride Syndrome
  - Linked to depression
  - Cannot be handled by pregnant patients and not to be used while conceiving - may decrease sperm count
  - Some reports show efficacy of slowing down hair loss
- **Other options**
  - Low-Level Laser Therapy: mixed evidence on efficacy and overall lower effective vs. pharmaceuticals
  - Hair Transplants: invasive; limited by donor hair availability; does not address ongoing alopecia in untreated areas - and requires concomitant hair loss suppression

## SIGNIFICANT CLINICAL UNMET NEED

# Even more limited treatment options exist for female androgenetic alopecia



**AFFECTS ~30M WOMEN  
IN THE UNITED STATES**

### *LIMITED TREATMENT OPTIONS DUE TO LACK OF EFFICACY AND SIDE EFFECTS*

- **Minoxidil - FDA Approved (Topical)**
  - Requires continuous use - compliance is an issue
  - Efficacy varies, and is driven by SULT1A1 activity in an individual, with higher activity is associated with a better response to minoxidil
  - Side effects: temporary hair shedding, scalp irritation, and changes in hair texture and sometime palpitations or hypotension
- **Finasteride**
  - Not approved by FDA for female androgenetic alopecia in childbearing women due to safety concerns including potential birth defects
- **Other options**
  - Low-Level Laser Therapy: mixed evidence on efficacy and overall lower effective vs. pharmaceuticals





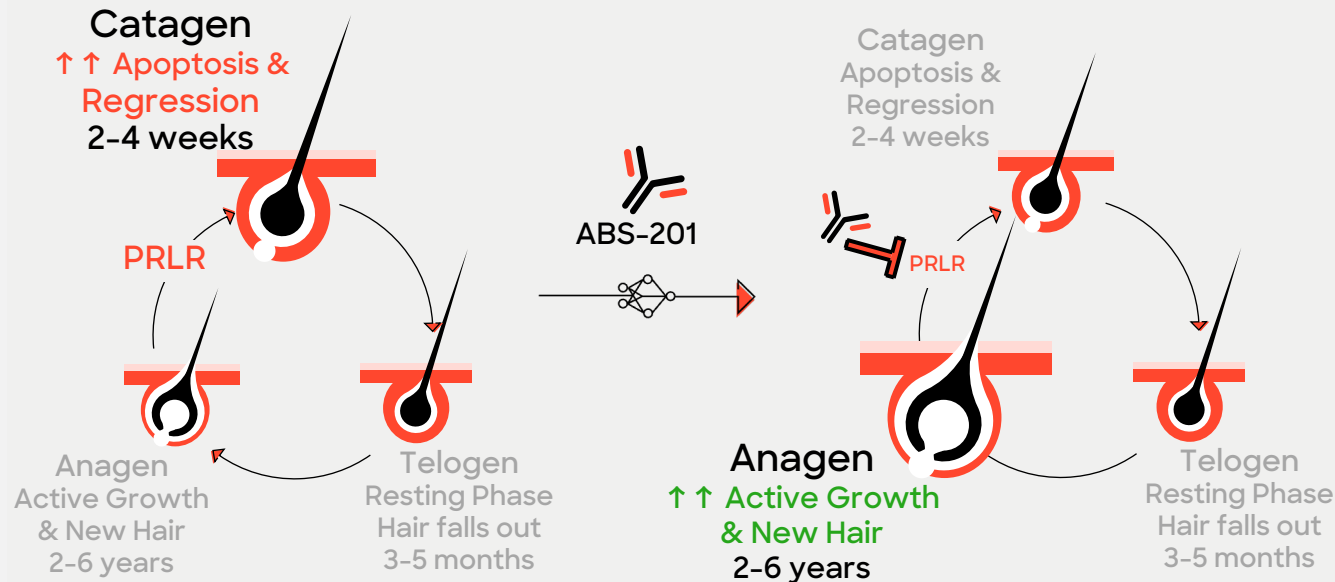
**LACK OF INNOVATION IN  
THE ANDROGENETIC  
ALOPECIA THERAPEUTIC  
LANDSCAPE OVER THE  
PAST 25+ YEARS**

**LAST FDA APPROVED  
THERAPY IN 1997**

## **Patients and clinicians need better treatment options for “hair re-growth”**

- Hair re-growth, not just slowing of hair loss
- Safe
- Minimal side effects
- Durable effect
- Convenient administration frequency
- FDA approved

# ABS-201 targets a novel mechanism that promotes hair into the anagen phase



## ABS-201 TARGET PRODUCT PROFILE:

- ✓ Significant hair **re-growth** (vs minoxidil)
- ✓ Safe with minimal side effects
- ✓ Durable effect
- ✓ Convenient infrequent dosing

## BENCH TO BEDSIDE

# Straightforward path for ABS-201 clinical development

## CLINICAL TRIALS FOR HAIR TREATMENTS ARE STRAIGHTFORWARD

- Ease of patient recruitment
- High level of KOL Interest
- Ability to conduct multi-center trials
- Non-invasive trial conduct

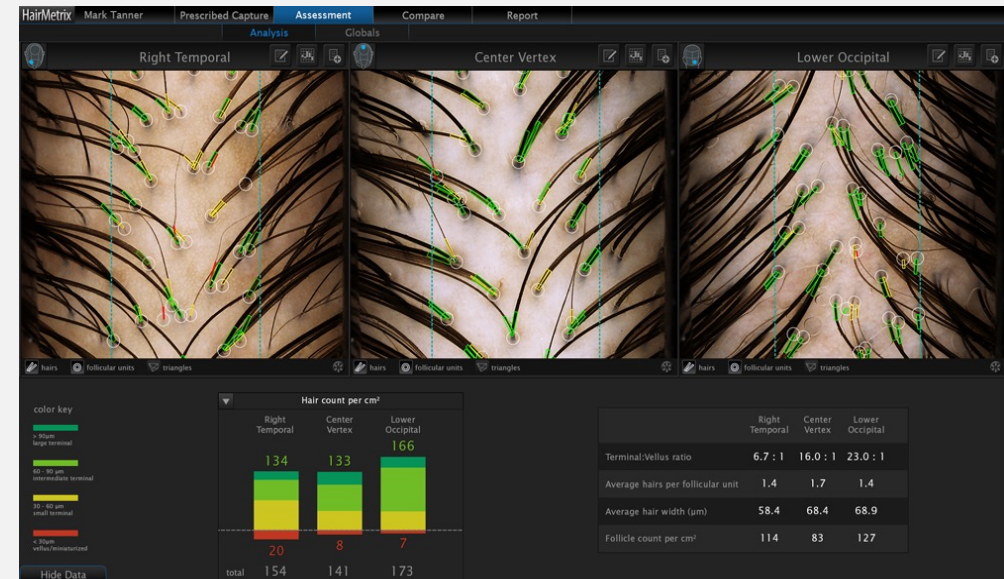
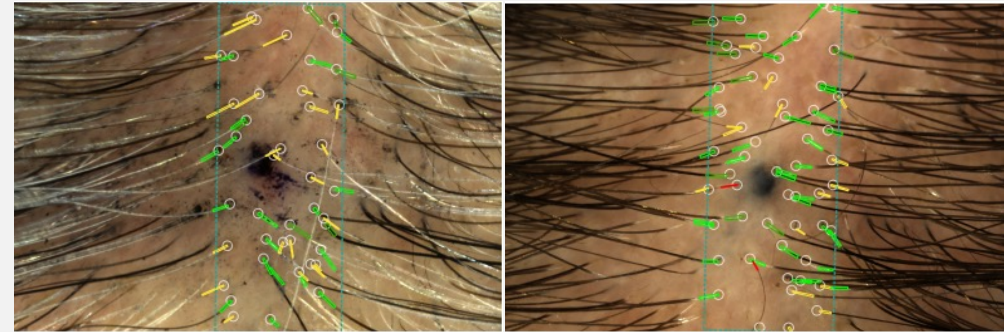
## WELL DEFINED ENDPOINTS WITH VALIDATED MEASURES

**Primary Endpoints:** Quantitative measurements with follicular dermatoscope (trichoscopy)

- Terminal Hair Growth
- Total Hair Count
- Total hair density (per cm<sup>2</sup>)

**Secondary Endpoints:**

- Patient Reported Outcomes as measured by validated scales accepted by the FDA (HairDex; Hair Specific Skindex-29 (FPHL); The Men's Hair Growth Questionnaire (MHGQ)); Women's Hair Growth Questionnaire (WHGQ)
- Hair color - regimentation



# Leading KOL network with extensive clinical and commercial reach

Over 500,000 alopecia patients treated each year by these KOL practice networks



DR. ANTHONY ROSSI  
Memorial Sloan  
Kettering Cancer  
Center



DR. KEN WASHENIK  
Bosley Medical Group



DR. MARIA K. HORDINSKY  
Univ. of Minnesota



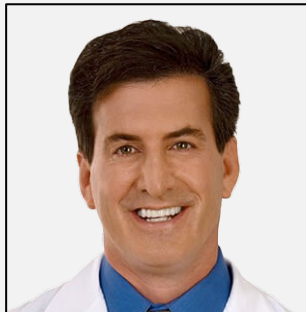
DR. NEIL S. SADICK  
Sadick Dermatology



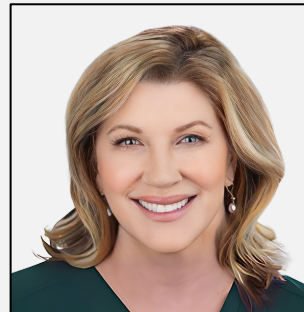
DR. MEENA SINGH  
Skin and Hair Center



DR. DORIS DAY  
Day Dermatology &  
Aesthetics



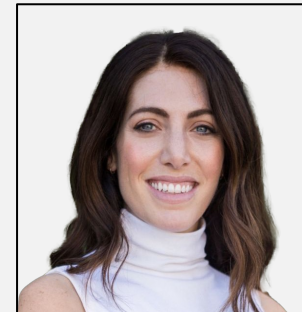
DR. MATT L. LEAVITT  
Advanced  
Dermatology and  
Cosmetic Surgery



DR. SUZANNE KILMER  
Laser & Skin Surgery  
Center of Northern  
California



DR. GLYNIS ABLON  
Ablon Skin Institute



DR. CHRISTINA RING  
ZENA Medical



DR. CHESAHNA KINDRED  
Kindred Hair & Skin  
Center

```
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library
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library.to_wet_lab(assay="ACE")
```

# MARKET OPPORTUNITY

MIKE JAFAR  
MEDICAL AESTHETICS STRATEGIC ADVISOR | BCG  
ADVISOR | ABS CI

```
from absci import lead_opt_model
lead_optimizer = lead_opt_model.load_late
library.naturalness =
lead_optimizer.naturalness(library)
lead_optimizer.optimize(library).to_wet_l
say="SPR")
```



## MIKE JAFAR

MEDICAL AESTHETICS STRATEGIC ADVISOR, BCG  
ADVISOR, ABSCI

### Professional Expertise:

Over 20 years in healthcare and medical aesthetics, played a pivotal role in launching or scaling iconic products:



Led initiatives at Allergan, Evolus, and Desktop Health, driving innovation in aesthetics and healthcare technology. Founder of JOYA Health and Xtresse

Over \$8B worth of M&A in the medical aesthetics category

Medical Aesthetics Strategic Advisor to the Boston Consulting Group (BCG)

# ABS-201 represents a new category offering hair re-growth



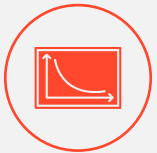
## Unlocks New Category Meeting Massive Consumer Need

- Lack of effective treatments and lack of innovation in the hair space opens opportunity for new science-based therapy
- An effective therapy that regrows hair has long been considered the last frontier of aesthetics



## Constant Consumer Demand Driven by Lifelong Focus on Skin & Hair

- The skin is a lifelong priority for consumers, fueling consistent demand for anti-aging, hydration, and appearance-enhancing treatments across all income and age groups



## Strong Willingness to Self-Pay Across Demographics

- Consumers across income levels invest in aesthetic treatments, with flexible pricing making procedures accessible and driving steady demand



## Appearance-Driven Culture Drive On-going Investment in Aesthetic Products

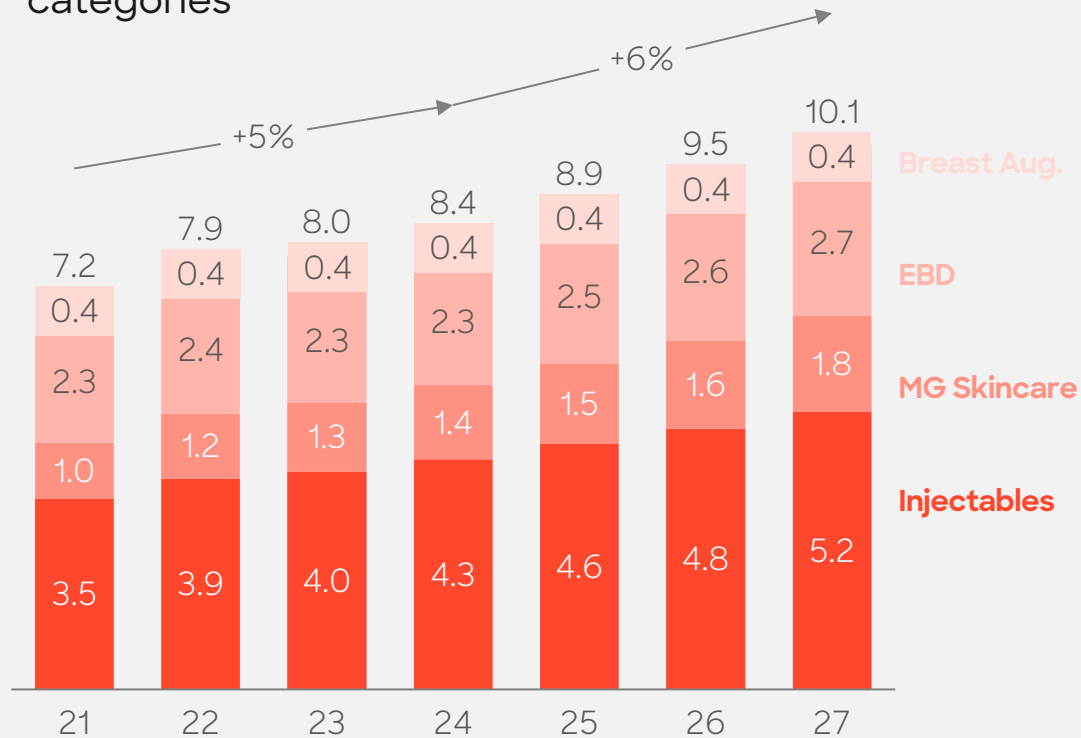
- The skin as a highly visible indicator of health and beauty in social media and professional settings, which drives consistent investment in aesthetic treatments

Source: BCG

# ABS-201 represents a new category of injectable therapy for large consumer-driven market with a wide-range of providers

## Injectables is the largest category of Aesthetics

Injectables growing at ~6% CAGR based in legacy categories

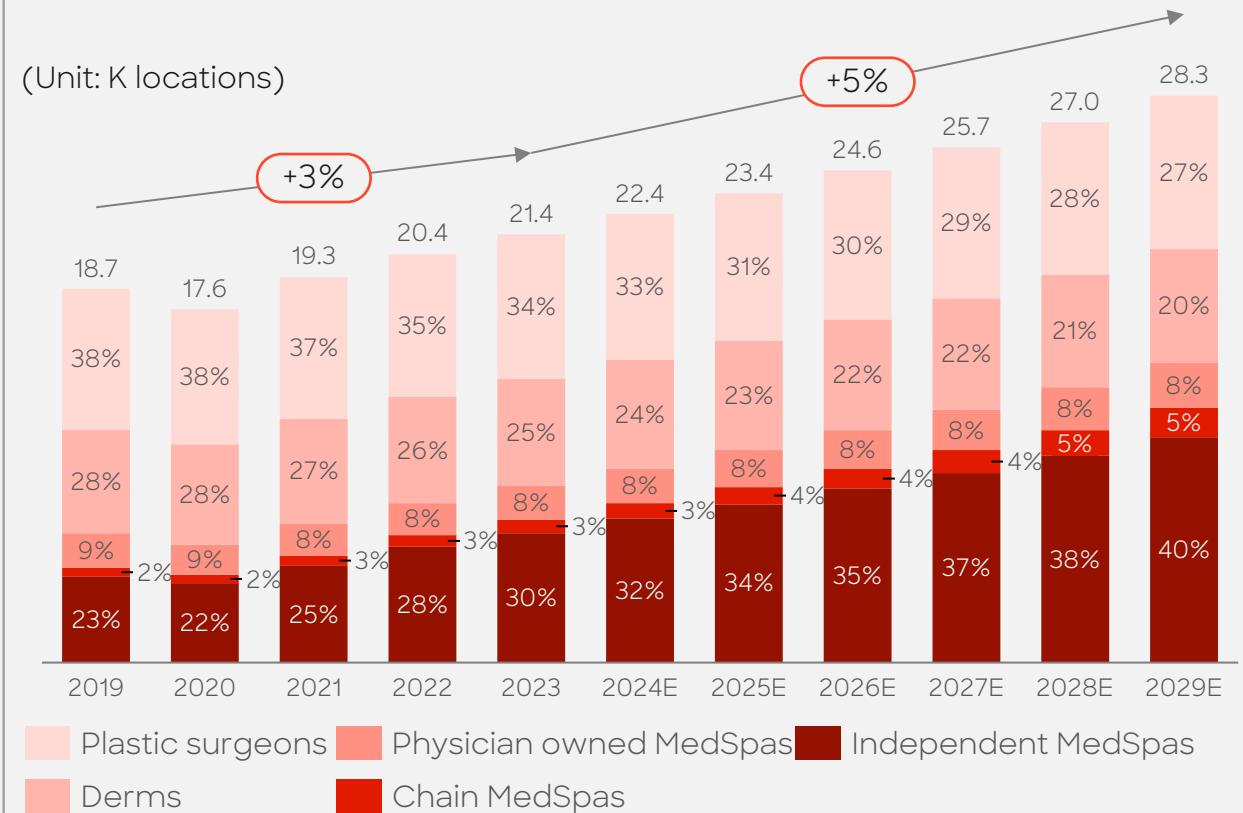


US Market outlook (\$Bn)

Source: IBIS; AmSpa

## Number of Core Medical Aesthetics Locations

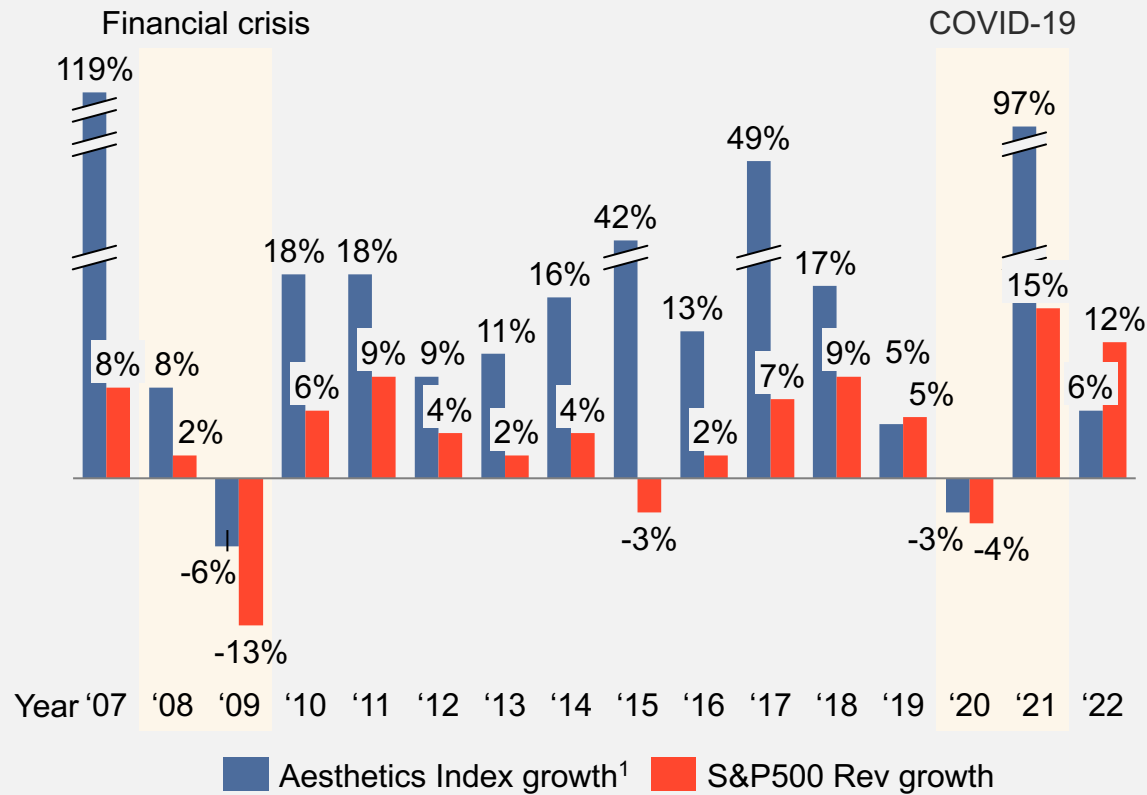
(Unit: K locations)





# ABS-201 as a new category fits into the rapidly growing and resilient consumer demand for appearance and wellness products

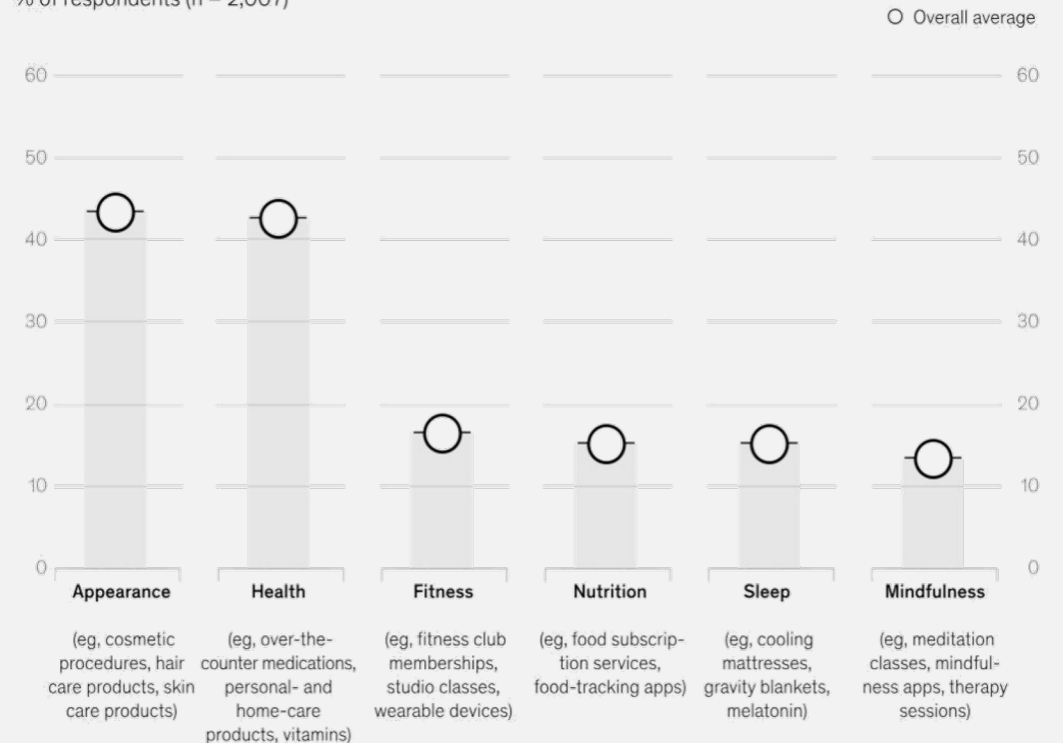
## IN PAST DOWNTURNS, THE AESTHETICS SECTOR HAS PROVEN MORE RESILIENT THAN THE BROADER MARKET<sup>1</sup>



1. Aesthetics Index is built using the weighted average revenue growth of Allergan Ax, Evolus, InMode, Cutera, Establishment labs.  
 2. Source: BCG Consumer Survey 2023, Expert interviews.

## SIGNIFICANT CONSUMER SPEND ON APPEARANCE & HEALTH

US health and wellness purchases, by product/service type and generation,<sup>1</sup>  
 % of respondents (n = 2,007)



<sup>1</sup>Average across all products in each category. Percentage of respondents who purchased at least once in past 12 months.  
 Source: McKinsey Future of Wellness Survey, Aug 2023

McKinsey & Company

# ABS-201's profile addresses massive market demand

**U.S. AESTHETICS INJECTABLES MARKET: \$4.3B**  
(e.g. Fillers and Botox)

**60M CONSUMERS CONCERNED ABOUT LINES/WRINKLES**

↓ 11.7% elect injectable therapy

**7M PATIENTS TREATED /yr**

**\$4.3B per year (OEM)**  
growing at 6% CAGR

**PROJECTED ABS-201 U.S ANDROGENIC ALOPECIA MARKET: \$7-14B**

Model uses aesthetic injectables conversion metrics as a proxy to predict annual ABS-201 treatments

**80M AMERICANS WITH ANDROGENIC ALOPECIA**

↓ 11.7% elect injectable therapy

**9M PATIENTS TREATED /yr**

**\$7-14B per year\***

\* Projections depend on pricing (relative to efficacy) assumptions. Additional upside potential from GLP1 side effects, possible hair color restoration

```
from absci_library import codon_optimizer
library
= codon_optimizer.reverse_translate(library)
library.to_csv("covid-antibody-designs.csv")
library.to_wet_lab(assay="ACE")
```

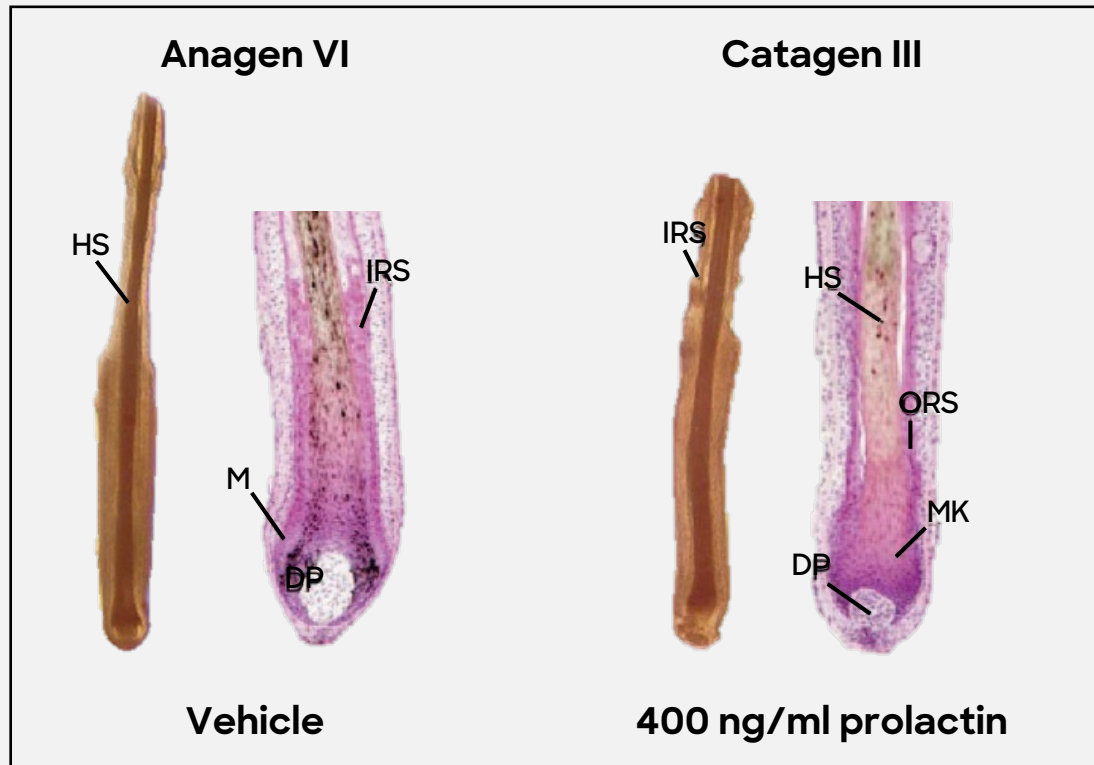
# ABS-201 DRUG CREATION AND DEVELOPMENT

CHRISTIAN STEGMANN, PHD  
SVP, DRUG CREATION

```
from absci import lead_opt_model
lead_optimizer = lead_opt_model.load_latest_model()
library.naturalness =
lead_optimizer.naturalness(library)
lead_optimizer.optimize(library).to_wet_lab(assay="SPR")
```

## Prolactin impacts on organ-cultured human hair follicles

### Prolactin-drives hair follicle regression in human *ex vivo* culture



Prolactin prematurely induces a catagen-like stage in organ-cultured human hair follicles<sup>1</sup> characterized by:

- Condensed shape of the dermal papilla
- Diminishment of the hair matrix volume
- Apparent cessation of pigmentation
- Inhibition of hair shaft elongation

Human genetic evidence suggests no safety liabilities targeting PRLR<sup>2</sup>

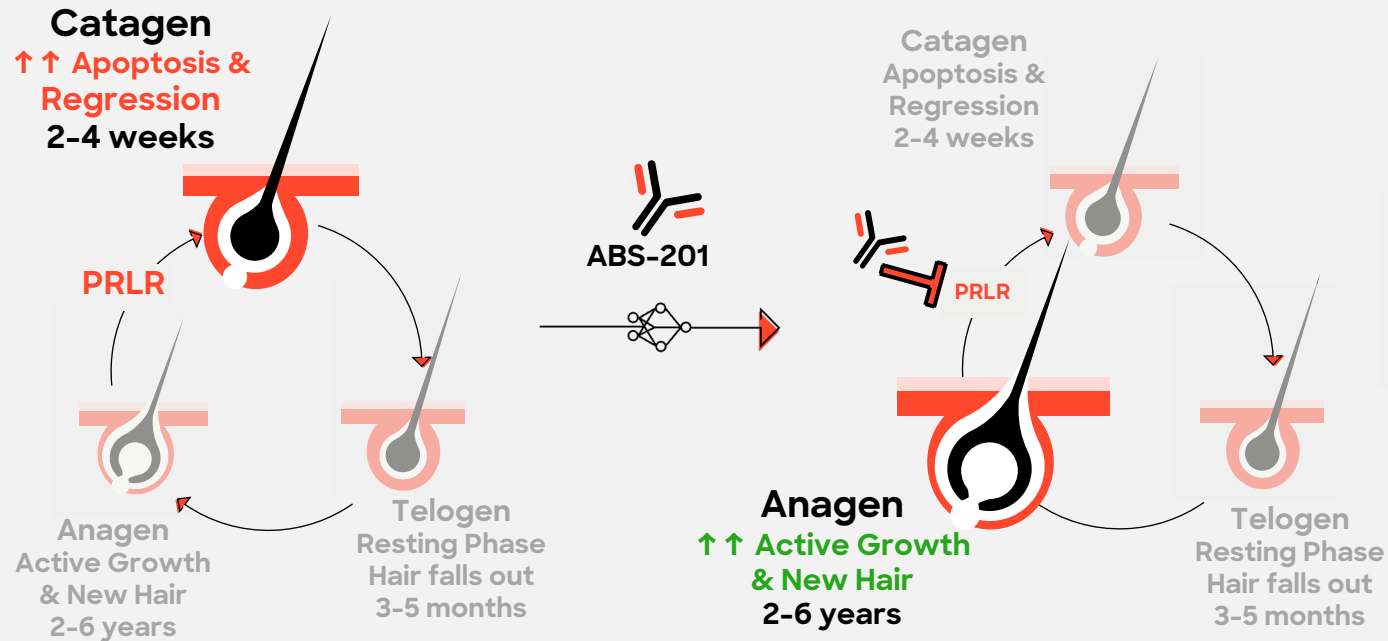
<sup>1</sup>doi: 10.2353/ajpath.2006.050468

<sup>2</sup>doi: 10.1056/NEJMoa1805171

## ABS-201, PRLR

# PRLR inhibition as a safe innovative alternative to current treatment options

## Proposed impact of ABS-201 on Hair Cycle Stages



## ABS-201 has the potential to:

- Shift the balance in hair cycle stage towards anagen phase<sup>1,2</sup> with:
  - active and new hair growth
  - prevention of telogen effluvium
- Promote a long-lasting effect after treatment cessation
- Prevent prolactin mediated telogen effluvium<sup>1,2</sup>
- Restore hair pigmentation<sup>2</sup>

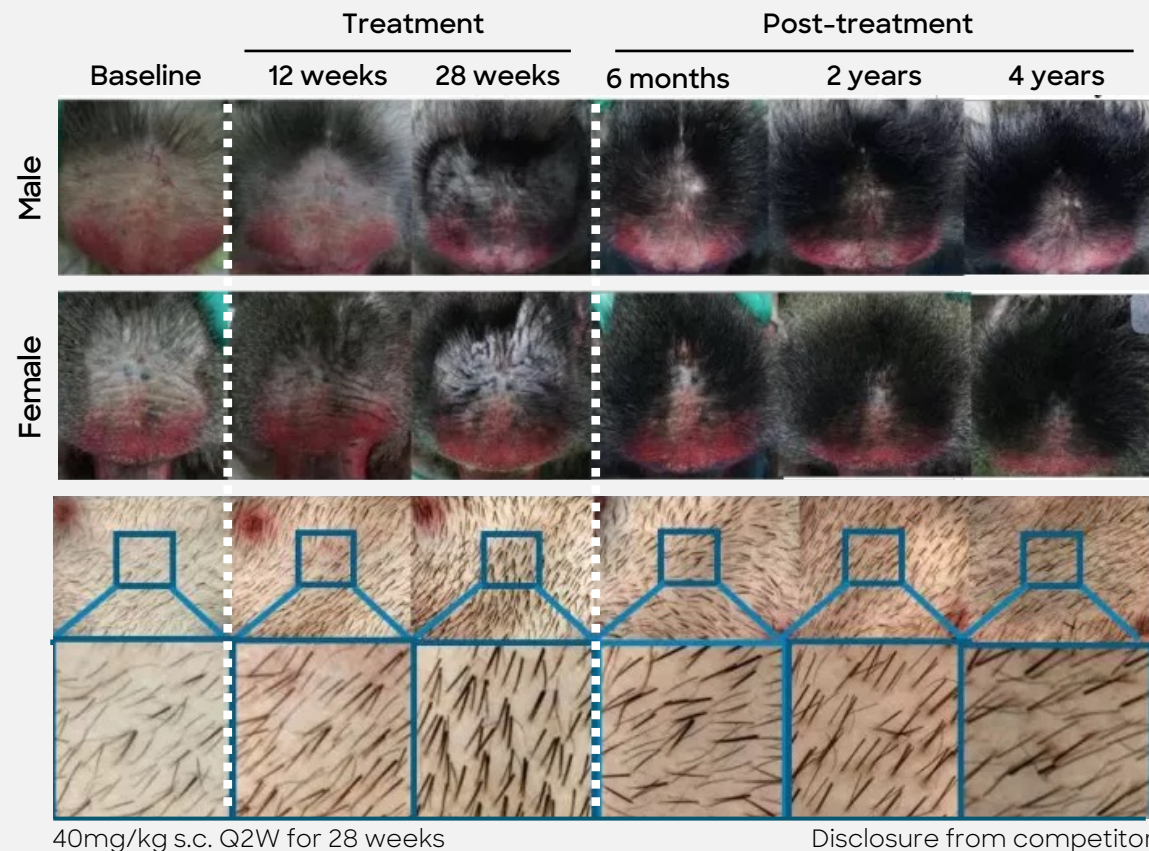
<sup>1</sup>doi: 10.1016/S0002-9440(10)64295-2

<sup>2</sup>doi: 10.2353/ajpath.2006.050468

## PRECLINICAL VALIDATION

# Treatment with an anti-PRLR mAb promotes and sustains long-term hair growth in NHP

### Top head view stump-tailed macaque phenotypic change over time



- Hair density & thickness improved with short treatment duration in primate model of androgenic alopecia
- Hair growth remains several years post cessation
- Hair regrowth observed for both male and female animals

**ABS-201, PRLR**

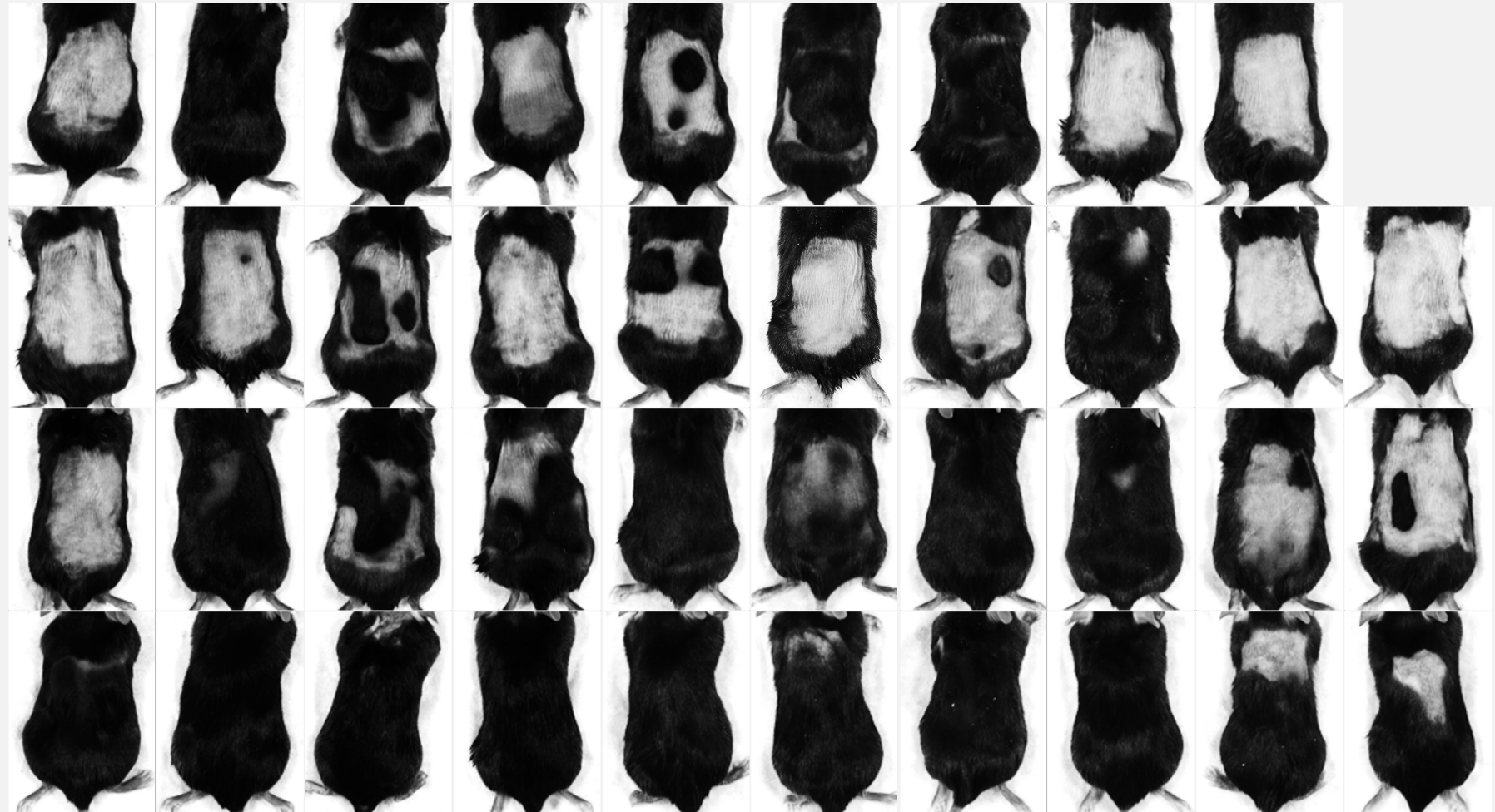
**ABS-201 shows superior efficacy vs. 5% topical minoxidil in 25d hair regrowth model**

**Untreated**  
(n=9)

**Isotype**  
30mg/kg i.p. 1QW  
(n=10)

**Minoxidil 5%**  
Topical once  
daily (n=10)

**ABS-201-A**  
30mg/kg i.p. 1QW  
(n=10)



## Excellent developability profile of ABS-201 development candidates

Desired attribute	Developability assessment	HMI-115#	ABS-201-A	ABS-201-B
High solubility	Solubility screening at various buffers <sup>1</sup>	Low	Great	Good
	Diffusion interaction parameter (high-concentration predictor) <sup>2</sup>	Low	Great	Great
Prolonged stability	Acidic stress forced degradation <sup>*,3</sup>	Affected	Not affected	Not affected
	Freeze-and-thaw susceptibility <sup>&amp;,4</sup>	Affected	Not affected	Not affected

# Estimated performance of a putative clinical competitor

\* Samples at 50 mg/mL incubated at pH ≈ 2.5, 25°C for up to 3 days.

& Samples at 50 mg/mL and subjected up to 5 cycles of freeze-and-thaw.

1. Low, good and great as <20%, 20% and >20% PEG solubility, respectively

2. Low, good and great as <15.0, 15.0-20.0, and >20.0 k<sub>D</sub>, respectively

3. Affected as loss of purity by NR-CGE > 4.0%

4. Affected as observed purity loss and high-molecular species formation by SEC



## Potential best-in-class PRLR antibody designed using generative AI



- High affinity and potency
- Delivery of promising candidates in just over 1 year
- Excellent developability profile enables high-concentration formulation and great stability
- Anticipated low immunogenicity
- Extended half-life and expected longer dosing intervals

## Continued progress with FiH expected in 2026

### • 4Q 2024

AI-designed development candidates

- ✓ High affinity
- ✓ High potency
- ✓ Favorable manufacturability
- Long half-life



### • INITIATED DEC 2024

IND-enabling studies



### • 1H 2026

FiH clinical development



## INTERNAL PIPELINE

# Absci's progress in Drug Creation

### > Continued advancement of lead assets

#### ABS-101

Continued progress of TL1A asset with FiH in 1H 2025

- New preclinical data supporting superior immunogenicity profile
- Phase 1 Interim data readout in 2H 2025

#### ABS-201

Development Candidate for PRLR (prolactin receptor) nominated early December 2024

IND-enabling activities initiated

### > Discovery of next assets

#### ABS-301

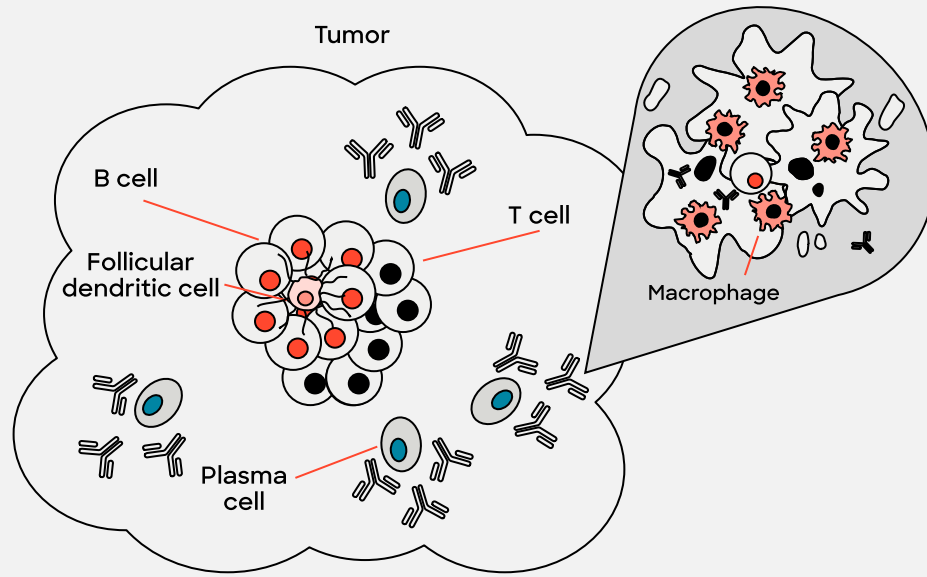
Progress of first-in-class asset with target validation and initial preclinical efficacy readouts in 1H 2025

#### NEW: ABS-501

Nomination of a potential best-in-class HER2 asset

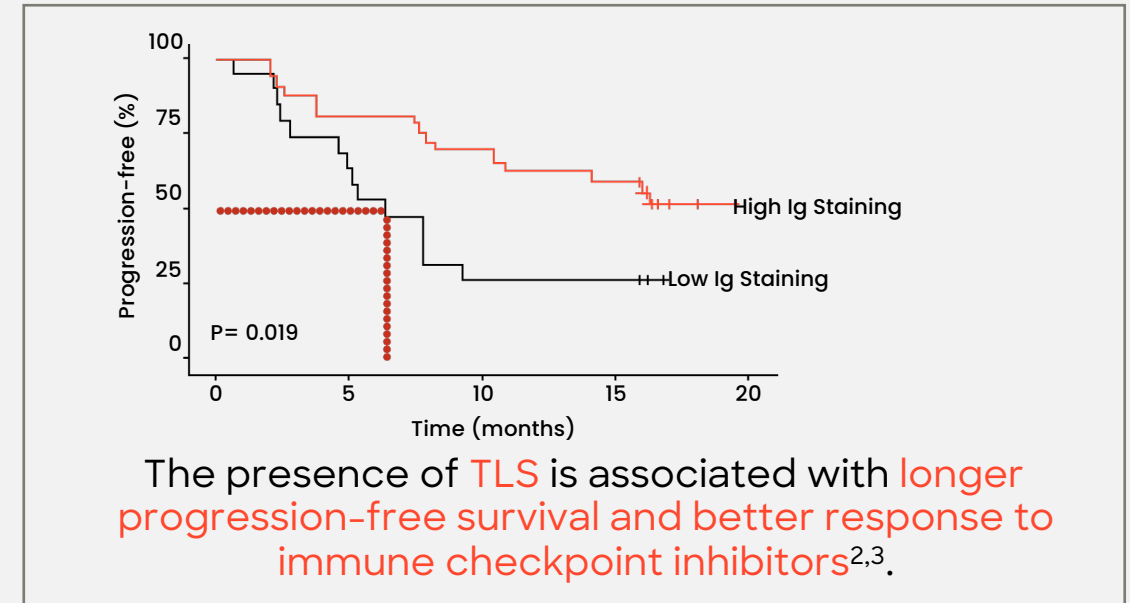
## TARGET DISCOVERY

# Tertiary Lymphoid Structures (TLS): The focus of Absci's Reverse Immunology approach



Tertiary lymphoid structures (TLS) are centers of immune activity, such as B-cell proliferation and antibody production, that develop in chronically inflamed tissues<sup>1</sup>.

Antibodies from TLS are specialized for local antigens and play a significant role in the progression of chronic diseases and cancer, setting them apart from the general population of antibodies in the peripheral blood<sup>2</sup>.



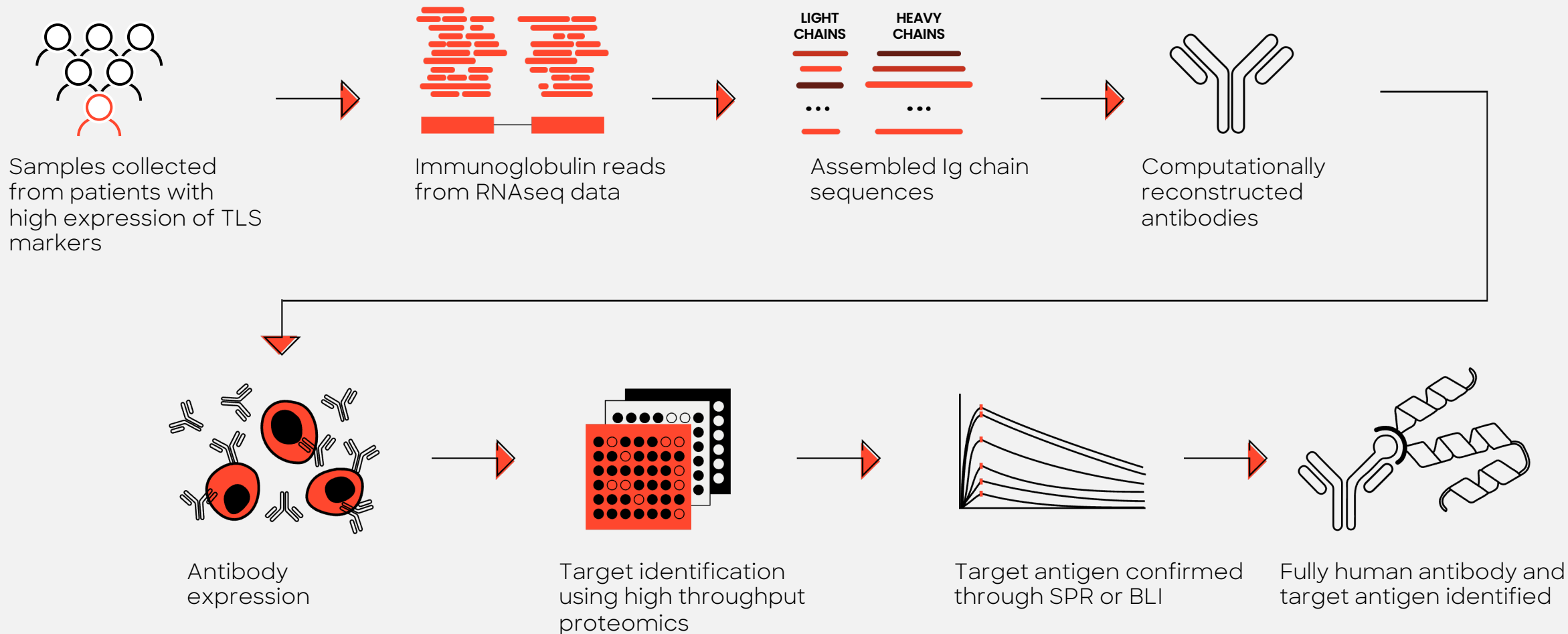
- Rapidly growing evidence illustrates correlation between TLS-derived antibodies in the tumor microenvironment and positive clinical outcomes<sup>2</sup>.
- TLS-derived antibodies have been shown to be associated with apoptosis of cancer cells in patients<sup>2</sup>.

<sup>1</sup> doi: 10.3389/fimmu.2018.01952

<sup>2</sup> doi: 10.1016/j.immuni.2022.02.001

<sup>3</sup> doi: 10.1038/s41586-019-1922-8

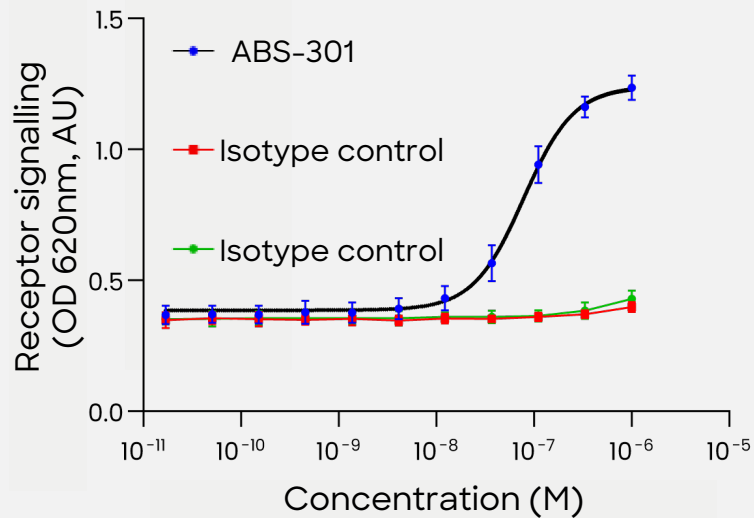
# ABS-301 | Reverse Immunology platform identifies the antigens targeted by endogenous antibodies produced in tumor lymphoid structures, TLS



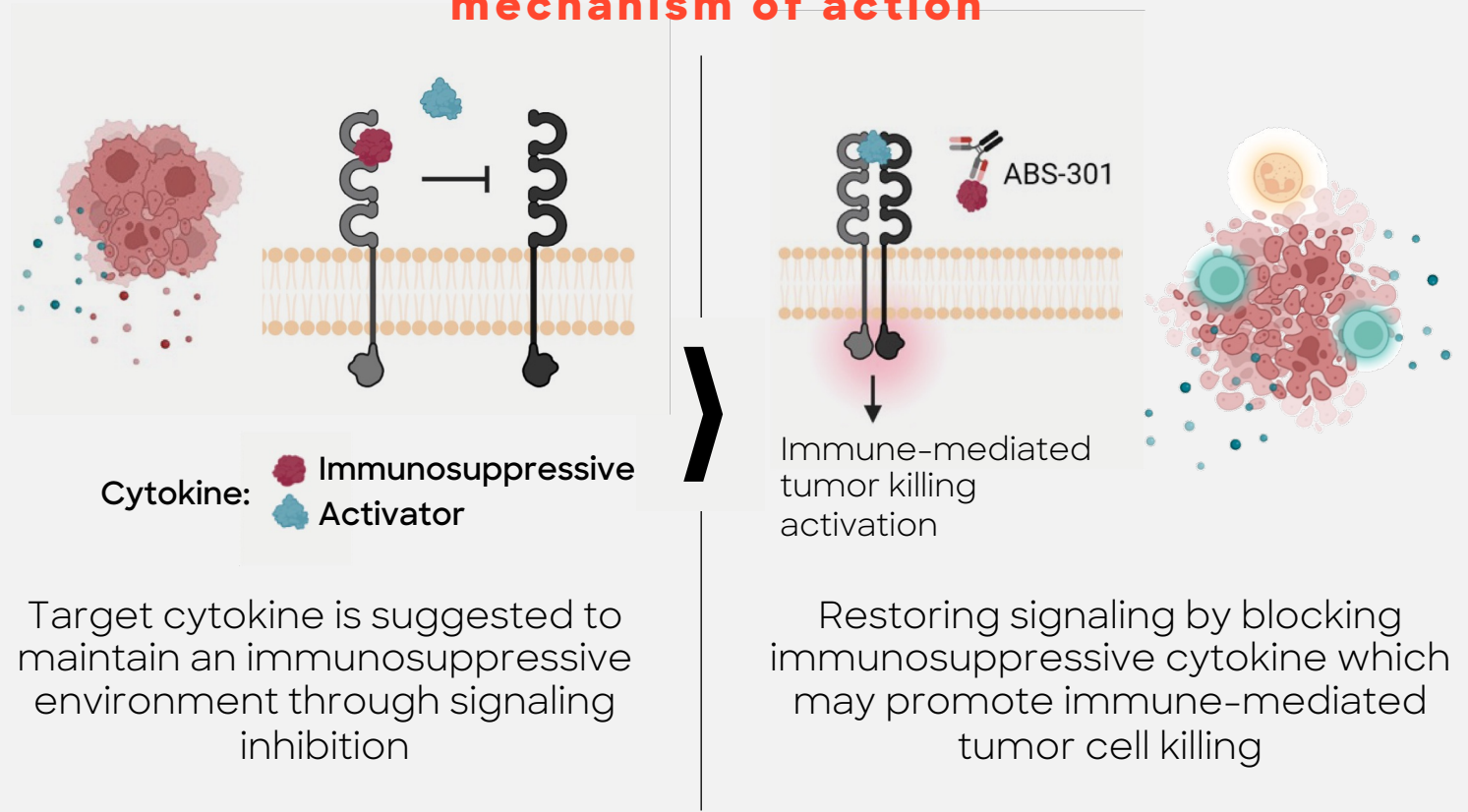
Reference, doi: 10.1101/2021.02.06.430058

# ABS-301 | A patient-derived antibody discovered by reverse immunology blocks an immunosuppressive cytokine

## ABS-301 rescues pro-inflammatory signaling through inhibition of immunosuppressive cytokine



## Target biology and proposed ABS-301 mechanism of action





## LUIS DIAZ JR., MD

HEAD OF THE DIVISION OF SOLID TUMOR ONCOLOGY | MSKCC

BOARD DIRECTOR | QUEST DIAGNOSTICS

SCIENTIFIC ADVISORY BOARD MEMBER | ABSCI

**Professional Expertise:** Grayer Family Chair and Head of the Division of Solid Tumor Oncology in Memorial Sloan Kettering Cancer Center’s Department of Medicine.

**Research Contributions:** Lab focused on developing novel tools and approaches for the diagnosis and treatment of cancer using genomics as a guide. Developed liquid biopsies for cancer detection, early diagnostic tools for ovarian and endometrial cancer, and led pembrolizumab study, securing the first FDA approval based on a biomarker rather than tumor location.

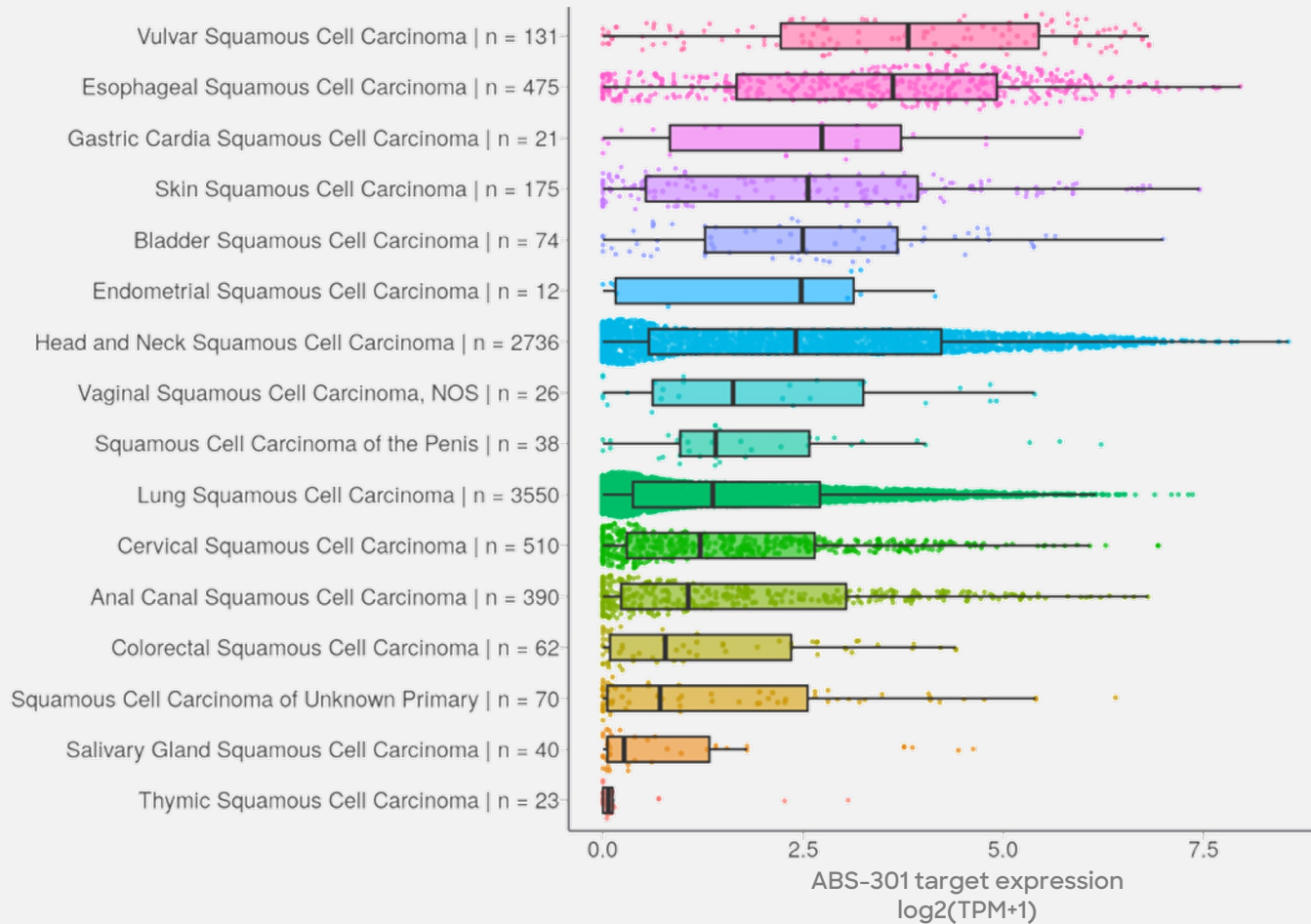
### Select Awards & Honors:

- Elected Member, National Academy of Medicine (2023)
- National Cancer Advisory Board (appointed by President Biden, 2021)

### Education & Training:

- MD, University of Michigan.
- Residency: The Johns Hopkins School of Medicine.
- Fellowship: The Johns Hopkins School of Medicine

# Expression of ABS-301's target suggests broad potential in squamous cell carcinomas



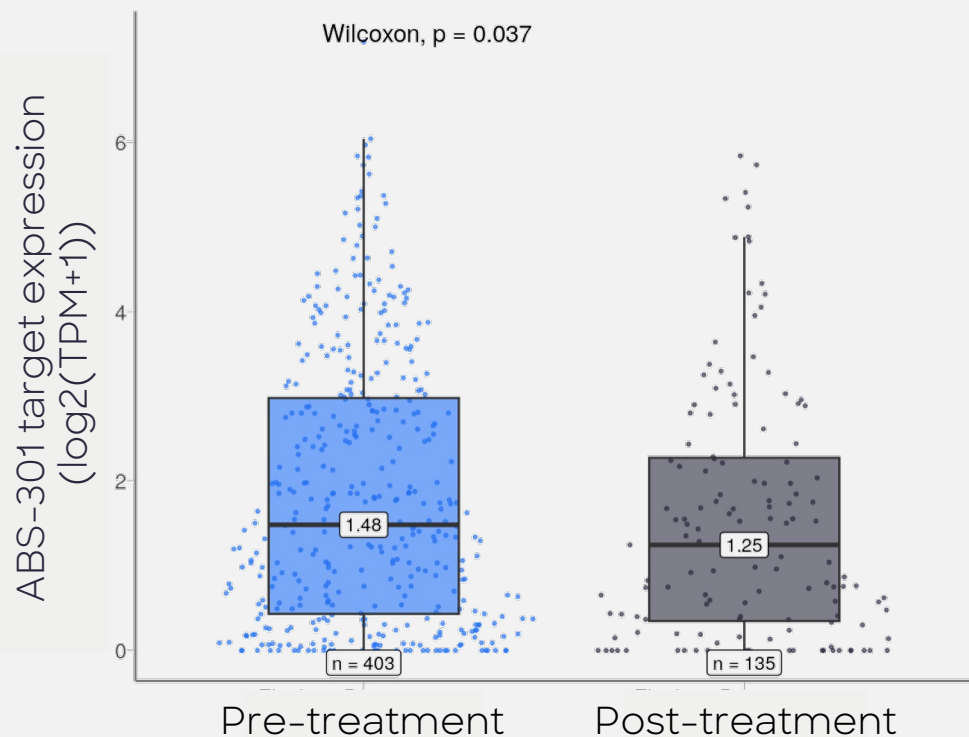
**Distribution of ABS-301 target expression across squamous cell carcinoma cohorts.**

Values shown are  $\log_2(\text{TPM}+1)$  normalized. Multiple biopsies from a patient are included in the analysis.  
Source: Tempus



# ABS-301 | Expression in Lung Squamous Cell Carcinoma (LUSC): no change with treatment and strong negative correlation with CD8+ T cell infiltration

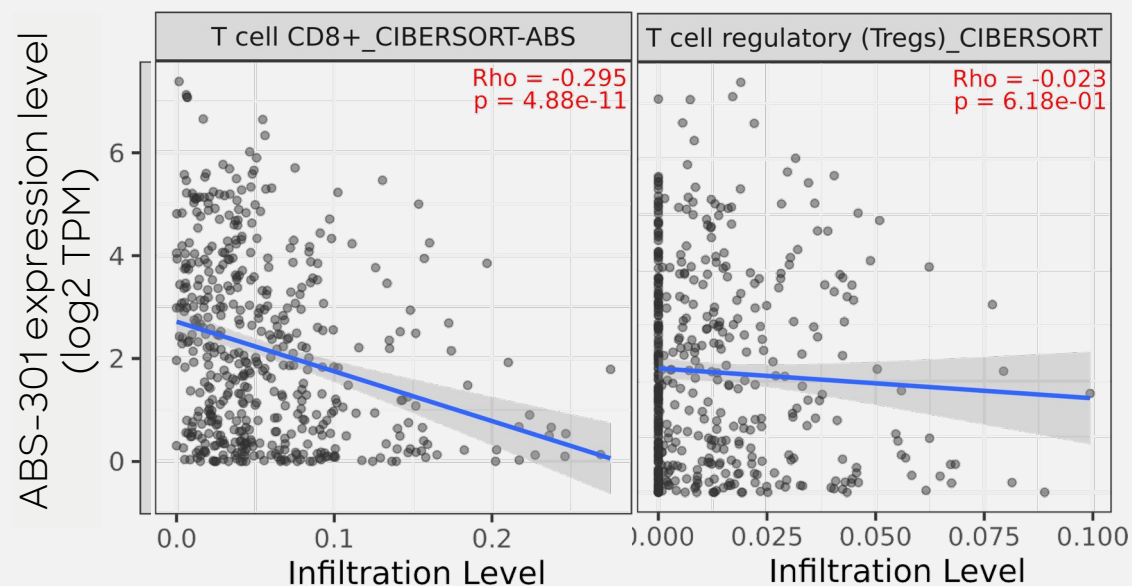
## Sustained target expression in LUSC



Source: Tempus

In LUSC, univariate analysis of ABS-301 expression indicate only a minor change in expression between pre- and post-treatment suggesting opportunity for combination therapy.

## CD8+ Infiltration negatively correlated with target expression in LUSC



Source: TCGA

ABS-301 target expression shows a strong negative correlation with CD8+ T cell infiltration with a minimal effect on Treg infiltration supporting immunosuppressive activity of target *in vivo*.

## ABS-301 | Broad potential in immuno-oncology

Based on literature and potential competitive molecules, the following indications could be of interest:

Indication	US Prevalence	Estimated 5-year survival rate*	US Sales in 2030
NSCLC	Calculated: ~202K in 2023	28%	\$27B
	SCC 30% of NSCLC cases Calculated: ~61K	24%	Calculated Sales: \$8.1B
Head and Neck SCC	~54K in 2022	68.5%	Calculated Sales: \$2.3B
Esophageal Cancer	~21K in 2022	20%	\$1.5B
	SCC ~20% of cases Calculated: ~4.2K		Calculated Sales: \$0.3B
Cervical Cancer	~14K in 2023		\$0.6B
	SCC 90% of cases Calculated: ~13K	67%	Calculated Sales: \$0.6B
Skin Cancer, non-melanoma	Incidence = ~3,300K	95-100%	\$1.0B
	SSC Incidence = ~700K	95%	Calculated Sales: \$0.2B

\*dependent on stage of diagnosis  
References provided in appendix

## INTERNAL PIPELINE

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### > Discovery of next assets

#### ABS-301

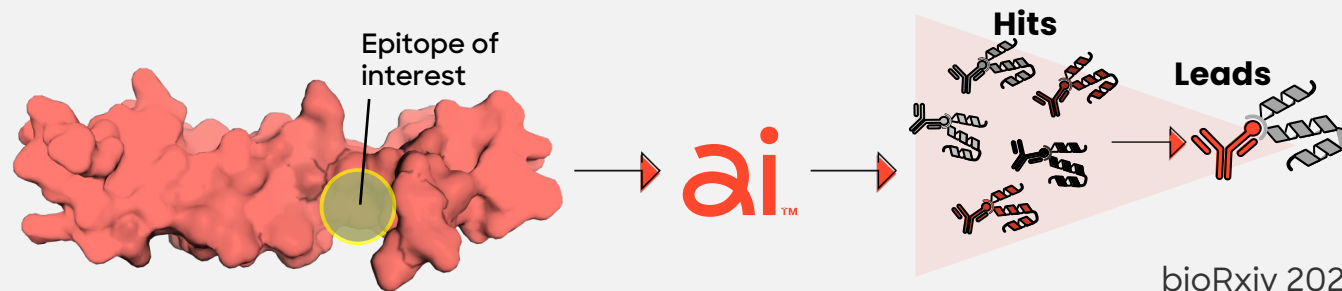
Progress of first-in-class asset with target validation and initial preclinical efficacy readouts in 1H 2025

#### NEW: ABS-501

Nomination of a potential best-in-class HER2 asset

# ABS-501, HER2 | Deploying *de novo* AI model on HER2 led to discovery of antibodies displaying molecular interactions distinct from trastuzumab

## Zero shot *de novo* AI discovery on HER2



- Hits with edit distance of up to 12 amino acids in HCDR3 region (13 aa, search space of  $20^{13}$ ) were screened
- Selected 50 hits with  $<10$  nM affinity were expressed as mAbs for binding affinity determination
- Top 11 antibodies were characterized *in vitro* and 3 leads evaluated *in vivo*

bioRxiv 2023.01.08.523187; doi: <https://doi.org/10.1101/2023.01.08.523187>

## AI-designed antibodies: same epitope, different HER2 contact preferences

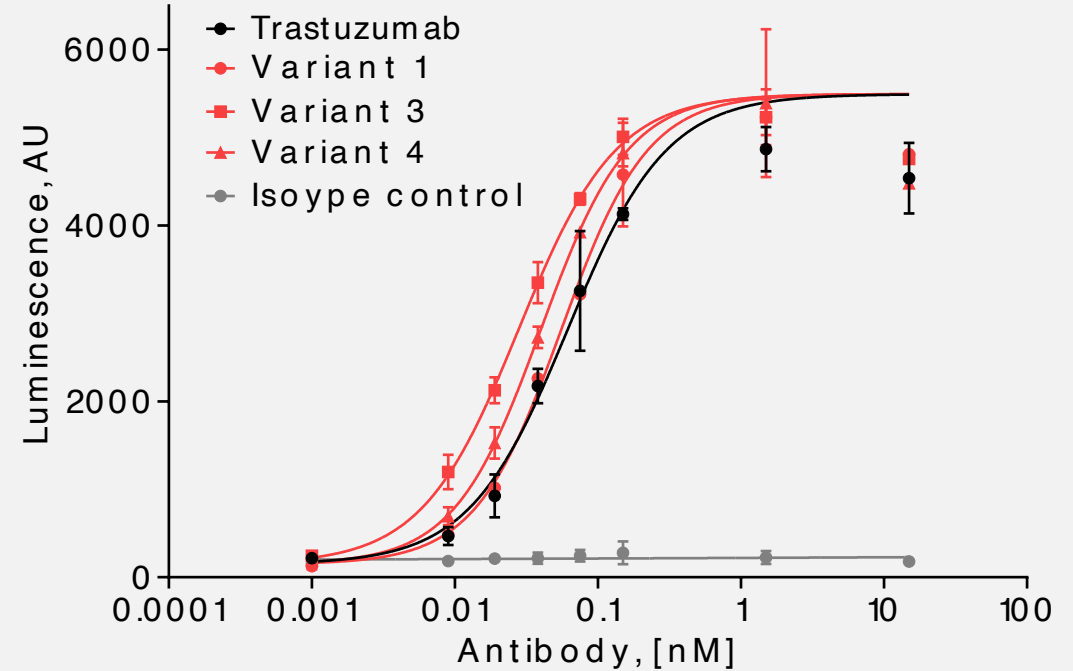
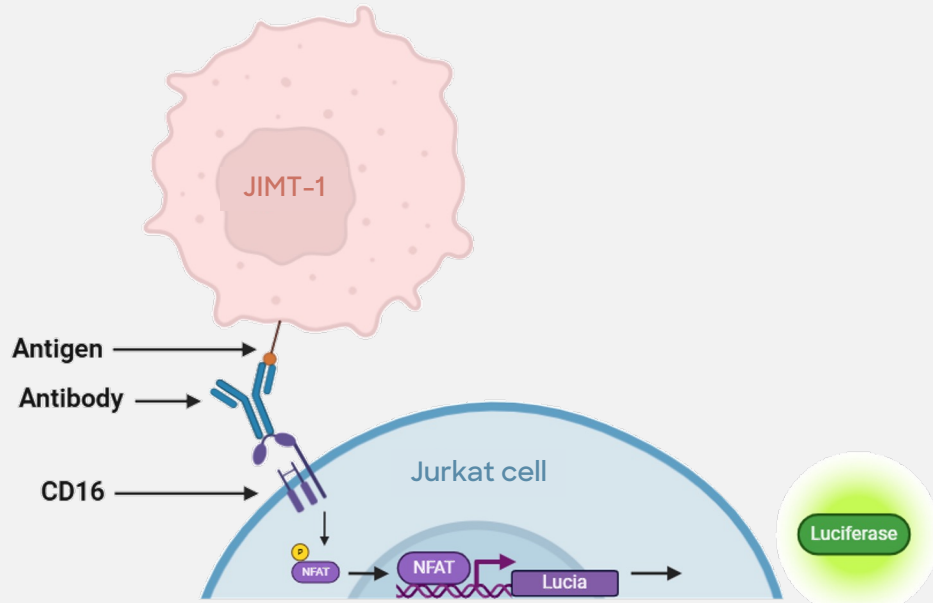
Variant #	Edit distance	K <sub>D</sub> (nM)	Epitope mapping view	Loop 581-590
Trastuzumab	0	1.07		
1	7	4.16		
3	7	9.75		
4	2	6.66		

- Not critical
- Partial
- Critical

# ABS-501, HER2 | AI-designed antibodies demonstrate measurable enhancement of ADCC activity compared to trastuzumab

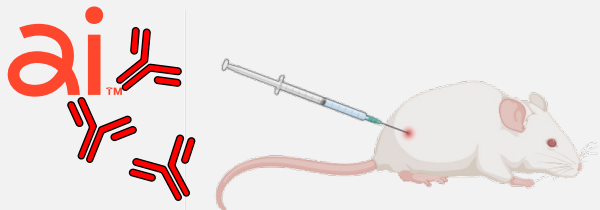
## ADCC assay principle

Luciferase signal driven by NFAT transcription factor positively correlates to ADCC activation against JIMT-1

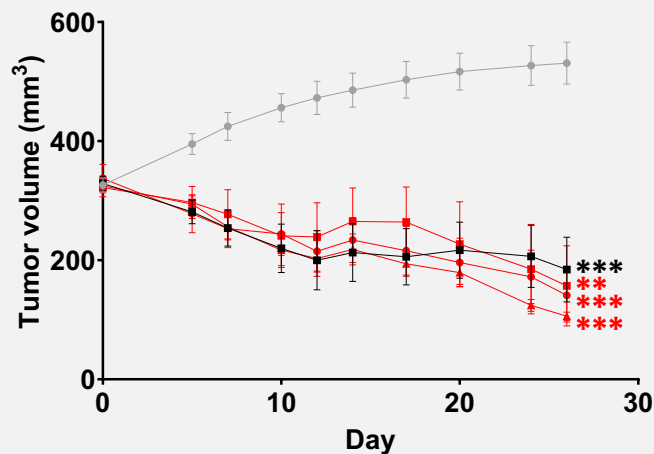


	Trastuzumab	Variant 1	Variant 3	Variant 4
EC50 (nM)	0.062	0.056	0.028	0.040
R squared	0.93	0.97	0.97	0.95
P value	N/A	Not significant	<0.0001	0.0015

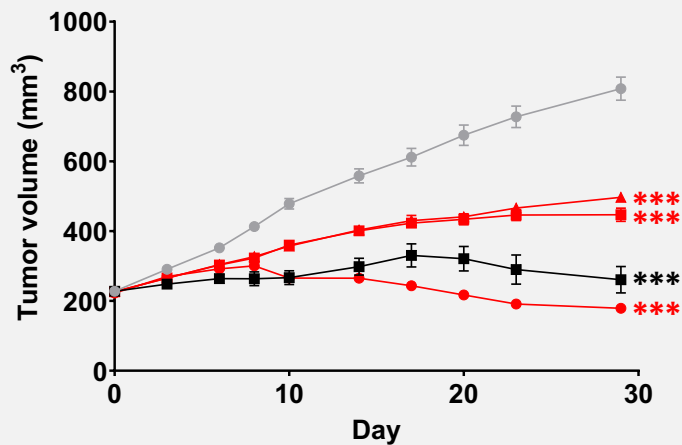
# ABS-501, HER2 | AI-designed antibodies suppress growth of trastuzumab-sensitive & resistant HER2+ breast tumors



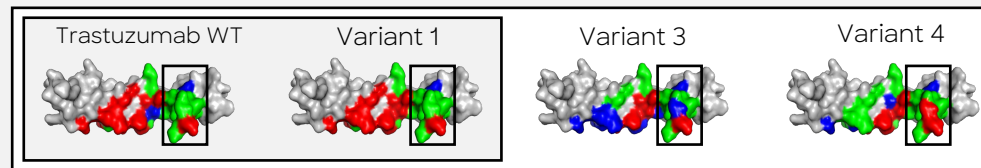
Mouse xenograft model using **EFM192A** (HER2+ BC; **Tz sensitive**)



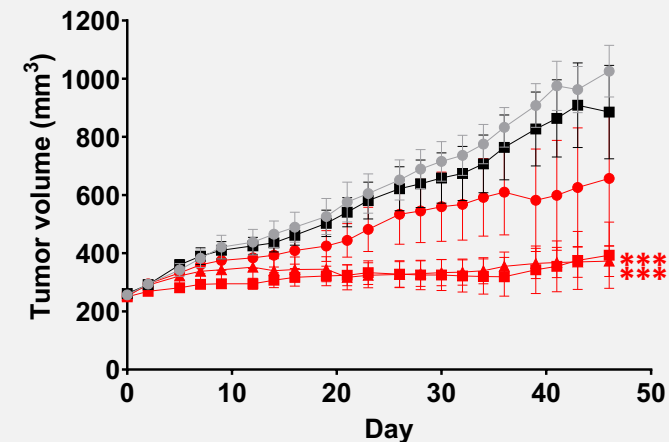
Mouse xenograft model using **MDA-MB-361** (HER2+ BC; **Tz sensitive**)



■ Not critical ■ Partial ■ Critical



Mouse xenograft model using **JIMT-1** (HER2-amp BC; **Tz resistant**)



● Isotype control ● Trastuzumab ● Variant 1 ■ Variant 3 ▲ Variant 4

Trastuzumab-sensitive EFM192A and MDA-MB-361 tumors respond to both trastuzumab (Tz) & AI-designed antibodies

JIMT-1 tumors are trastuzumab resistant but sensitive to variants 3 and 4

Xenograft studies conducted by Dr. Dennis Slamon's team at UCLA

2-way ANOVA \*\* P<0.001 and \*\*\*P<0.0001 vs isotype control

# ABS-501, HER2 | AI-designed variants create opportunities to address unmet medical need

## Currently exploring breast cancer as opportunity: alternative to or post Enhertu®

› Despite Enhertu's good efficacy, leading oncologists are only moderately satisfied due to toxicity (e.g. interstitial lung disease); less toxic therapy and effective treatment post-Enhertu are key unmet needs.



*"Post-Enhertu is really where the action is right now in the field. I think the first company that comes up with something that has significant benefit in Enhertu progressive disease is going to win." - KOL*

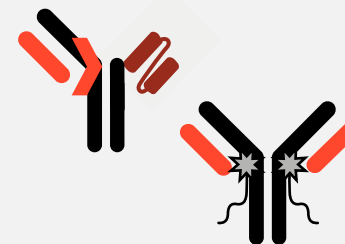
## Multiple paths possible for therapeutic development:

› Modality switch or combination opportunities under consideration to address unmet medical needs



Later-line treatment regimens for HER2-positive cancer:

- Monotherapy
- Combination therapy with targeted small molecules



Enhancing efficacy and expanding indications (e.g. Enhertu resistance):

- Antibody-drug conjugates (ADCs)
- Multi-specific antibodies



## DENNIS SLAMON MD, PHD

PROFESSOR OF MEDICINE AND CHIEF OF THE DIVISION OF HEMATOLOGY/ONCOLOGY | UCLA  
EXECUTIVE VICE CHAIR FOR RESEARCH | UCLA DEPARTMENT OF MEDICINE  
DIRECTOR OF CLINICAL/TRANSLATIONAL RESEARCH AND DIRECTOR OF THE REVLON/UCLA  
WOMEN'S CANCER RESEARCH PROGRAM

**Research Contributions:** Pioneering research that identified the HER2/neu oncogene that is amplified in 25–33% of breast cancer patients which led to the development of the breast cancer drug trastuzumab


### Select Awards & Honors:

- 2024 Szent-Györgyi Prize
- 2002 Jacob Heskell Gabbay Award in Biotechnology and Medicine
- 2007 Gairdner Foundation International Award
- 2017 Komen Brinker Award for Scientific Distinction
- 2019 Lasker-DeBakey Clinical Medical Research Award
- 2019 Sjöberg Prize

### Education & Training:

- MD, University of Chicago
- Residency: Internal Medicine, University of Chicago Hospitals
- Fellowship: UCLA School of Medicine





Slamon, Dennis J.

```
from absci_library import codon_optimizer
library = codon_optimizer.reverse_translate(library)
library.to_csv("covid-antibody-designs.csv")
library.to_wet_lab(assay="ACE")
```

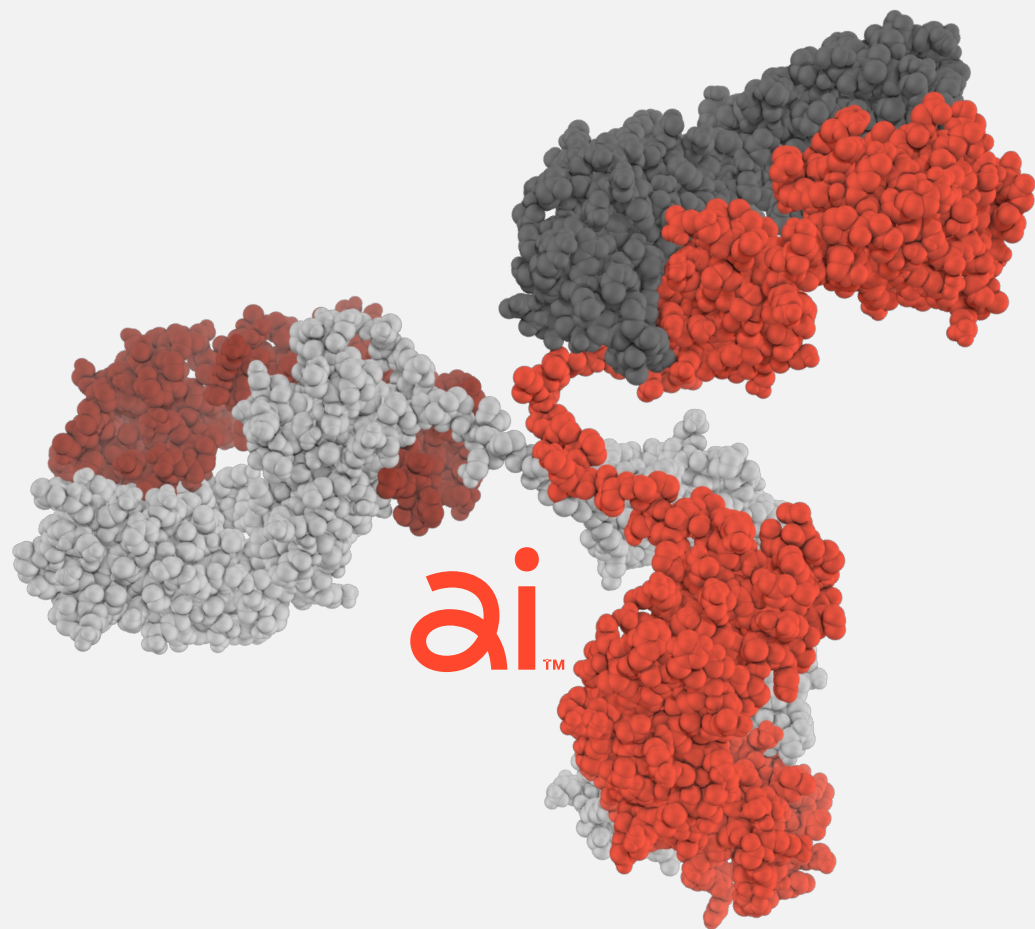
```
from absci import de_novo_model
model = de_novo_model.load_latest()
antigen = model.load_pdb("7olz.pdb", chain="A")
antibodies = model.predict(antigen, N=300000)
```

# ZACH JONASSON, PHD

CHIEF FINANCIAL OFFICER & CHIEF BUSINESS  
OFFICER

```
library = []
for antibody in antibodies:
    naturalness = model.naturalness(antibody.sequence)
    library.append((antibody, naturalness))
```

# We are advancing our AI platform capabilities and applying them to design novel and differentiated assets

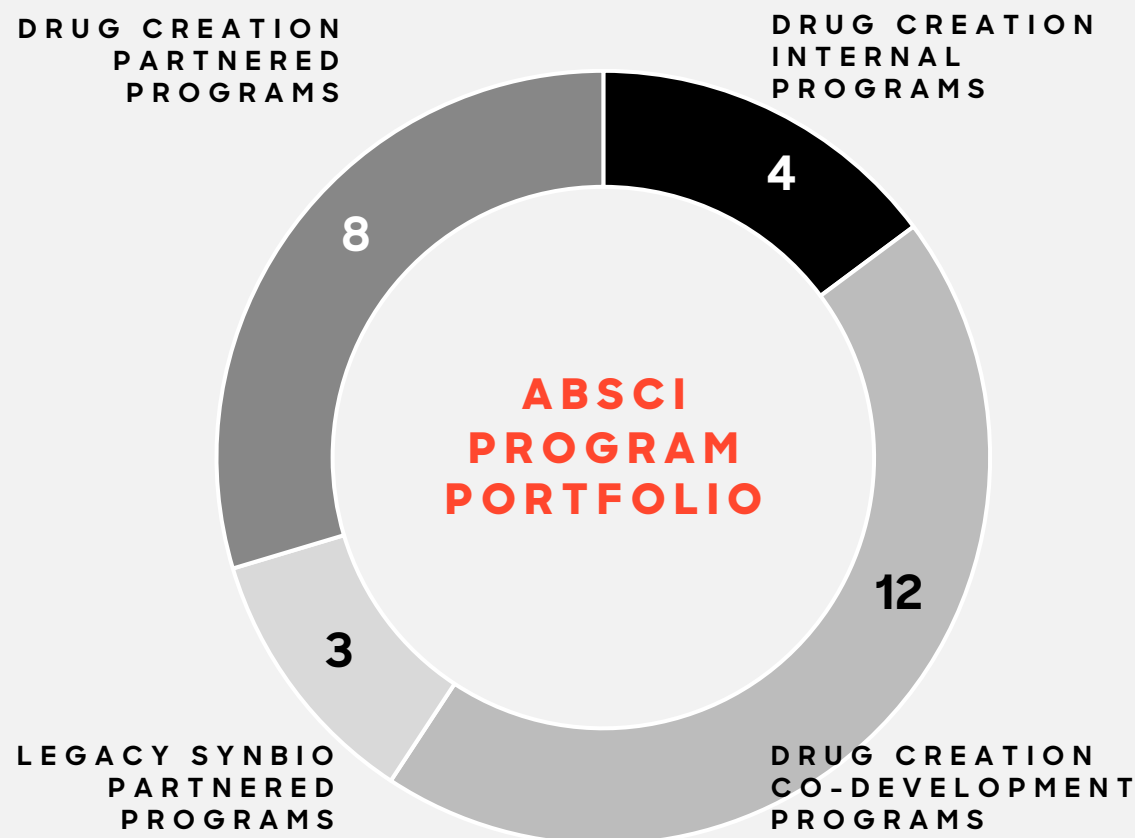


- EPITOPE-SPECIFIC + EPITOPE INTERFACE OPTIMIZATION
- ENHANCED POTENCY AND MOA
- ADDRESS DIFFICULT TARGET CLASSES
- DIFFERENTIATED PHARMACOLOGY FEATURES

# Balanced portfolio of novel and differentiated programs

## PORTFOLIO INCLUDES:

- › Wholly-owned internal Drug Creation Programs targeting multi-billion dollar opportunities
- › Partnered Drug Creation Co-Development Programs targeting multi-billion dollar opportunities
- › Partnered Drug Creation Programs with over \$1.5 B in potential deal value not including royalties
- › Partnered Legacy SynBio Programs



## PARTNERSHIP MODELS

## DEAL STRUCTURE

## EXAMPLE PARTNERS

### Drug Creation Partnerships

- Upfront payments
- R&D payments
- Election fees
- Clinical + commercial milestone payments
- Royalties



### Co-Development Partnerships

- Joint contribution of IP, know-how, technology
- 50/50 development cost
- 50/50 out-license transaction economics
- Opt-out ability



### Internally Developed Assets

- Asset Sale or  
Out-license with
- Upfront payment
  - Clinical and commercial milestone payments
  - Royalties

Partnering Discussions Currently Ongoing for Specific Programs



## KARL ZIEGELBAUER, PHD

CHIEF SCIENTIFIC OFFICER, ALMIRALL

### **Professional Expertise:**

Chief Scientific Officer at Almirall since 2021. Leading strategic partnerships and developing and expanding Almirall's R&D pipeline in Medical Dermatology.

and previously Senior Vice President and Head of Open Innovation & Digital Technologies at Bayer Pharmaceuticals.

Three decades of experience in drug discovery in international markets such as Germany, Japan, and the United States.


### **Research Contributions:**

Co-authored more than 50 scientific publications covering basic research as well as drug discovery topics.

### **Education & Training**

PhD in Biochemistry from the University of Tübingen (Germany).



A man with short, light-colored hair and glasses is sitting in a light-colored office chair. He is wearing a dark blue, long-sleeved button-down shirt. He is smiling at the camera. Behind him is a large, leafy green plant in a dark planter. To the right, there is a modern office cabinet with glass doors on top and orange doors on the bottom. A portion of a yellow table is visible in the bottom right corner. The background wall is a neutral, light color.

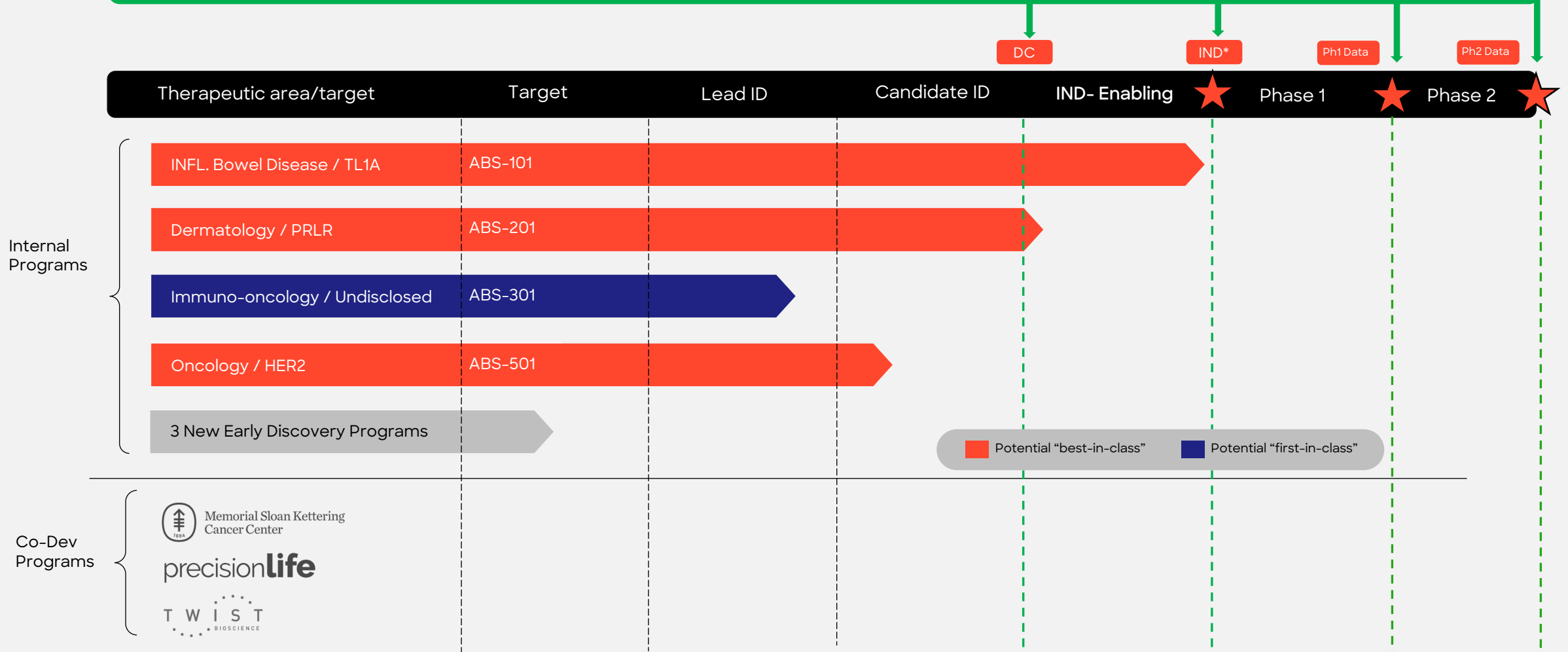
Karl

# PIPELINE

## Internal pipeline and co-development pipeline monetization strategy

**For each pipeline program:**

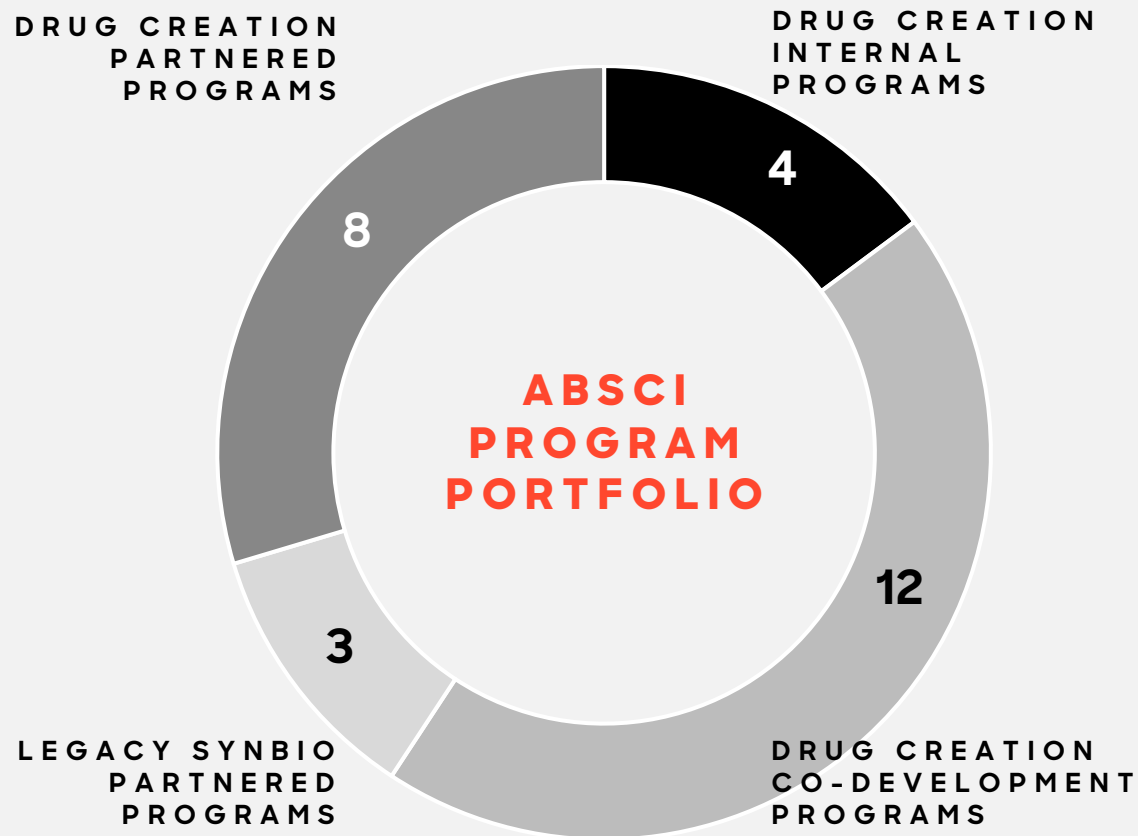
Assess partner/out-license options during development and transact at value inflection points





# VALUE INFLECTION POINTS

## Portfolio provides near-term value inflection points and monetization options



ABS-101 (anti-TL1A antibody) Interim Phase 1 Readout (2H 2025)

ABS-301 (novel immuno-oncology program) Development Candidate Package (1H 2025)

Progress on Milestones from our Existing Drug Creation Partnerships; Addition of New Partnerships with Large Pharma, Biotech, and Others

Co-development partnership program advancements

**4 NAMED INTERNAL, WHOLLY OWNED PROGRAMS**

**23 PARTNERED AND CO-DEVELOPMENT PROGRAMS**

Strong pipeline of interest in our AI Drug Creation Platform

**\$127.1M**

Cash, cash equivalents, and short-term investments as of September 30, 2024, which based on our current plans is expected to fund our operations into the first half of 2027

```
from absci_library import codon_optimizer
library = codon_optimizer.reverse_translate(library)
library.to_csv("covid-antibody-designs.csv")
library.to_wet_lab(assay="ACE")
```

# SEAN MCCLAIN

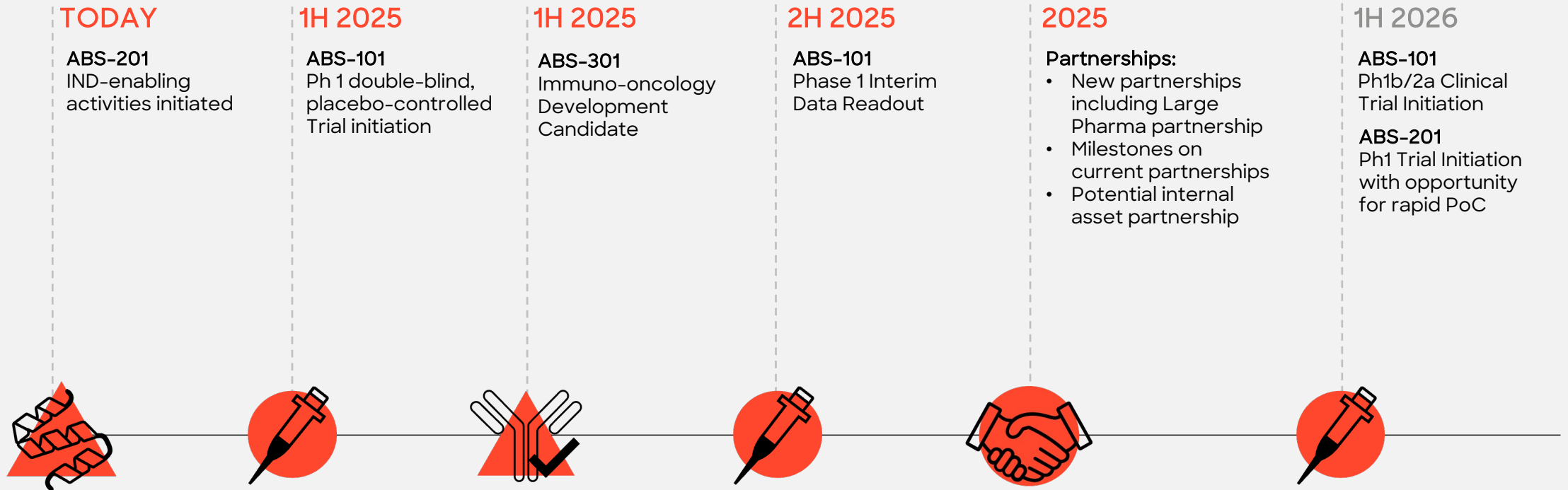
FOUNDER & CEO

```
from absci import de_novo_model
model = de_novo_model.load_latest()
antigen = model.load_pdb("7olz.pdb", chain="A")
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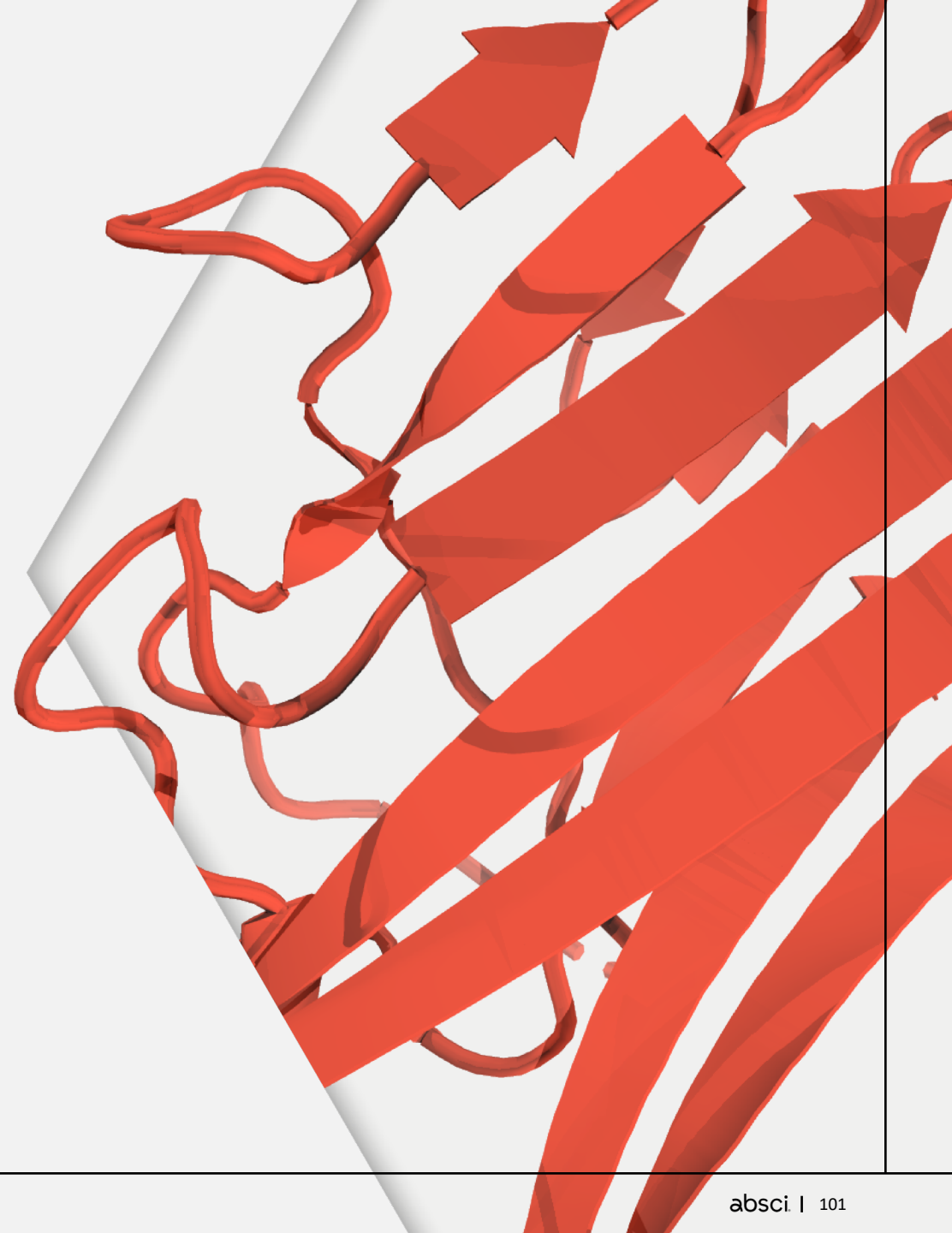
```
library = []
for antibody in antibodies:
    naturalness = model.naturalness(antibody.sequence)
    library.append((antibody, naturalness))
```

## CATALYSTS

# Leading AI platform driving numerous near-term value inflection points

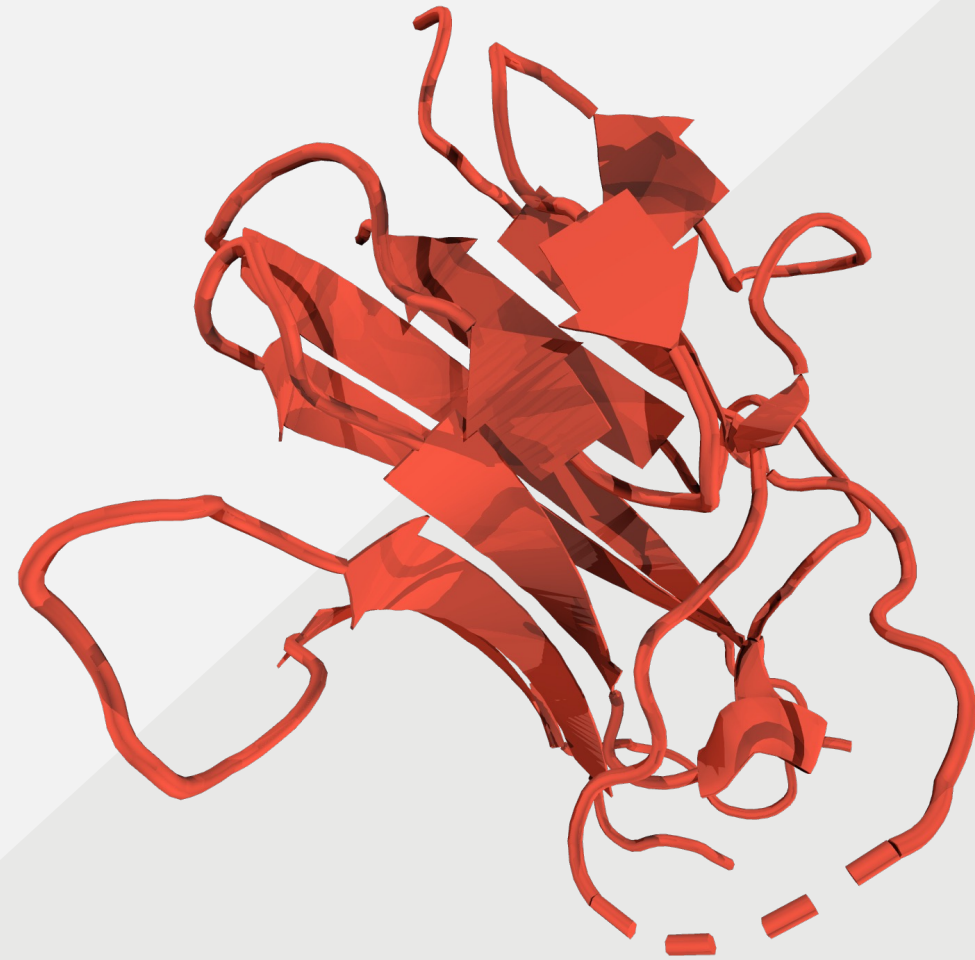


# Q&A



Mission:

**CREATE BETTER BIOLOGICS  
FOR PATIENTS, FASTER.**



# APPENDICES



## CASE STUDY - pH SENSITIVITY

# Models optimize pH sensitivity without introducing developability liabilities

mAb	K <sub>D</sub> at pH 5.8 (nM)	K <sub>D</sub> at pH 7.4 (nM)	Affinity fold (pH 7.4 over pH 5.8, Δ)	Stability T <sub>m</sub> (°C)	Aggregation by AC-SINS (shift nm)	Polyreactivity score, DNA ELISA (AU)	Polyreactivity score, insulin ELISA (AU)	*Sequence
Parental	24.1	4.4	5.5	67.0	-0.3	1.184	1.291	XXXXXXXXXXXXXXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXXXXXXXXXXXXXXXXXXXX
Variant 1	54.7	7500	-138	64.5	-0.3	1.515	1.747	KXXXXEXHXXXXX...XXXXXEXXXXXXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXXXXXXXXXXXXXXKXXX
Variant 2	9.6	180	-18	67.0	-0.7	1.328	1.369	XXXXXEXHXXXXX...XXXXXEXXXXXXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXHXXXXXXXXXXXXX
Variant 3	7.4	121	-16	65.8	-0.3	0.929	0.96	XXXXXEXHXXXXX...XXXXXEXXXXXXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXHXXXXXXXXXXXXX
Variant 4	8.9	127	-14	66.5	-0.3	1.692	1.89	KXXXXXXHXXXXX...XXXXXEXXXXXXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXXXXXXXXXXXXXXXXX
Variant 5	902	44.7	+20	68.5	-0.7	1.403	1.364	XXXEXXXXXXXXXX...XXXXXXXXXXDHXXXHX...XXXXXXXXXXRXXX...XXXXXXXXXXXXXXXXXXXXXXXXXX
Variant 6	609	36.0	+17	68.0	-1.3	1.382	1.605	XXXXXXXXXXXXXXXXXX...XXXXXXXXXXDHXXXHX...XXXXXXXXXXDRXXX...XXXXXXXXXXXXXXXXXXXXXXXXXX
Variant 7	272	16.7	+16	68.0	-0.7	1.6	1.719	XXXEXXHXXXXX...XXXXXXXXXXHXXXXX...XXXXXXXXXXXXXXXXXX...XXXXXXXXXXXXXXXXXXXXXXX
Variant 8	262	17.2	+15	67.5	1.0	1.018	1.031	XXXXXXRXXXXX...XXXXXXXXXXDHXXXHE...XXXXXXXXXXDRXXX...XXXXXXXXXXXXXXXXXXXXXXX
Low T <sub>m</sub> control	NA	NA	NA	55.8	-0.3	1.987	2.198	NA
Polyreactive control	NA	NA	NA	68.0	14.7	2.798	8.392	NA

\*X indicates positions of sampled amino acids from four regions on the heavy chain that included framework and CDRs. Substitutions of model-identified hits are shown in red.

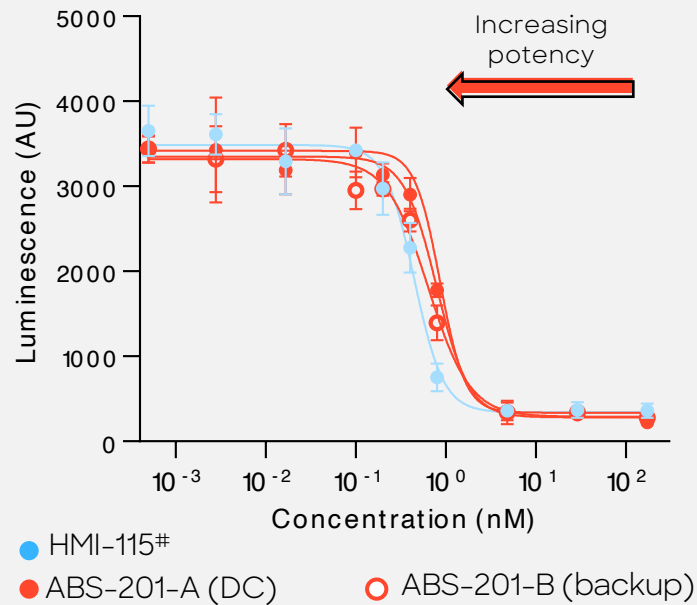
Distinct from classical histidine scanning, Absci's lab-in-the-loop workflow allows the model to search the entire relevant ionizable combinatorial space and identifies 10x-100x pH sensitive variants in both directions **without introducing developability liabilities.**



# ABS-201, PRLR

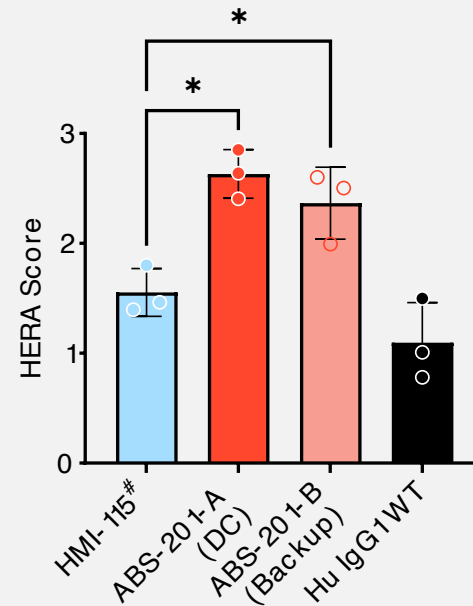
## Drug candidates are potent binders with extended half-life and favorable immunogenicity

### HIGH IN VITRO POTENCY



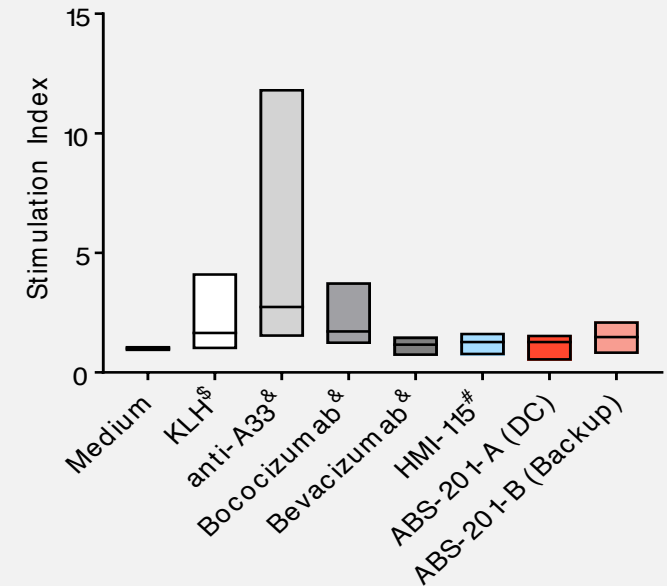
PathHunter assay

### EXTENDED HALF-LIFE



FcRn recycling *in vitro*

### FAVORABLE IMMUNOGENICITY



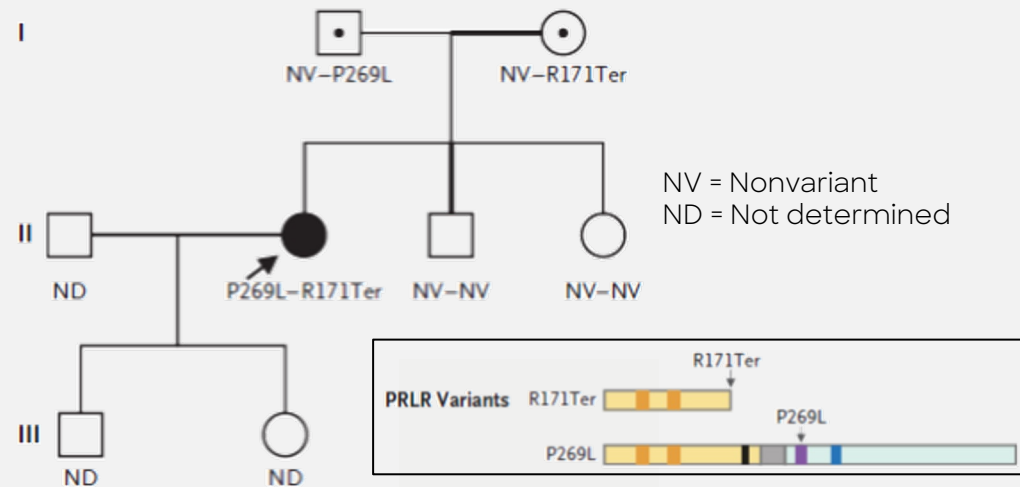
PBMC ex vivo immunogenicity assay

#Estimated performance of a putative clinical competitor

§Positive control; &High, mid and low standards as anti-A33, bococizumab and bevacizumab, respectively.

# Prolactin receptor (PRLR) inhibition anticipated to be safe and well tolerated

**Female compound heterozygous PRLR loss-of-function carrier lacks complete PRLR signaling**



A 35-year-old woman with postpartum agalactia and hyperprolactinemia, otherwise in **good health**, with no apparent impact on fertility and completely normal serum electrolytes and hormone levels (except prolactin, PRL).

**Serum electrolyte & hormone levels are within the normal range, except prolactin**

Description	Analyte	Values	Range	
Serum electrolytes	Na <sup>+</sup> - [mEq/l]	140	135-145	
	K <sup>+</sup> - [mEq/l]	5.1	3.6-5.0	
	Cl <sup>-</sup> - [mEq/l]	104	98-108	
	Ca <sup>2+</sup> - [mEq/l]	9.6	8.6-10.1	
	IP - [mEq/l]	4.6	2.5-4.5	
	DHEAS* [μg/dL]	215	23-266	
	IGF-1 [ng/ml]	119	111-279	
Serum hormones (Early prolif. Phase)	LH [IU/l]	4.5	1.68-15	
	FSH [IU/l]	8.8	1-9	
	TSH [mIU/l]	1.07	0.5-5.0	
	Estradiol [pg/ml]	57	20-350	
	Prolactin [ng/ml]	200	<25	
	Serum hormones (Mid luteal phase)	LH [IU/l]	3.7	0.61-16.3
		FSH [IU/l]	3.4	1-9
Estradiol [pg/ml]		82	30-450	
Progesterone [ng/ml]		14.5	2-25	
Testosterone [ng/ml]		0.24	0.15-0.46	
Prolactin [ng/ml]	188	<25		

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## ABS-301 | Broad potential in immuno-oncology

Based on literature and potential competitive molecules, the following indications could be of interest:

Indication	US Prevalence	Estimated 5-year survival rate*	US Sales in 2030 <sup>12</sup>
NSCLC	Calculated: ~202K in 2023 <sup>1</sup>	28% <sup>2</sup>	\$27B
SCC	30% of NSCLC cases Calculated: ~61K	24% <sup>3</sup>	Calculated Sales: \$8.1B
Head and Neck SCC	~54K in 2022 <sup>4</sup>	68.5% <sup>4</sup>	Calculated Sales: \$2.3B
Esophageal Cancer	~21K in 2022 <sup>5-7</sup>	20% <sup>7</sup>	\$1.5B
SCC	~20% of cases Calculated: ~4.2K		Calculated Sales: \$0.3B
Cervical Cancer	~14K in 2023 <sup>8</sup>		\$0.6B
SCC	90% of cases Calculated: ~13K	67% <sup>9</sup>	Calculated Sales: \$0.6B
Skin Cancer, non-melanoma	Incidence = ~3,300K <sup>10</sup>	95-100% <sup>11</sup>	\$1.0B
SSC	Incidence = ~700K <sup>10</sup>	95% <sup>11</sup>	Calculated Sales: \$0.2B

\*dependent on stage of diagnosis

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