

1. Proteins as drugs

2. Designing protein drugs *de novo*

Tanja Kortemme

# Introduction

---

- About me:  
Background in Physical Chemistry & Biophysics  
Ph.D. in Biophysics from EMBL, Heidelberg, Germany
  
- Research in quantitative biology (computational & experimental)  
<http://kortemmelab.ucsf.edu>
  
- Development of computational protein design methods since 1999  
<https://www.rosettacommons.org>
  
- Earlier review on our work: Mandell et al, “Computer-aided design of functional protein interactions” *Nature Chemical Biology* 2009
- Recent highlight: Glasgow et al, “Computational design of a modular protein sense-response system” *Science* 2019

# Agenda

---

## Background

- Why protein drugs

## State of the field

- Principles of engineering & optimizing antibody drugs

## Current research

- Creating protein drugs from scratch
- Computational protein design (Rosetta)
- Successes and challenges

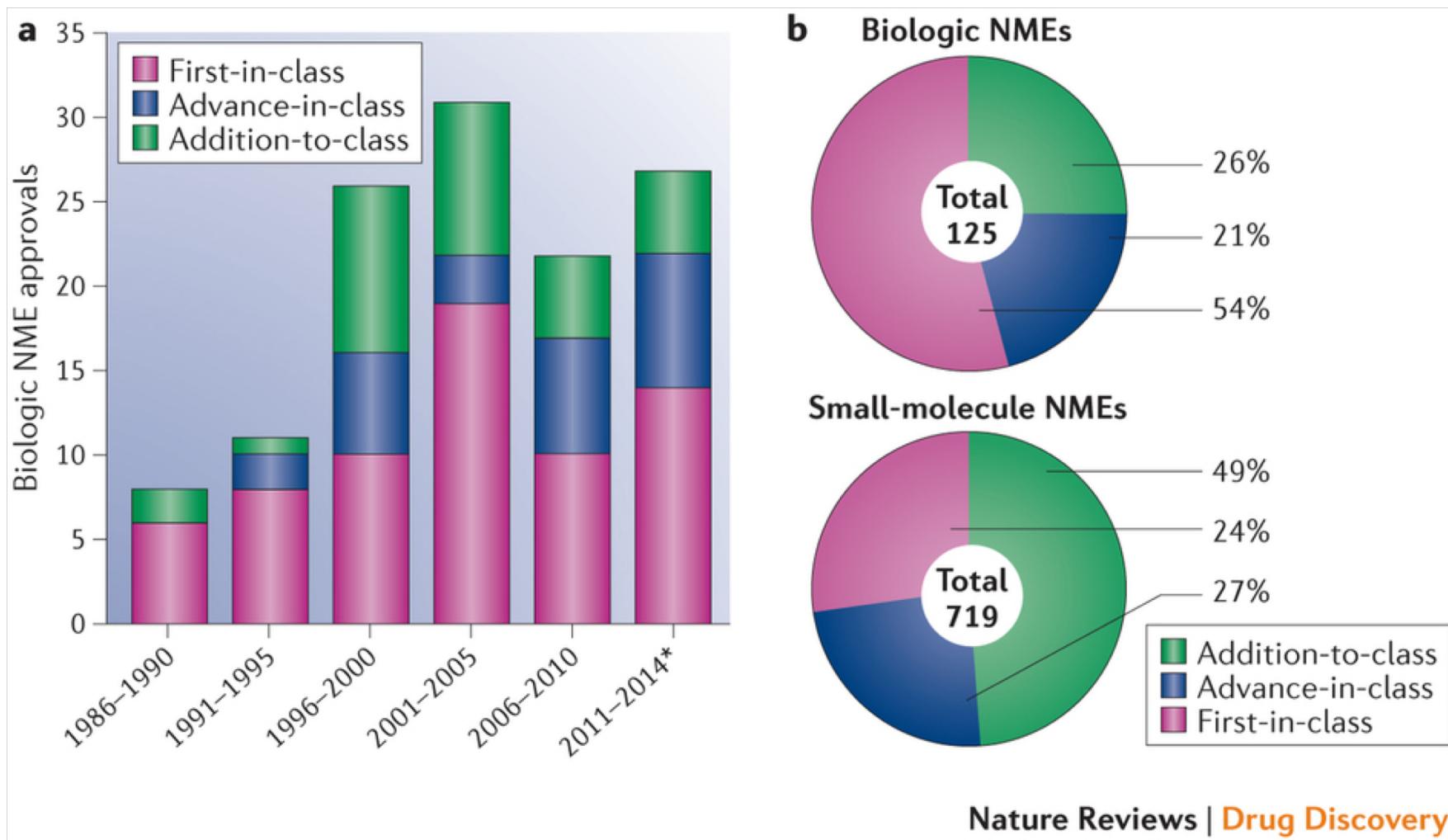
# Proteins are important therapeutics

---

- > 200 protein therapeutics on the market

Name	Target disease	Market (2011)
Adalimumab/Humira Pen Etanercept/Enbrel	Rheumatoid arthritis	\$8 billion
Infliximab/Remicade	arthritis	
Rituximab/Rituxan	Non-Hodgkin's B-cell lymphoma	\$7 billion
Bevacizumab/Avastin Trastuzumab/Herceptin	Colorectal cancer Breast cancer	\$6-5 billion
Insulin glargine/Lantus	Type I and II diabetes	\$4.8 billion
Epogen (erythropoietin)	Renal anemia	\$2.5 billion

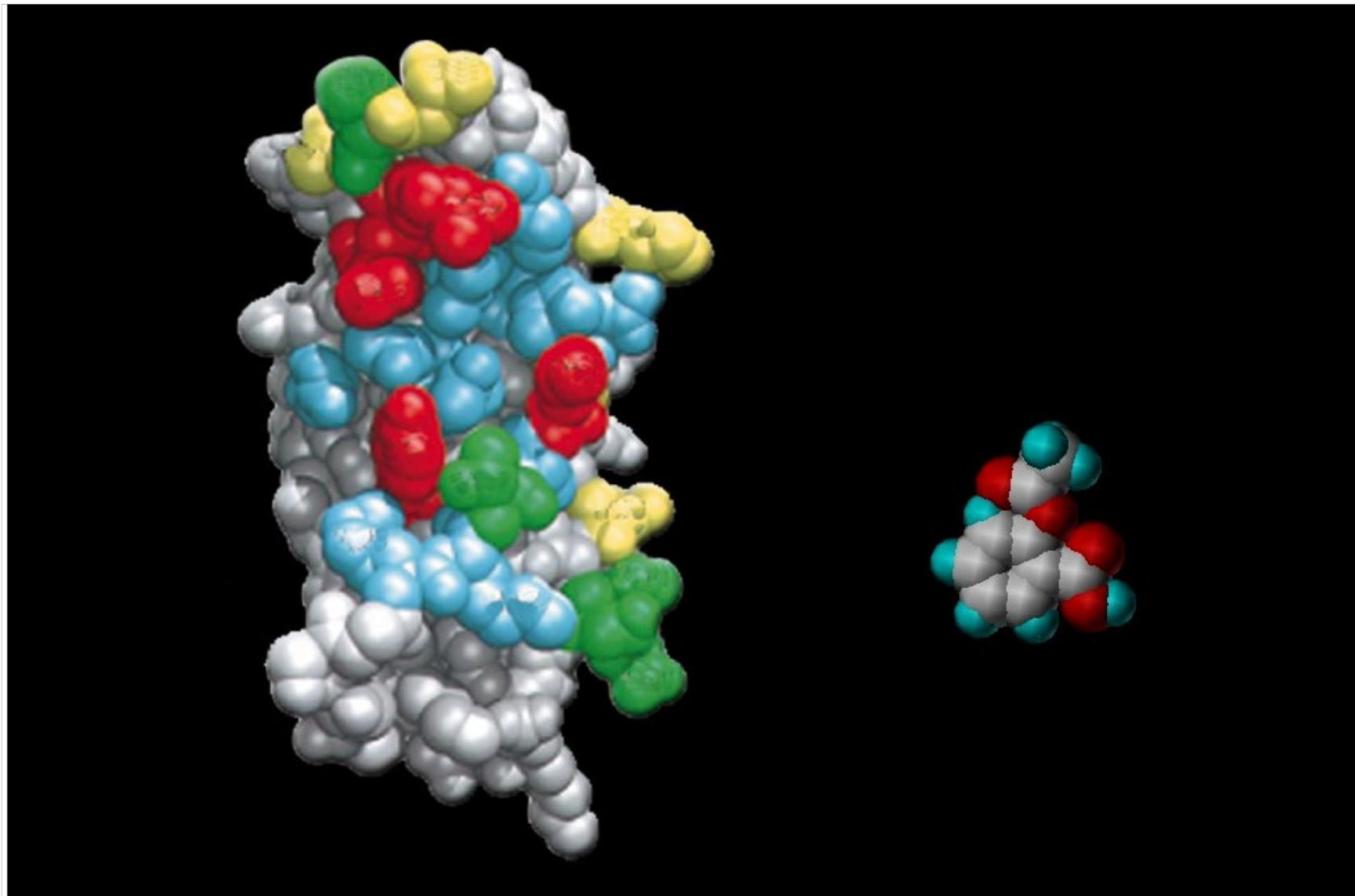
# Biologics are innovative



**a** | Biologic NME approvals by the US Food and Drug Administration in 1986–2014, split into innovation categories and 5-year time periods. **b** | Comparison of the innovativeness of biologic and small-molecule NMEs approved in 1986–2014. NME, new molecular entity. \*The last bar is only a 4-year time period.

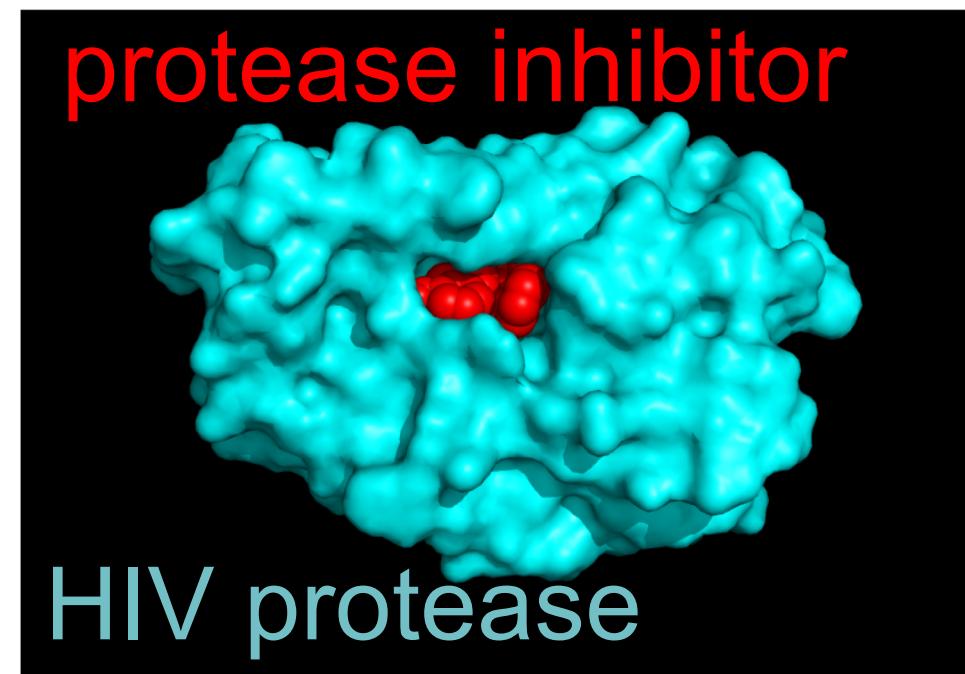
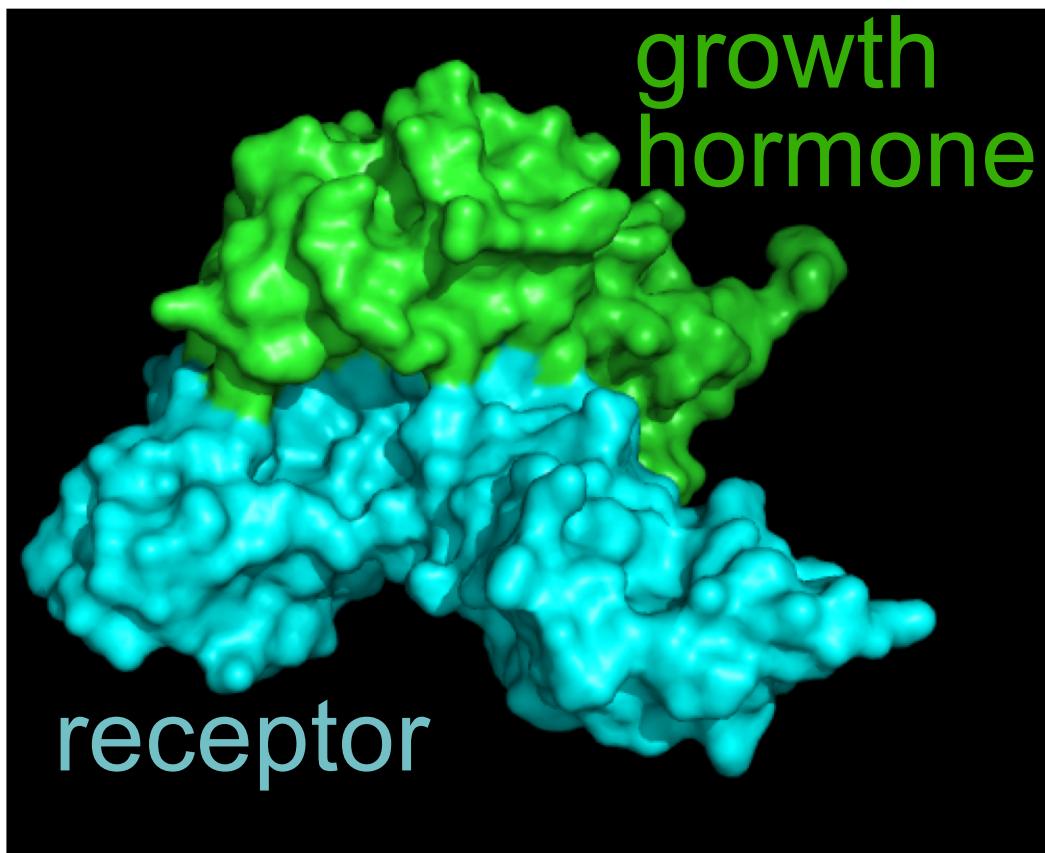
# Protein and small molecule drugs have different targets

---



# Protein and small molecule drugs have different targets

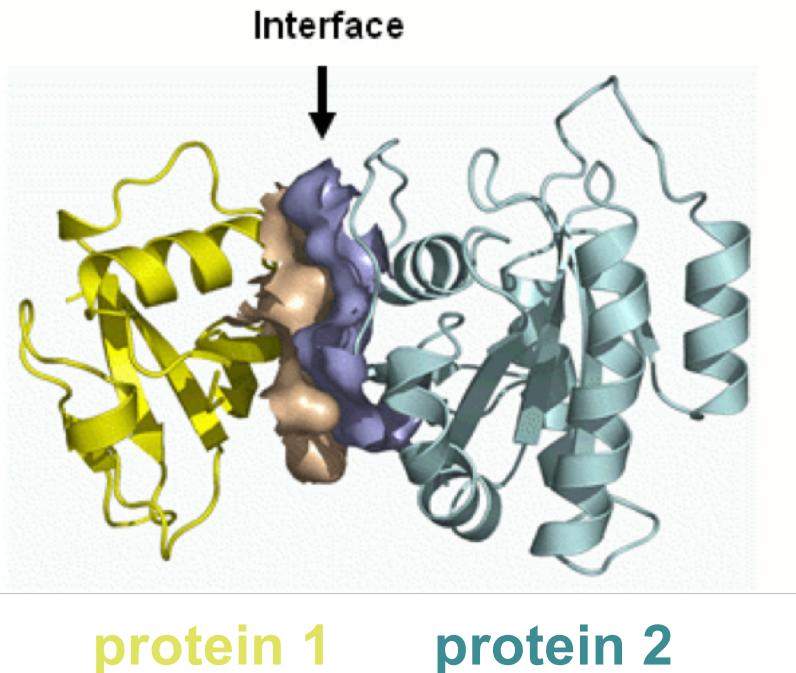
---



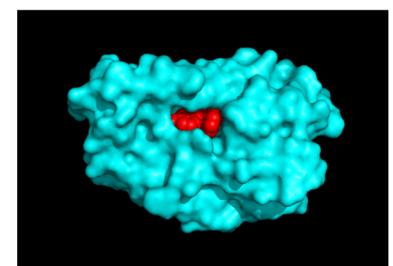
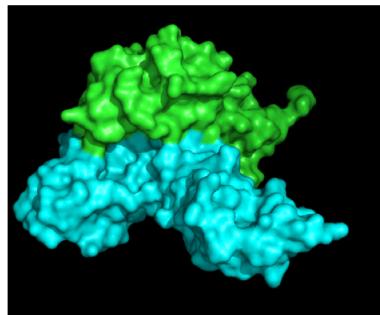
- flat interface
- many interactions
- binding pocket (“druggable”)
- fewer interactions

# Proteins can be very potent in blocking protein-protein interactions

---

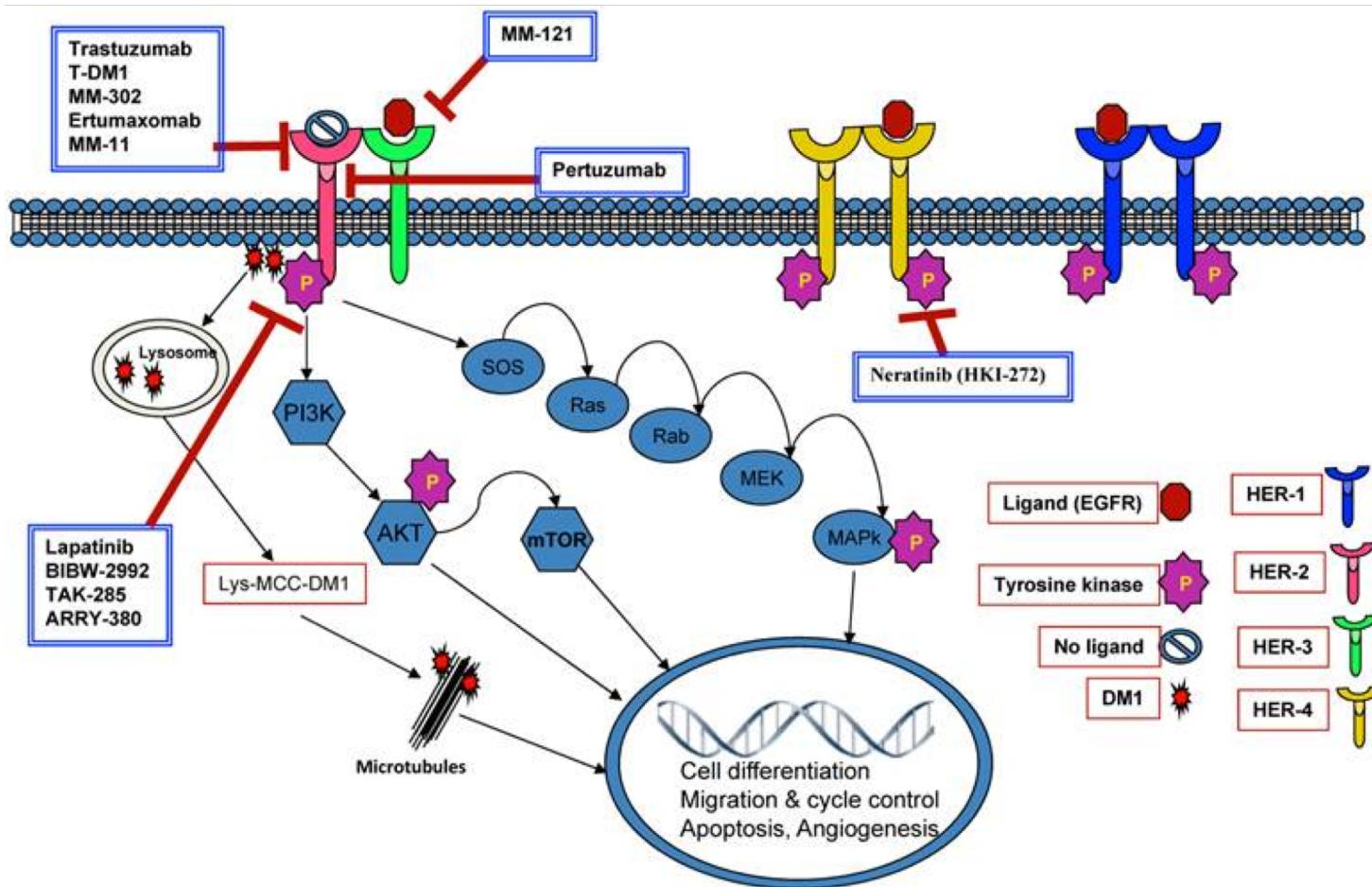


- a small molecule drug may not have enough surface/binding energy to efficiently block a protein-protein interaction \*

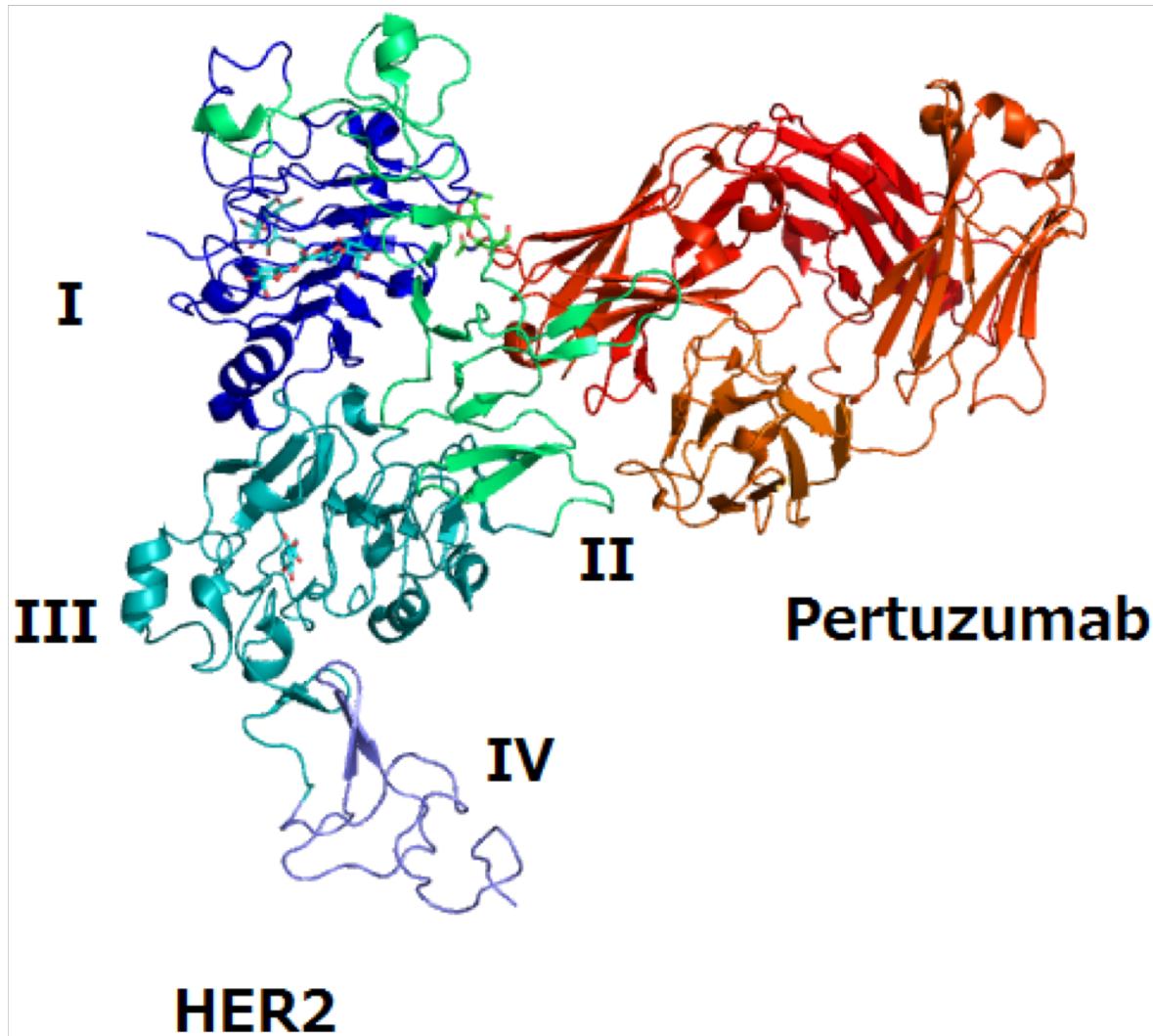


\* large protein-protein interfaces are difficult to disrupt with a small molecule, although sometimes possible: see Jim Wells & Chris McClendon, *Nature* 2007

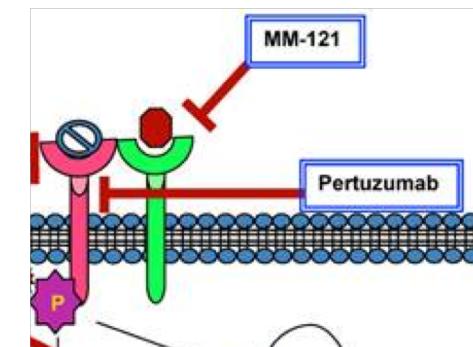
# Protein-protein interactions important in cancer are key targets for inhibition



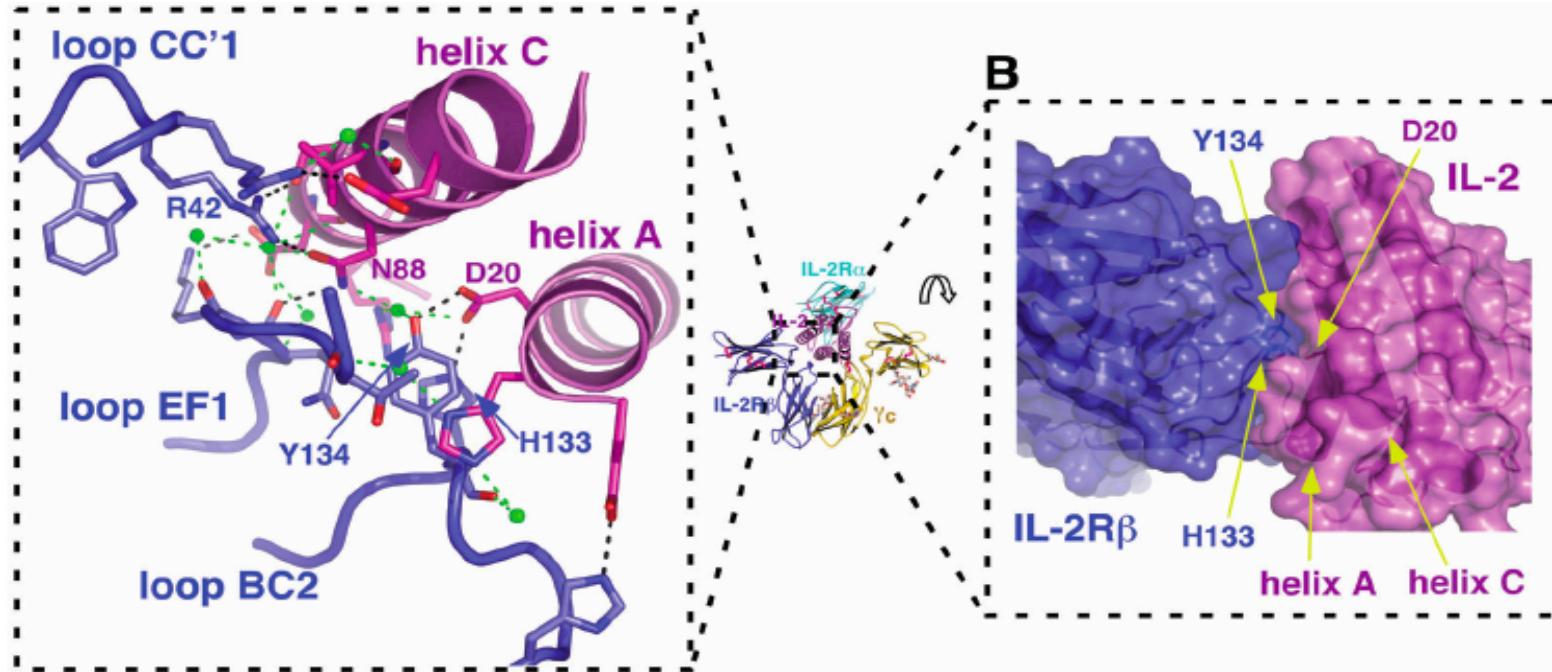
# Proteins drugs can be very potent in blocking protein-protein interactions



- Pertuzumab inhibits dimerization of Her2 with other Her receptors



# Proteins can be very specific, because they can form many defined interactions with their targets



Example **IL-2/ IL-2 receptor**, Wang et al., Science 310: 1159-1163, 2005

A protein drug may be able to form similar specific interactions to distinguish a target, such as the IL-2 receptor, from a paralog that has slightly different amino acids in the interface.

# Agenda

---

## Background

- Why protein drugs:
  - **potent** as protein-protein interaction inhibitors
  - can perform “**endogenous**” **activity**: insulin, human growth hormone, IL-2
  - often very **specific**

# Agenda

---

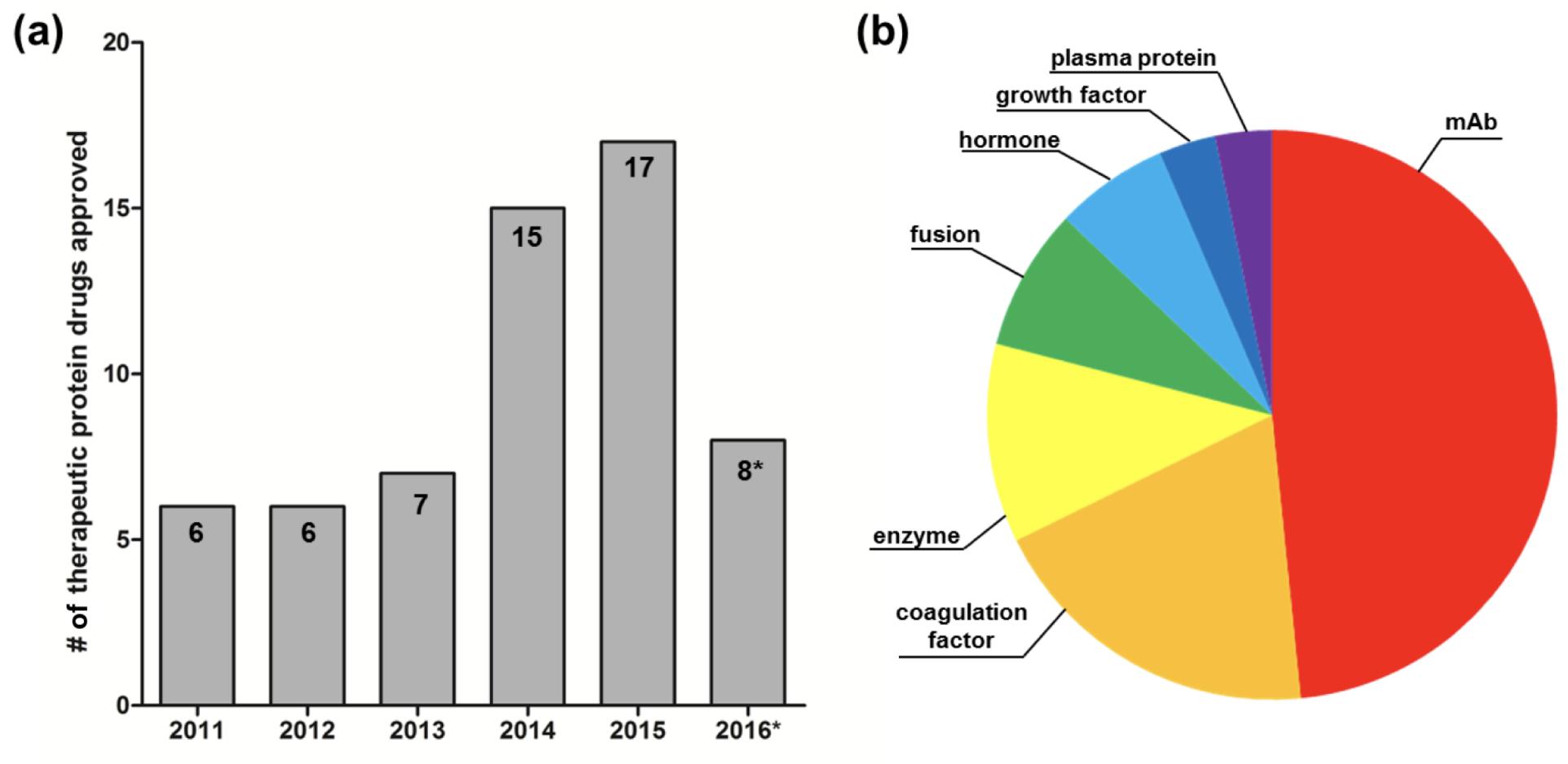
## Background

- Why protein drugs

## State of the field

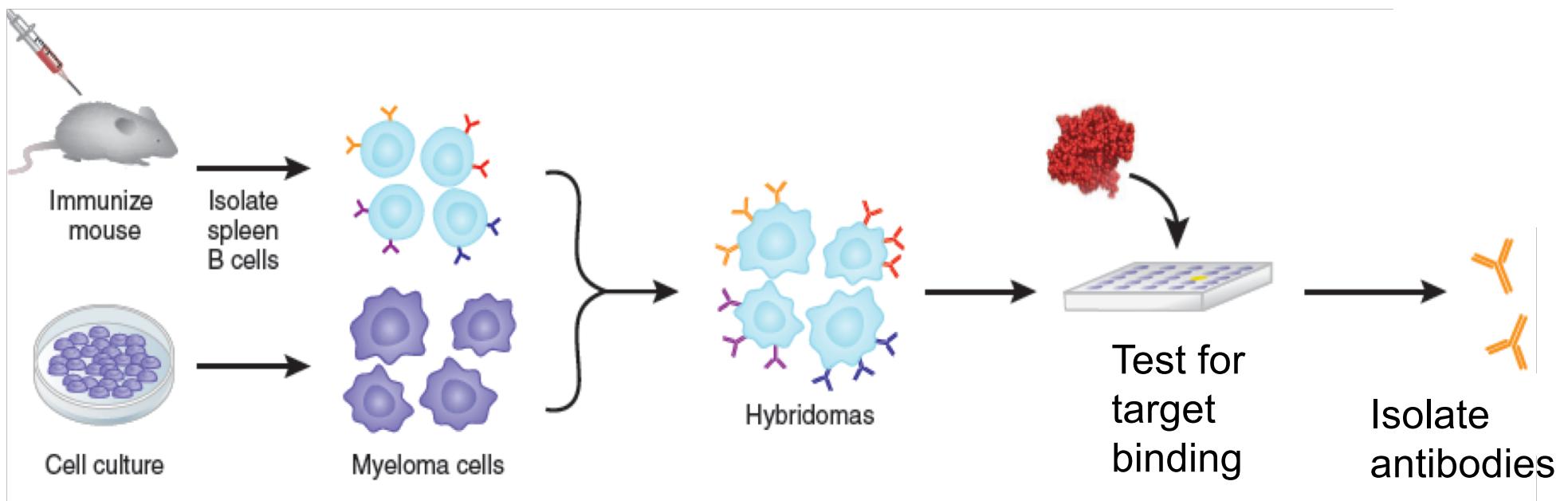
- Principles of engineering & optimizing antibody drugs

# Protein drug FDA approval 2011-2016



**Figure 2. U.S. Food and Drug Administration (FDA)-approved therapeutic proteins (2011–2016\*).** (a) Bar graph showing the number of therapeutic protein FDA approvals by year (2011–2016\*). (b) Pie chart showing the distribution of FDA-approved therapeutic proteins (2011–2016\*) by drug class. \*January 1, 2011, through August 31, 2016.

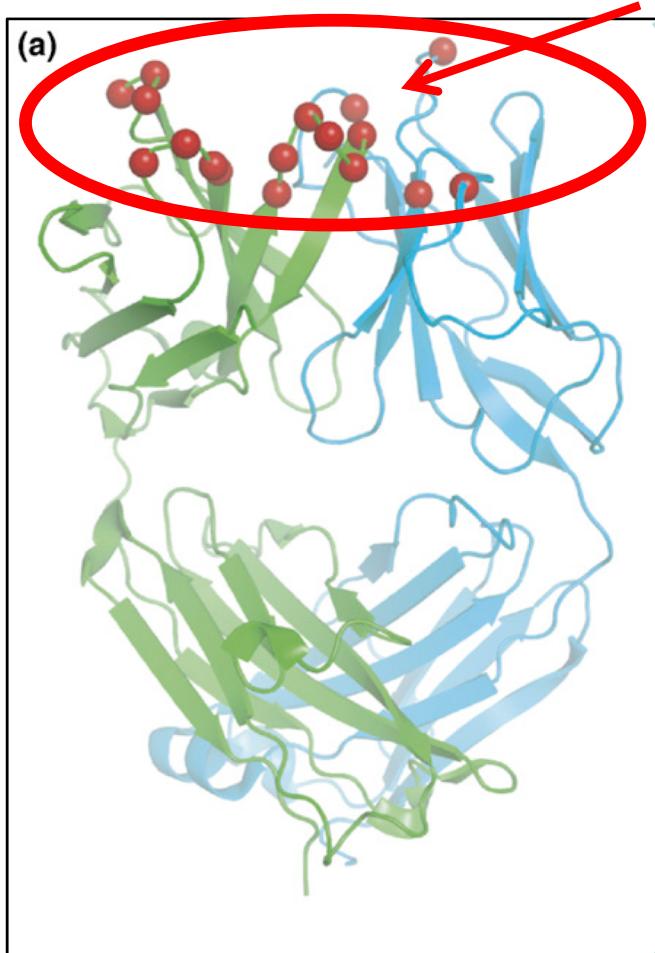
# Monoclonal antibodies can be generated *via* immunizing mice



But the problem with this approach: antibodies are from mouse and can lead to immune reactions in humans!

*(but even completely human antibodies can cause immune reactions)*

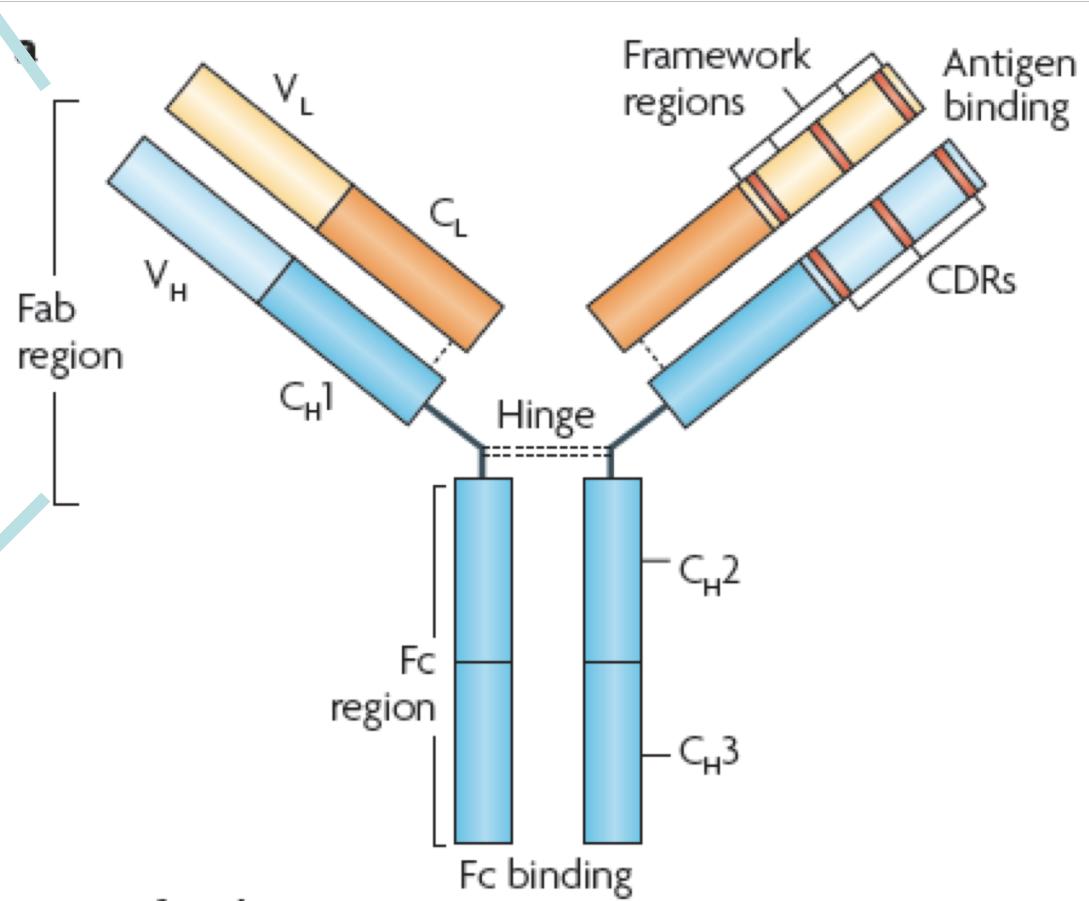
# Antibody Structure



**Antigen-binding Fragment (Fab)**

CDRs (complementarity-determining regions) that bind the antigen are formed by a small region: loops in the heavy (H1-H3) and light chains (L1-L3)

Red spheres indicate amino acid residues mainly responsible for target recognition



# Different classes of therapeutic antibodies have a different fraction of mouse amino acid sequence

These are most desired

**Approved therapeutic antibodies**

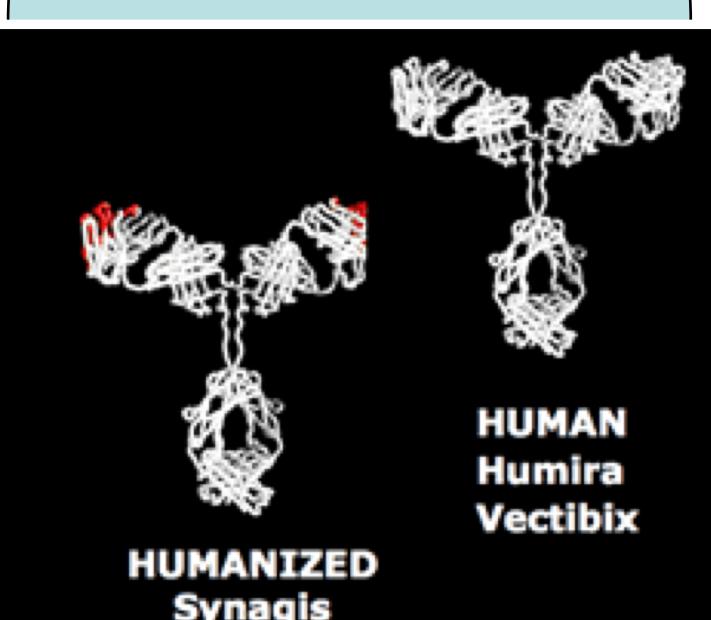
**Mouse**  
**human**

**MOUSE**  
**OKT3**  
**BEXXAR**  
**Zevalin**

**CHIMERIC**  
**Rituxan**  
**Remicade**  
**Reopro**  
**Simullect**  
**Erbilux**

**HUMANIZED**  
**Synagis**  
**Herceptin**  
**Zenapax**  
**Myelotarg**  
**Campath**  
**Xolair**  
**Raptiva**  
**Avastin**  
**Tsyabri**  
**(Actemra-Japan)**

**HUMAN**  
**Humira**  
**Vectibix**



The diagram illustrates the structure of an antibody molecule. It is divided into two main regions: the Fab (Fragment antigen-binding) region at the top and the Fc (Fragment crystallizable) region at the bottom. The Fab region contains the variable light chain (V<sub>L</sub>) and the constant light chain (C<sub>L</sub>). The Fab region is further divided into Framework regions and CDRs (Complementarity Determining Regions). The Fc region contains the variable heavy chain (V<sub>H</sub>) and the constant heavy chains C<sub>H</sub>1, C<sub>H</sub>2, and C<sub>H</sub>3. The Fc region is also labeled with Fc binding.

**a**

Fab region

V<sub>L</sub>

C<sub>L</sub>

V<sub>H</sub>

C<sub>H</sub>1

Hinge

C<sub>H</sub>2

C<sub>H</sub>3

Framework regions

CDRs

Antigen binding

Fc region

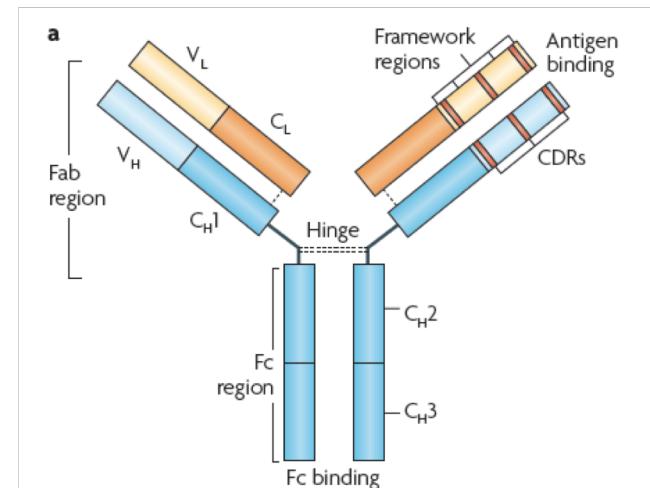
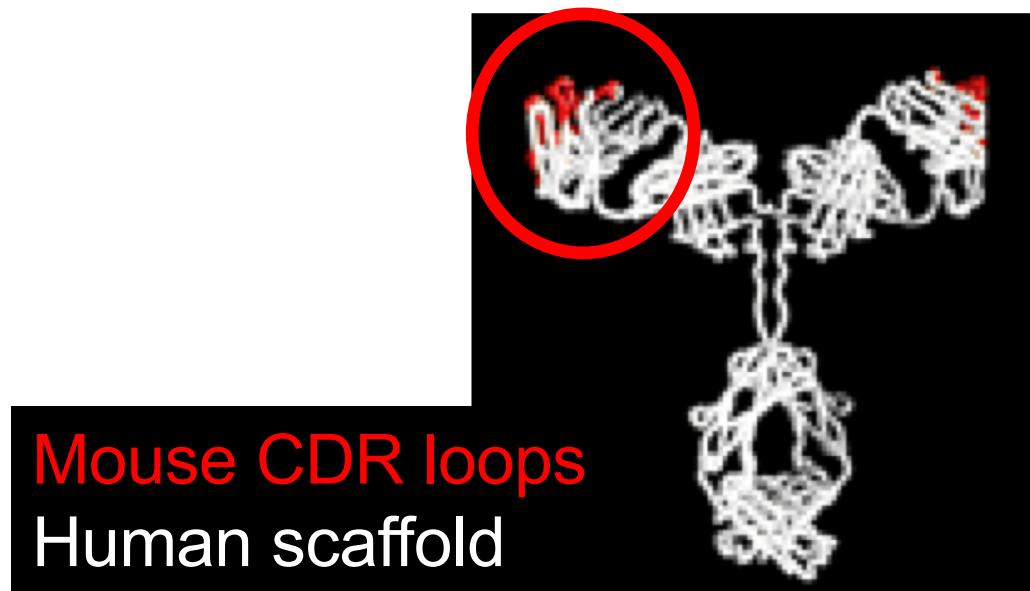
Fc binding

## Types of mAbs

Murine	Entirely murine amino acids
Chimeric	Human constant (C) + murine variable (V) regions
Humanized	Murine complementarity determining regions (CDRs)
Human	Entirely human amino acids

# “Humanized” antibodies are a common solution to decease the immunogenicity of mouse antibodies

- “transplant” the CDR loops from the mouse antibody onto a human constant “scaffold”
- How? Insert into the gene sequence



The problem with this approach: “transplanting” generally decreases potency (small sequence differences in the scaffold can change the precise structure / positioning of the CDR loops )

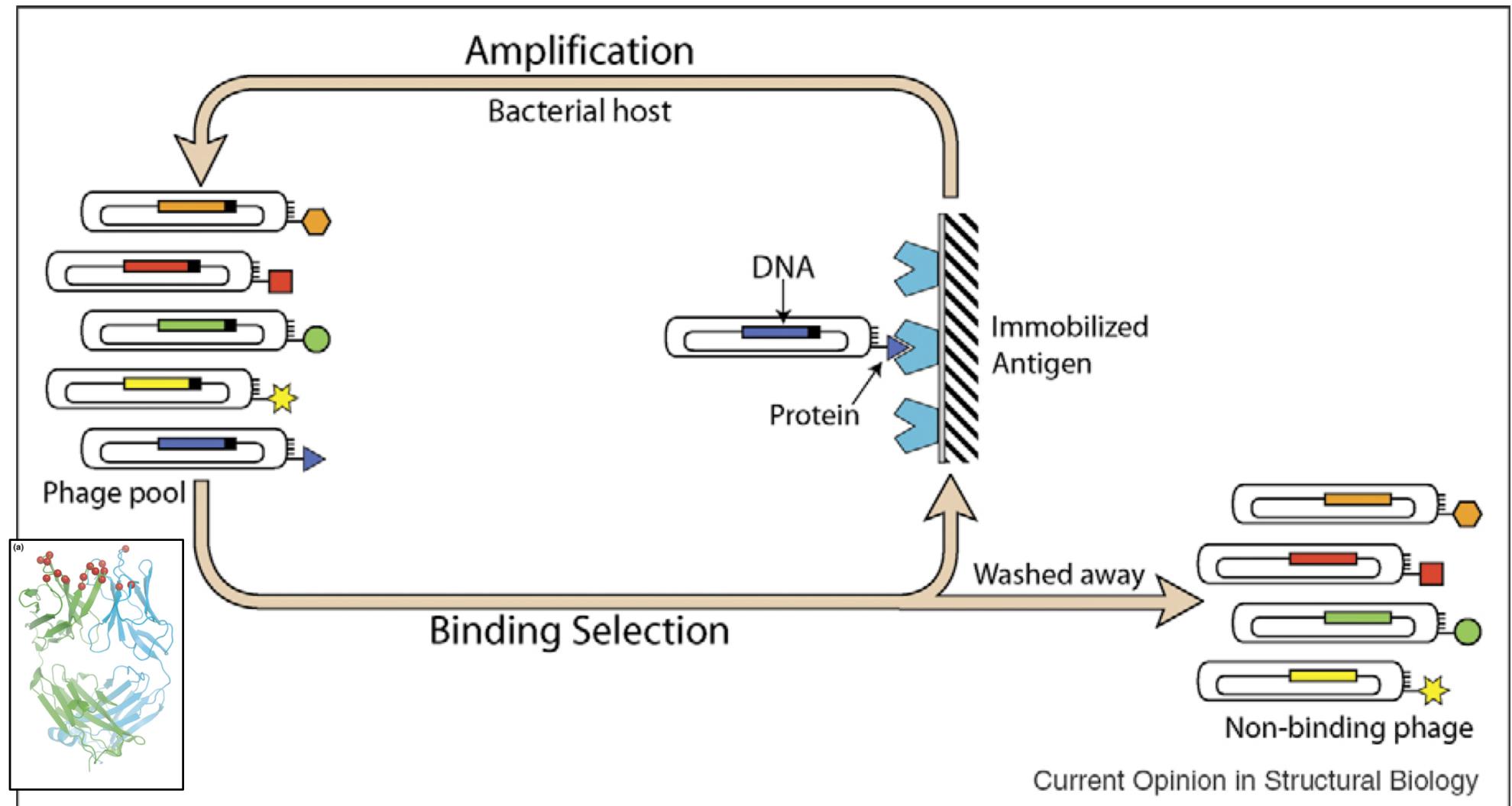
# Can we use a human antibody scaffold in the first place?

---

- Can be done using a technique called “**phage display**”: select from “libraries” of antibody variants where the scaffold is human and the red residues are varied
- Phage display can also be used to **improve potency**: antibody “affinity maturation” by selection in the laboratory



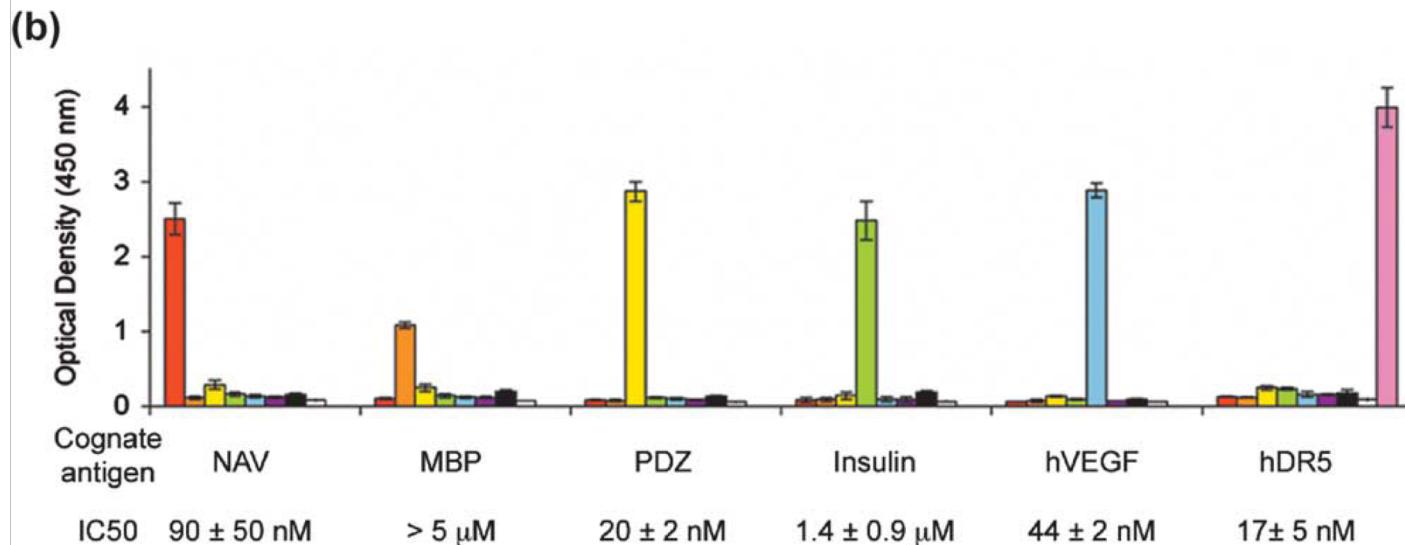
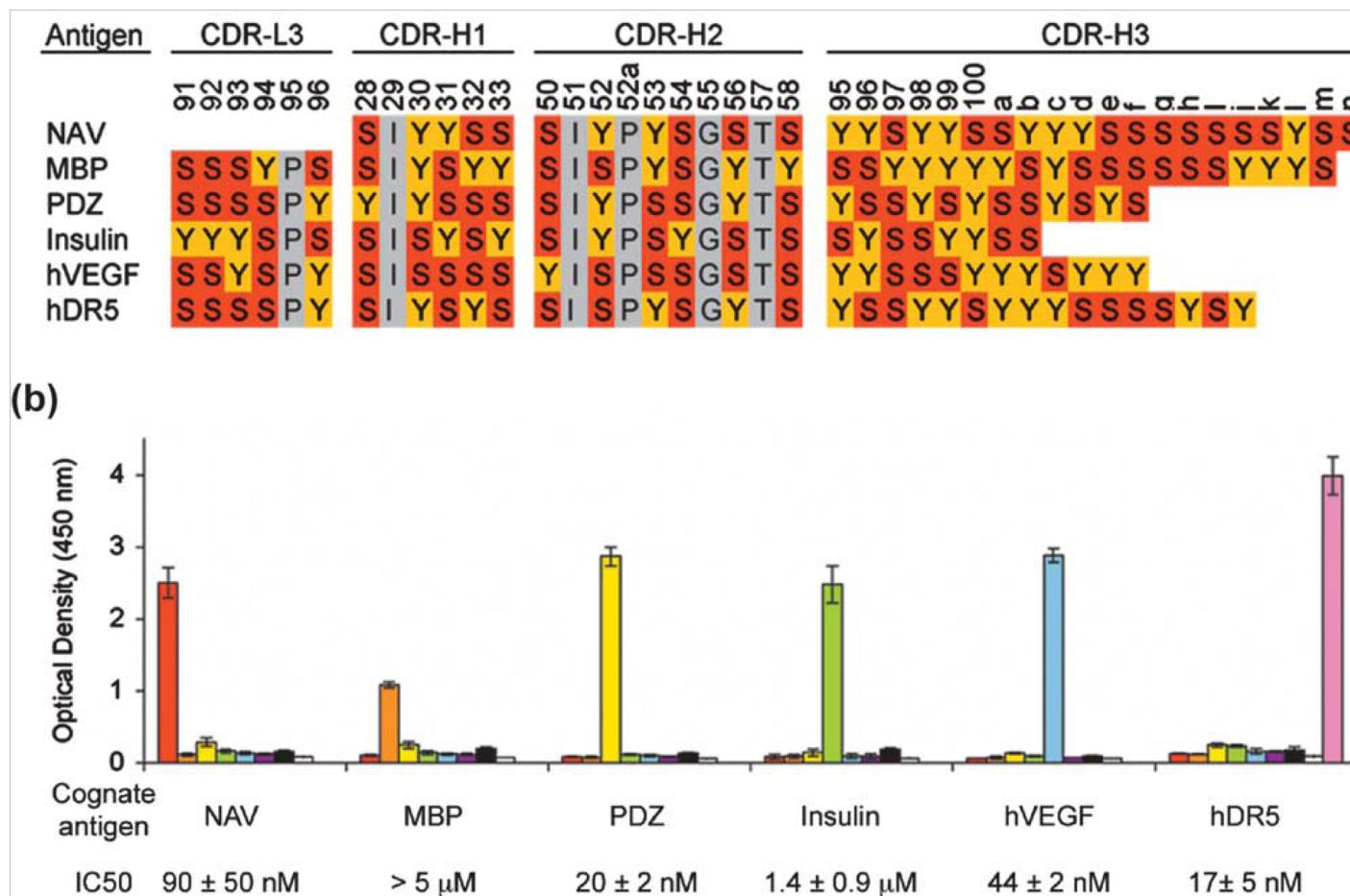
# “Phage display” can efficiently generate specific antibodies (and also optimize them)



Sidhu & Koide, 2007

# By phage display, antibodies can be selected to be specific for diverse antigens

The same “library” of antibody variants (that had the sequences of their CDR loops randomized) was screened against 6 targets to yield 6 different specific antibodies



# Agenda

---

## Background

- Why protein drugs

## State of the field

- Principles of engineering & optimizing antibody drugs

## Current research

- Creating protein drugs from scratch
- Computational protein design (Rosetta)
- Successes and challenges

# Can we design (non-antibody) protein therapeutics *de novo*?

---

- let's first think about why we would want to...

# Can we design (non-antibody) protein therapeutics *de novo*?

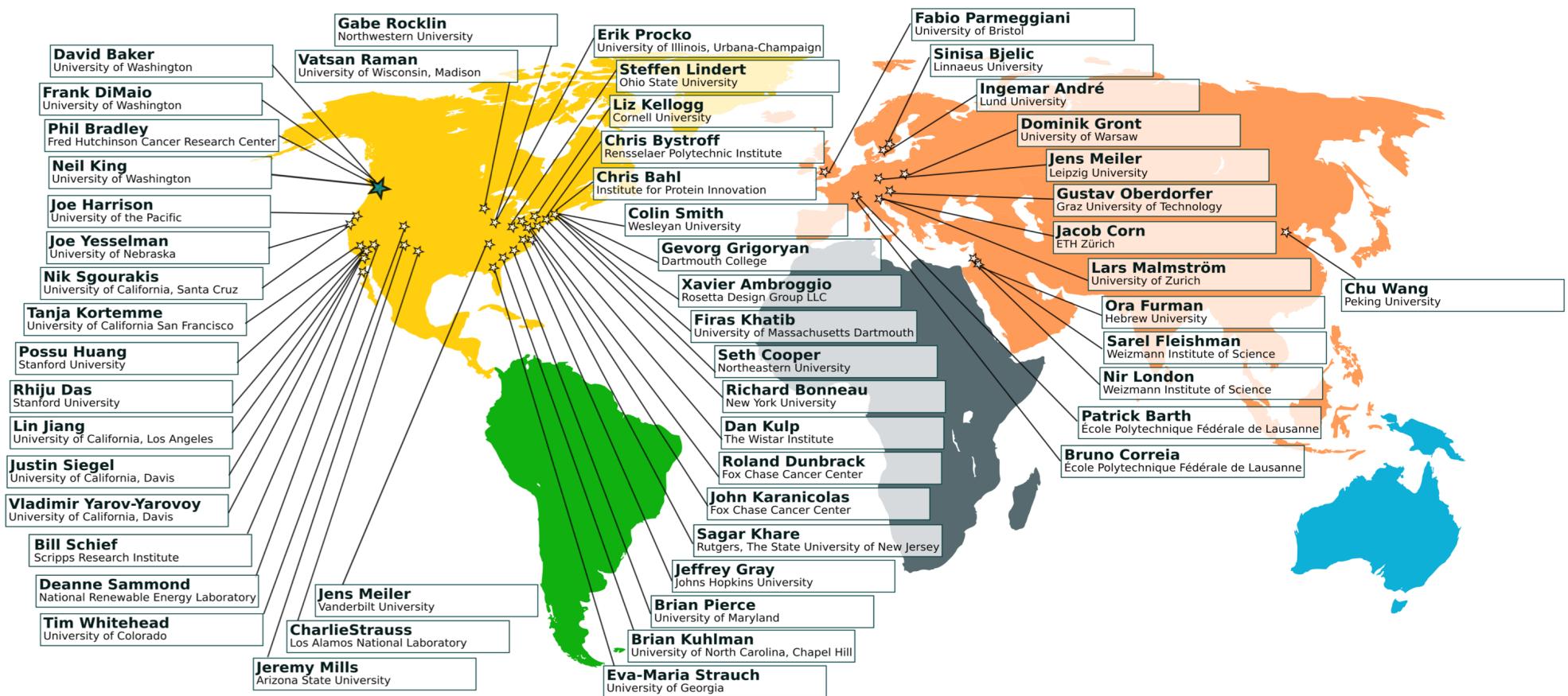
---

- Two case-studies:
  - Vaccines
  - Potent and selective mimics of endogenous proteins

Before we get to applications: How does *de novo* protein design work? -> in Rosetta

# A large community of research labs develop Rosetta to model & design proteins

Source code free for academia: [rosettacommons.org](http://rosettacommons.org)  
Licensed for a fee by >70 companies

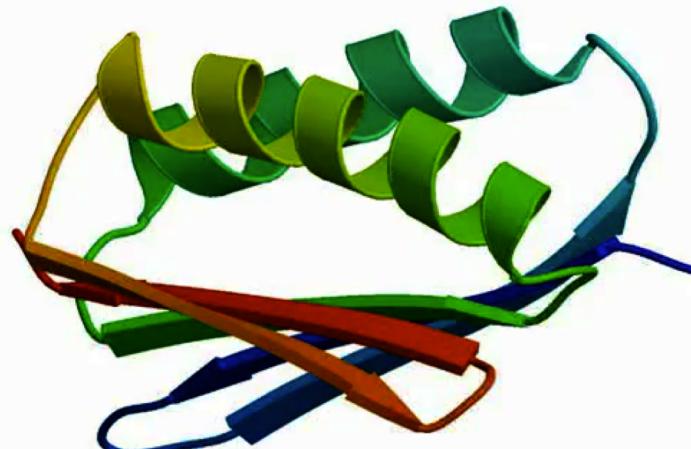


# Computational protein design (Rosetta) is an optimization problem

---

## INPUT

Design Objective:  
Structure &  
Function  
(represented at  
all-atom level)



## OUTPUT

Amino acid  
Sequences  
Optimized for  
Design Objective &  
Structure

Score each structure-sequence  
combination, find “best”

Monte-Carlo simulated annealing  
Genetic Algorithms  
SAT solvers  
Dead-end Elimination

...

The key challenge is that the possible space is  
absolutely enormous

---

possible sequences for 100 residue protein:  $20^{100} \sim 10^{130}$   
(most proteins are larger; only a small fraction will be functional)

number of atoms in the universe:  $\sim 10^{80}$

## The key challenge is that the possible space is absolutely enormous

---

possible sequences for 100 residue protein:  $20^{100} \sim 10^{130}$   
(most proteins are larger; only a small fraction will be functional)

number of atoms in the universe:  $\sim 10^{80}$

number of different proteins on earth today:  $\sim 10^{12}$

number of proteins sampled in evolution:  $10^{21} - 10^{50}$

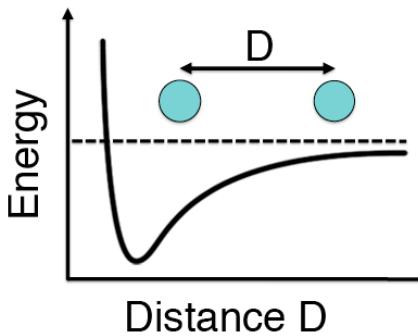
## As a consequence, need to make simplifications

---

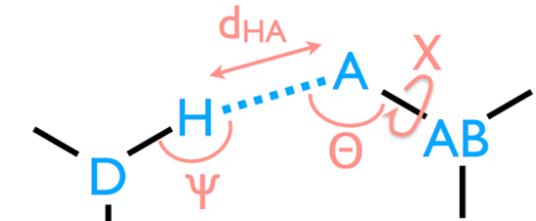
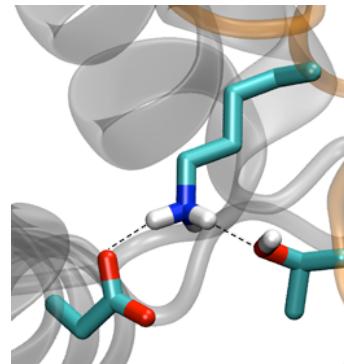
- How we "sample" space (reduced degrees of freedom)
- How we "score" solutions (approximate energy function)

# Rosetta all-atom energy function

## 1. van der Waals packing

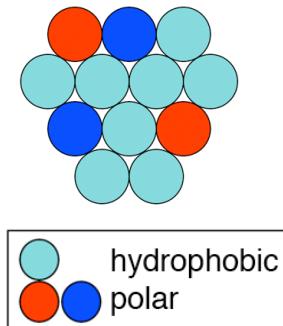


## 2. hydrogen bonding



Kortemme & Baker, PNAS 2002  
Kortemme et al., JMB 2003

## 3. implicit solvation

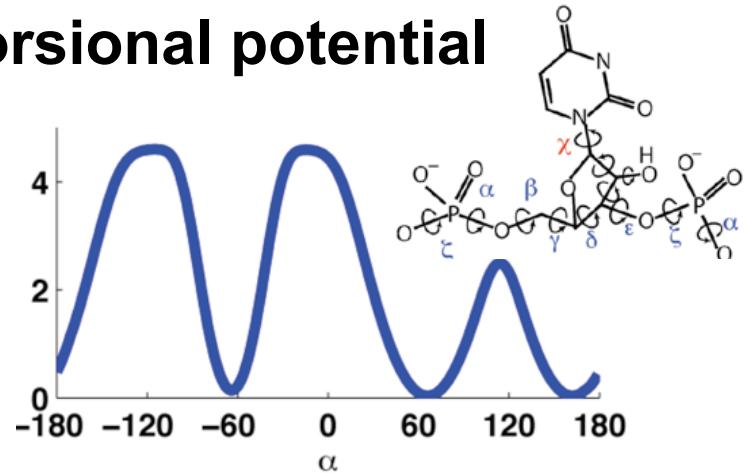


$$\Delta G_i^{slv} = \Delta G_i^{ref} - \sum_{j \neq i} f_i(r_{ij}) V_j$$

$\Delta G^{ref}$ : reference solvation free energy  
 $V_j$ : Volume of group j  
 $f_i$ : solvation free energy density of i  
 $r_{ij}$ : distance between i and j

Lazaridis & Karplus, Proteins 1999

## 4. torsional potential



## 5. electrostatic repulsion (screened)

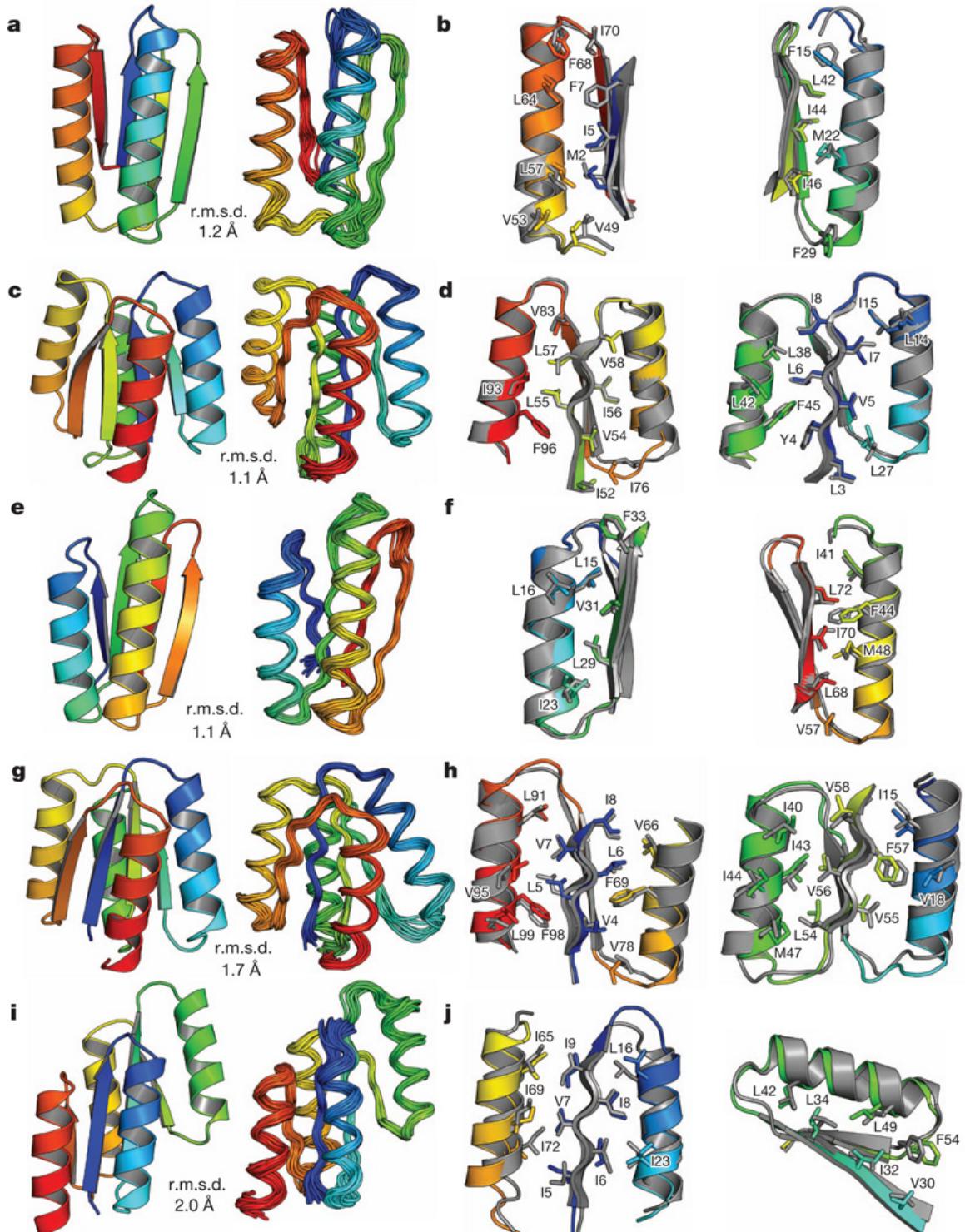
free energy - configurational entropy

## As a consequence, need to make simplifications

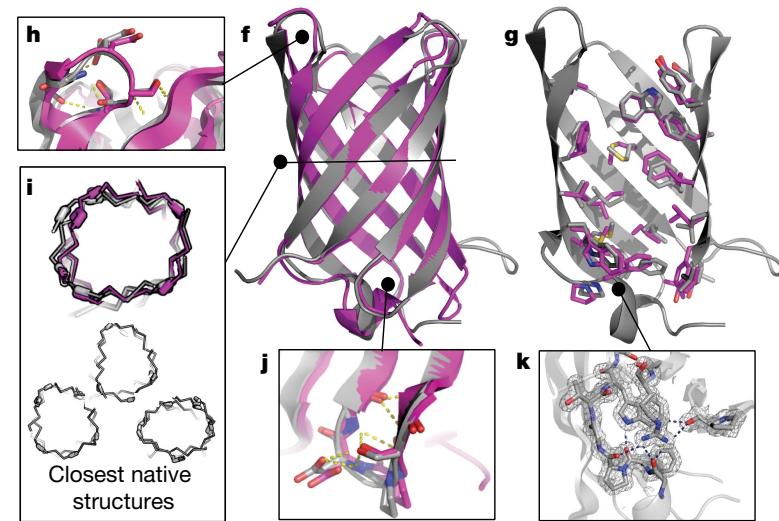
---

- How we "sample" space (reduced degrees of freedom)
- How we "score" solutions (approximate energy function)

Both of these are main causes of errors and determine current state of the field.



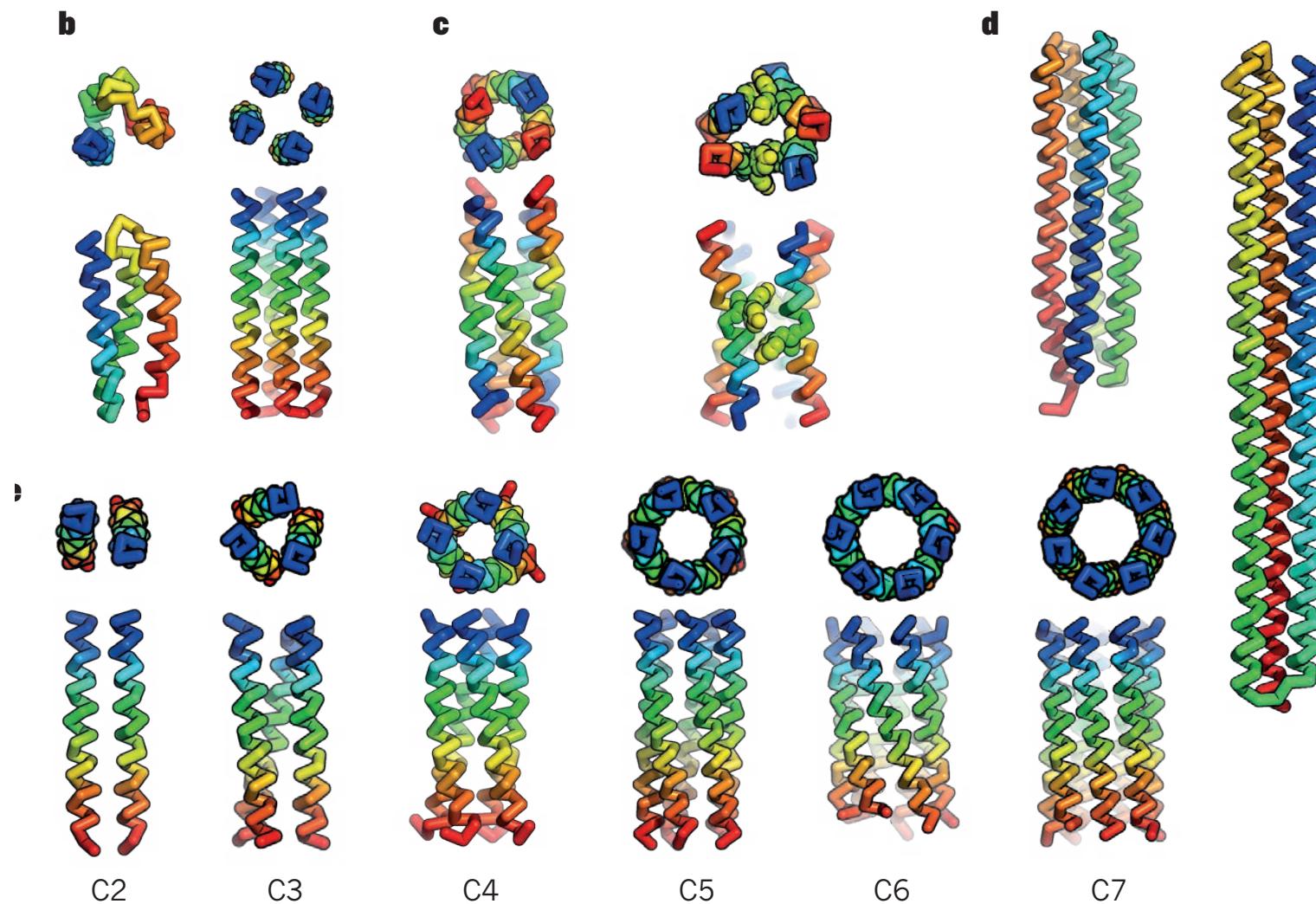
Despite these simplifications, many design successes: small, "idealized" folds



Koga et al, Nature 2012 DOI:  
[10.1038/nature11600](https://doi.org/10.1038/nature11600)  
Dou, Vorobieva et al, Nature 2018  
DOI: [10.1038/s41586-018-0509-0](https://doi.org/10.1038/s41586-018-0509-0)

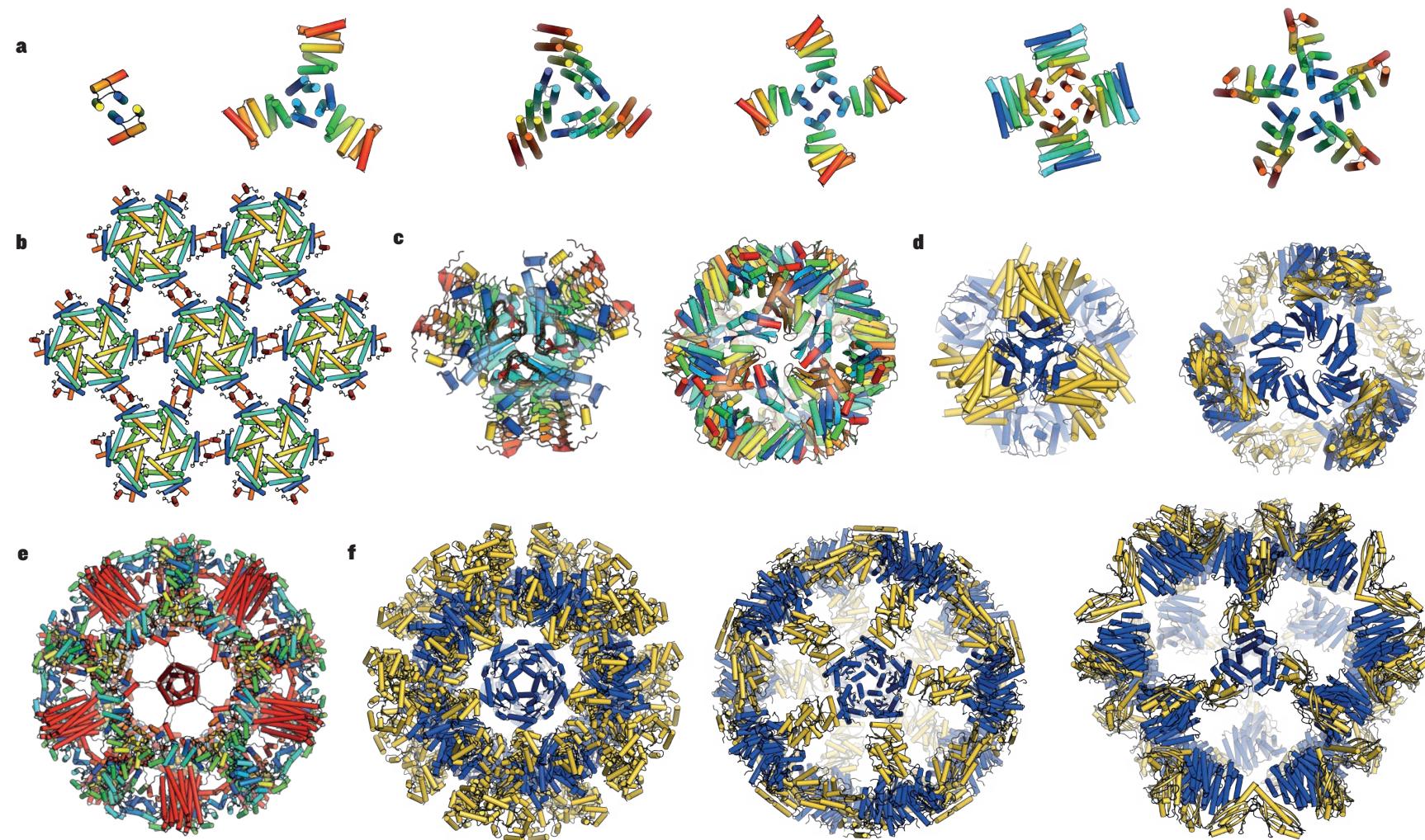
In particular, a wealth of architectures from helical bundles (that can be “functionalized”, more later)

---



# Helical structures can be assembled into a range of higher-order architectures

---



# Computational protein design: state of the field

---

(>3 decades of fundamental work, enormous progress in applications in the last 15 years!)

- *de novo* folds built from rules: a, a/b and all-beta proteins
- new architectures, symmetrical assemblies & materials
- helical bundles, can be functionalized

# Current challenges are in designing function

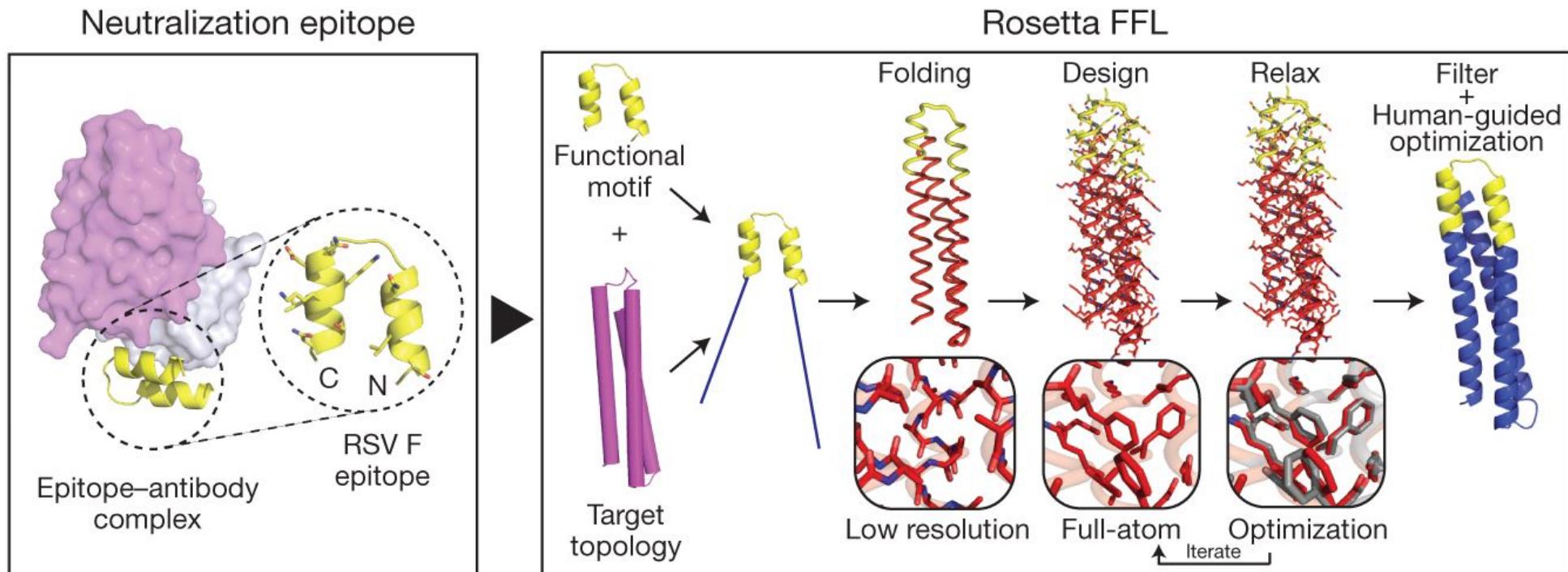
---

- sensors
- switches
- efficient enzymes
- machines
- ... many complex and composite functions

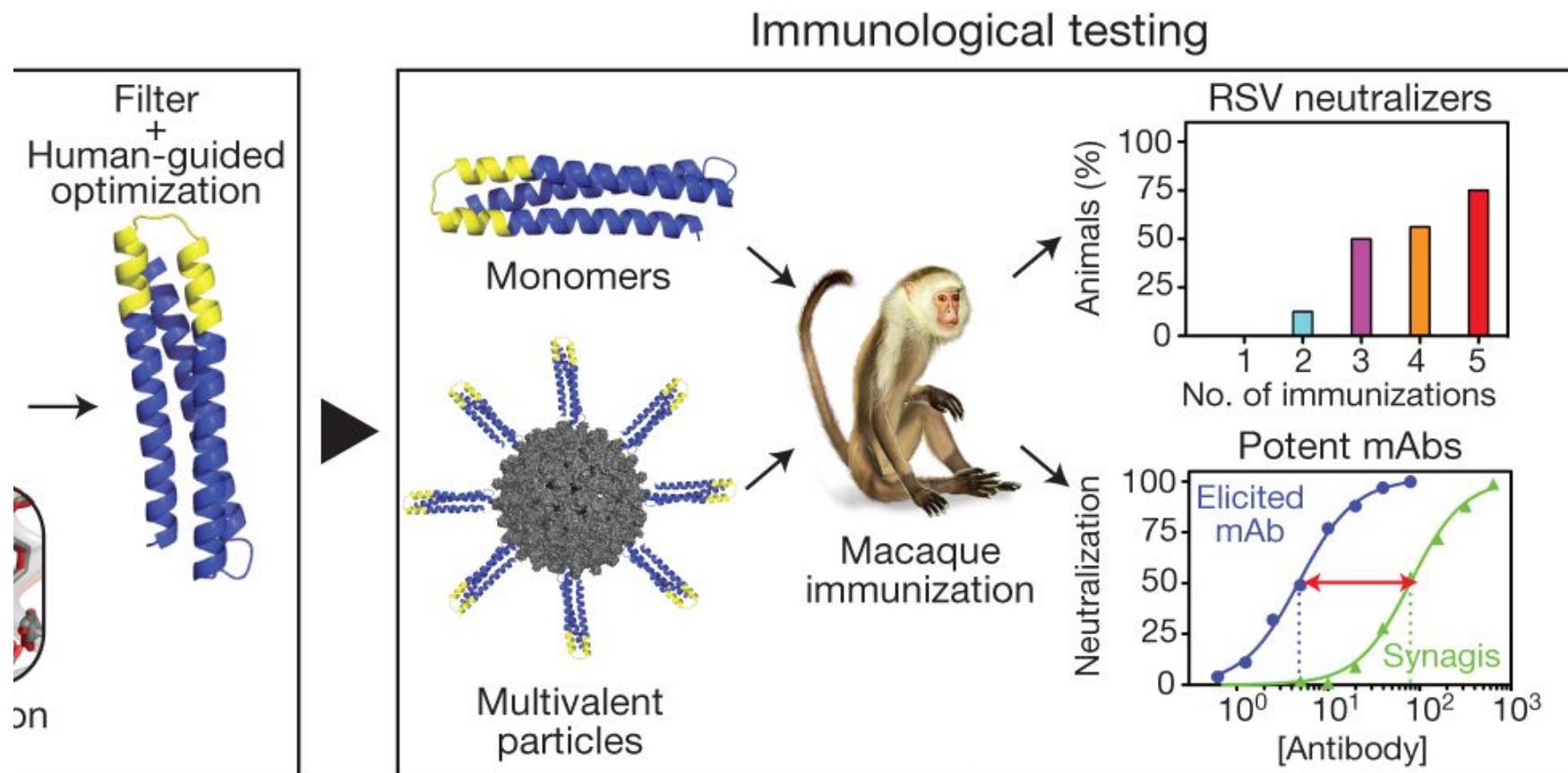
key difficulties:

- precise control over irregular functional geometries
- often polar recognition
- switchable states (not deep minima)

# An approach to circumvent this problem: “transplant” functional region to new stable protein



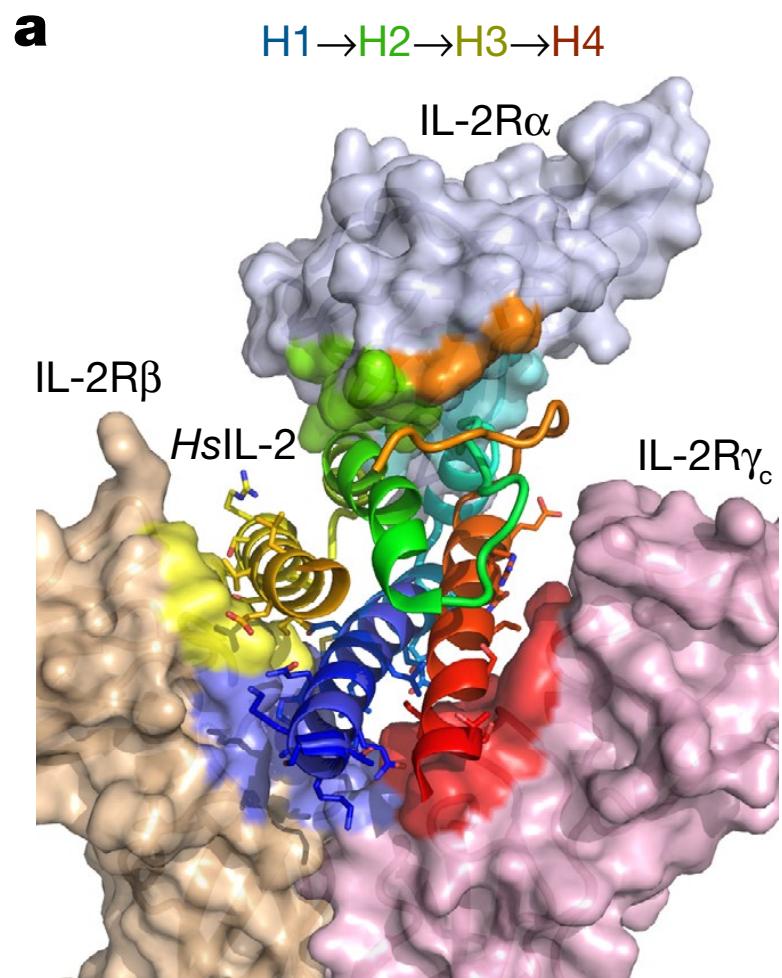
# Proof of principle for epitope-focused vaccine design



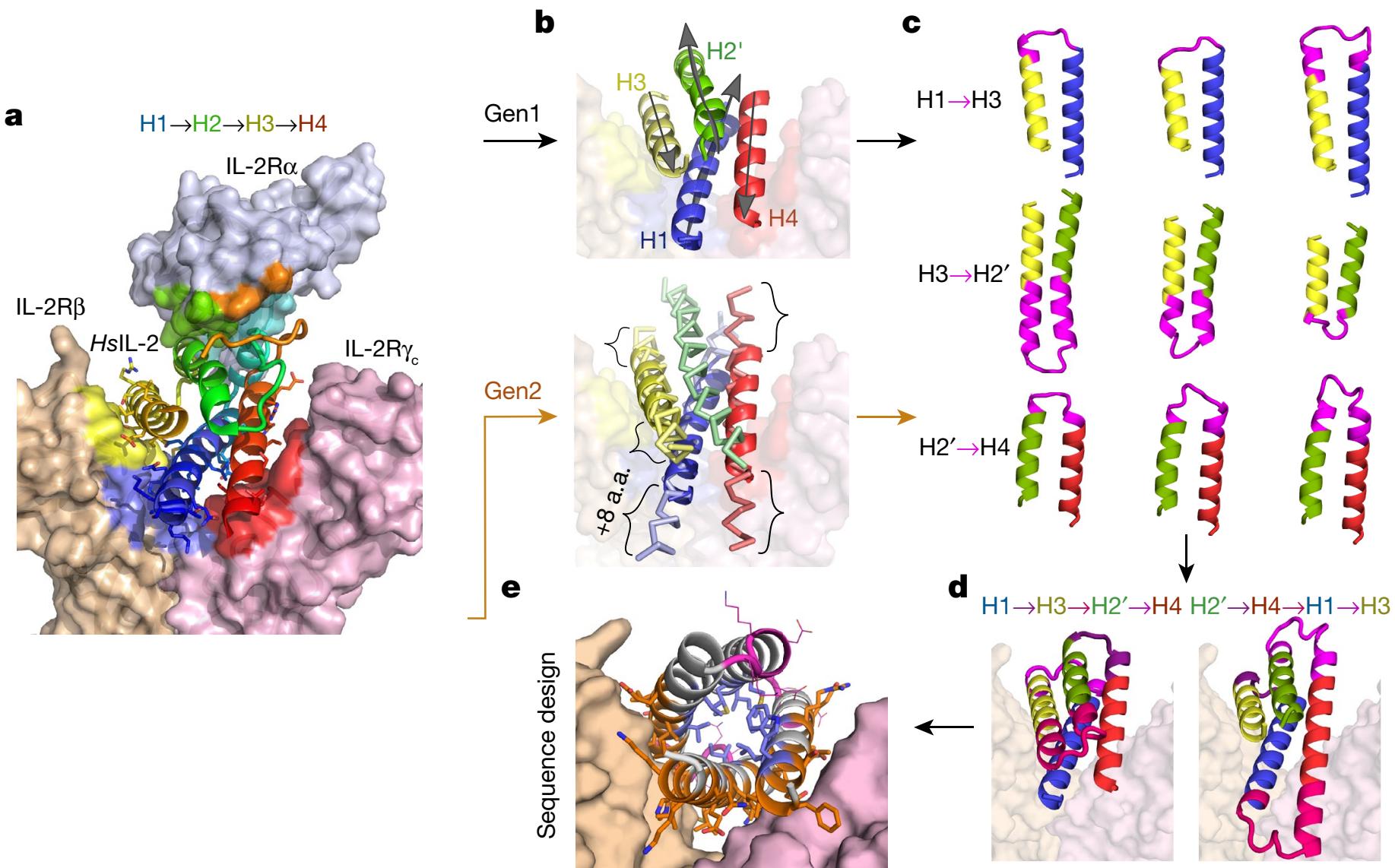
# *De novo* design of potent and selective and protein mimics as therapeutics (case study IL-2)

Challenges with using IL-2 as drug:

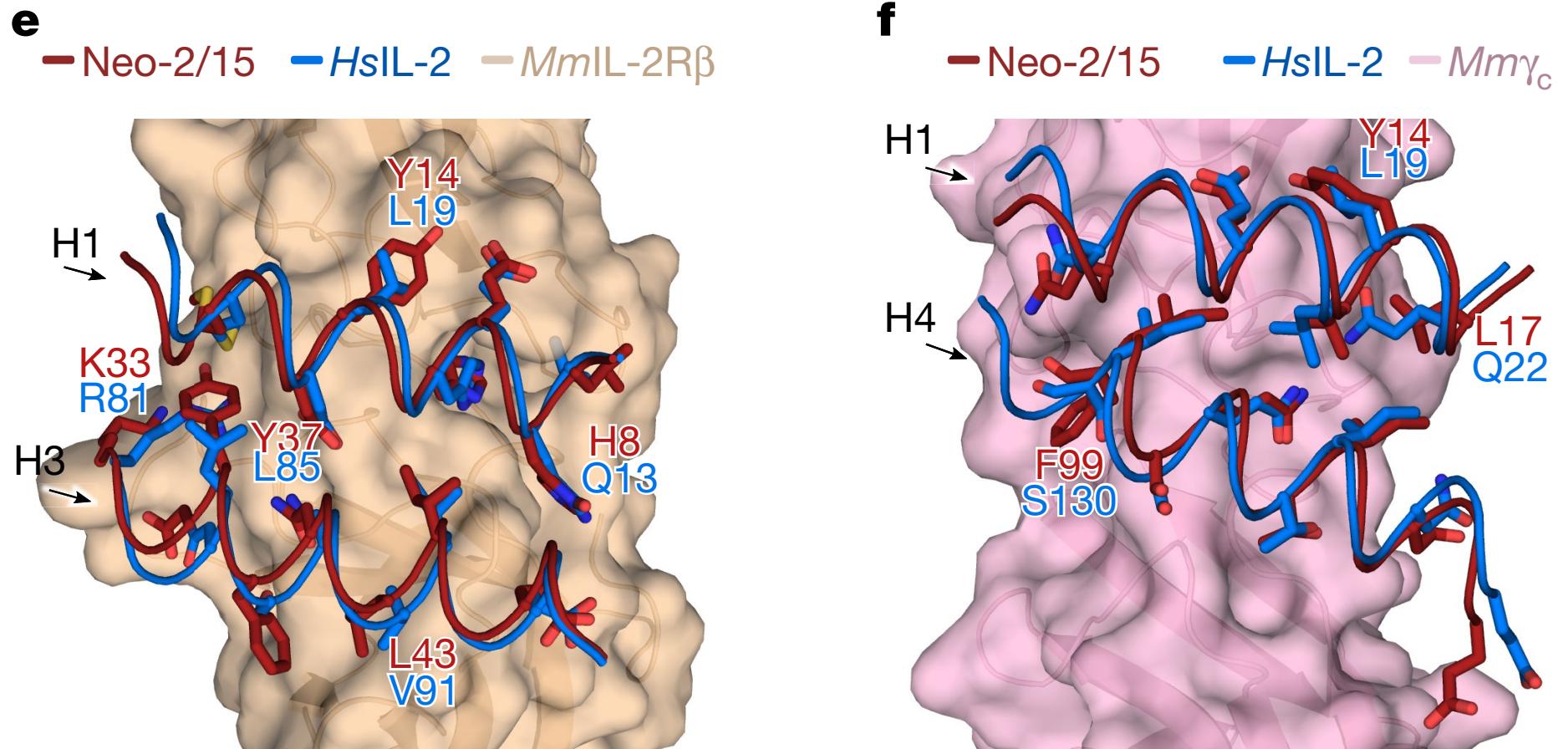
- marginal stability, aggregation
- toxicity (perhaps by interaction with IL-2R $\alpha$ )
- immune response will also target endogenous IL-2
- previous engineering efforts compromised activity and / or stability



# A design strategy for IL-2 mimics

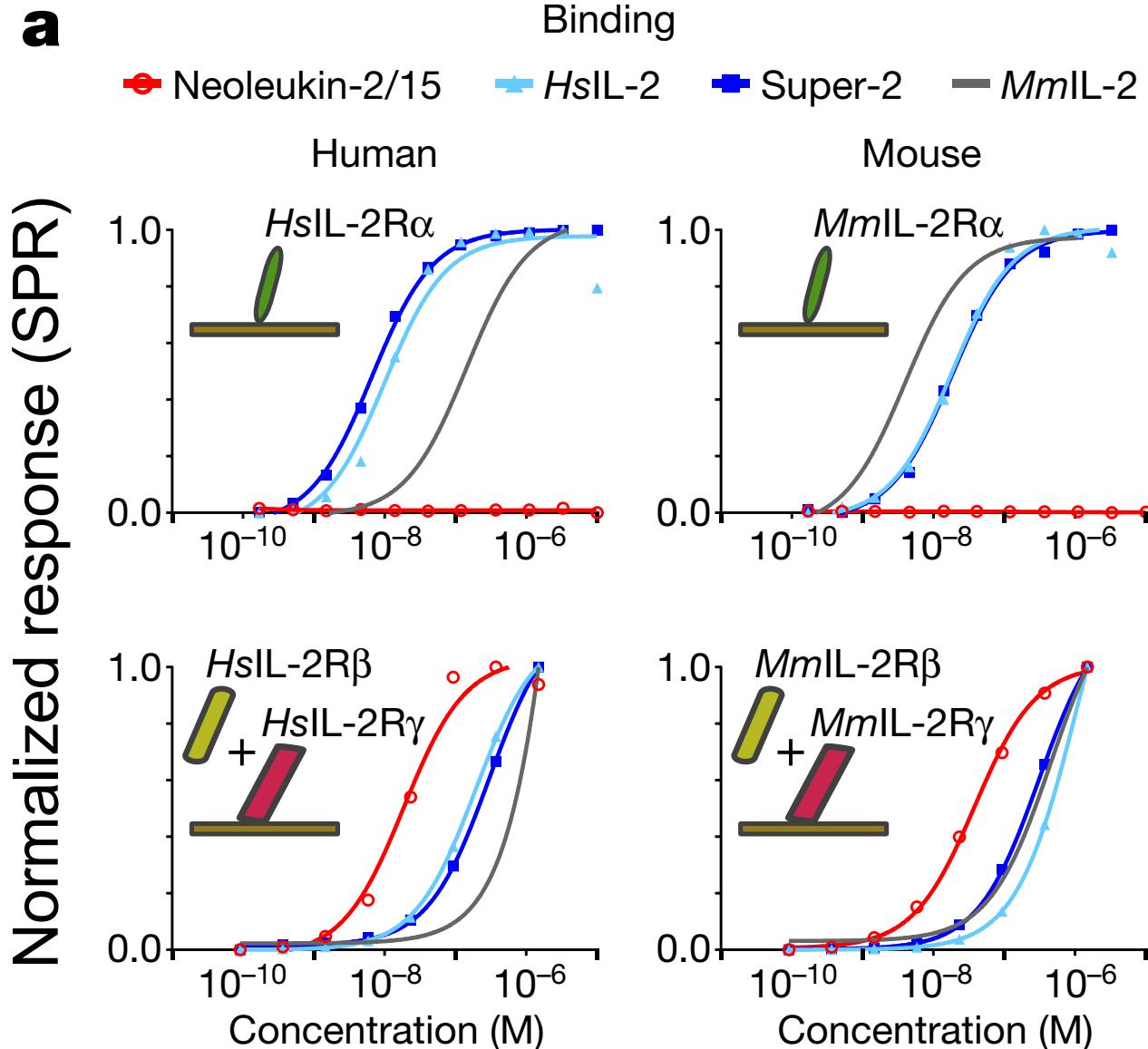


# The design structure confirms the preserved (and improved) interactions with beta and gamma



Outside these regions, Neo-2/15 is quite different from IL-2:  
Sequence identity to human 14%, mouse 24%

# Neo-2/15 has the desired binding properties (*in vitro*)

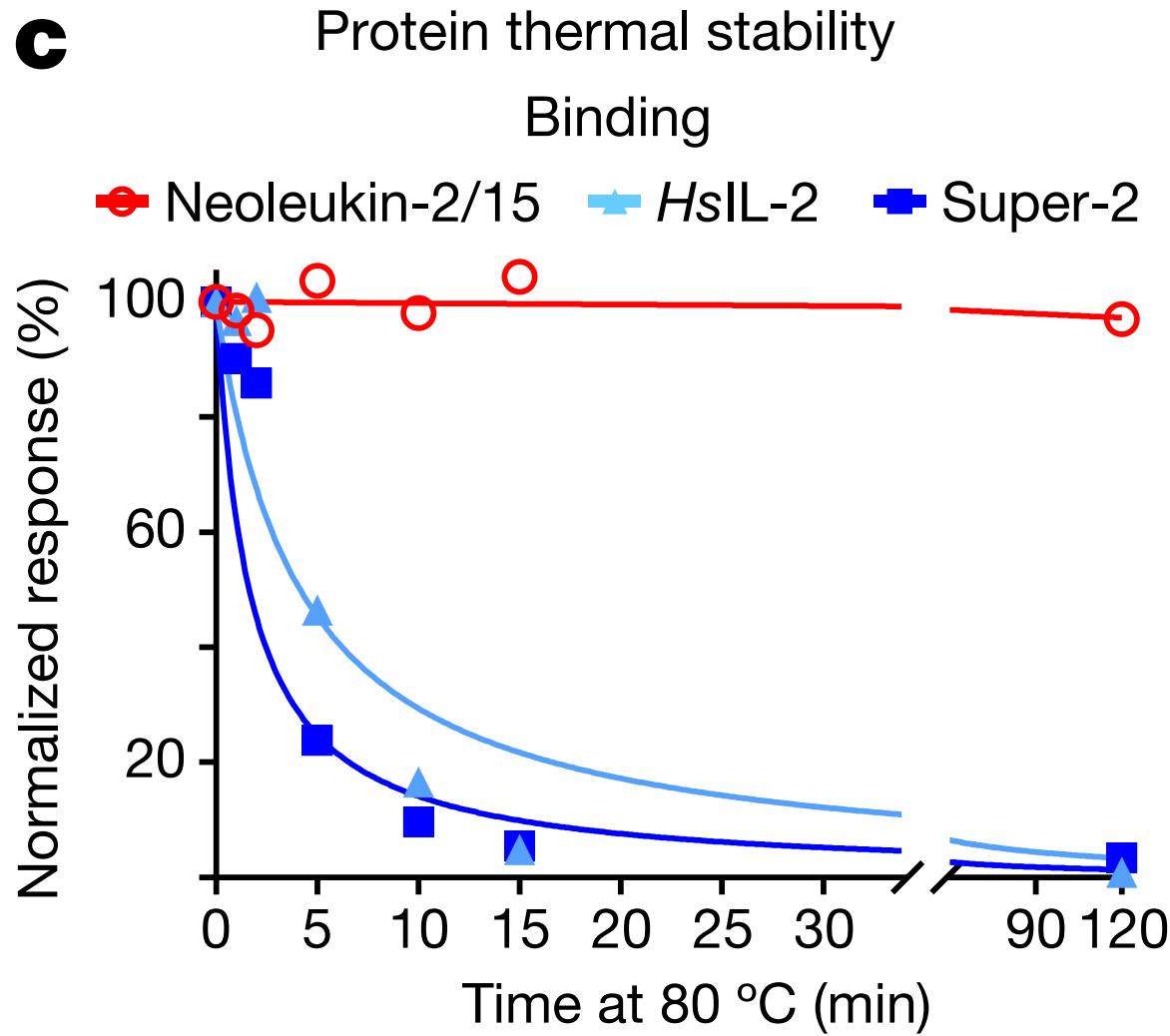


no detectable  
binding to alpha

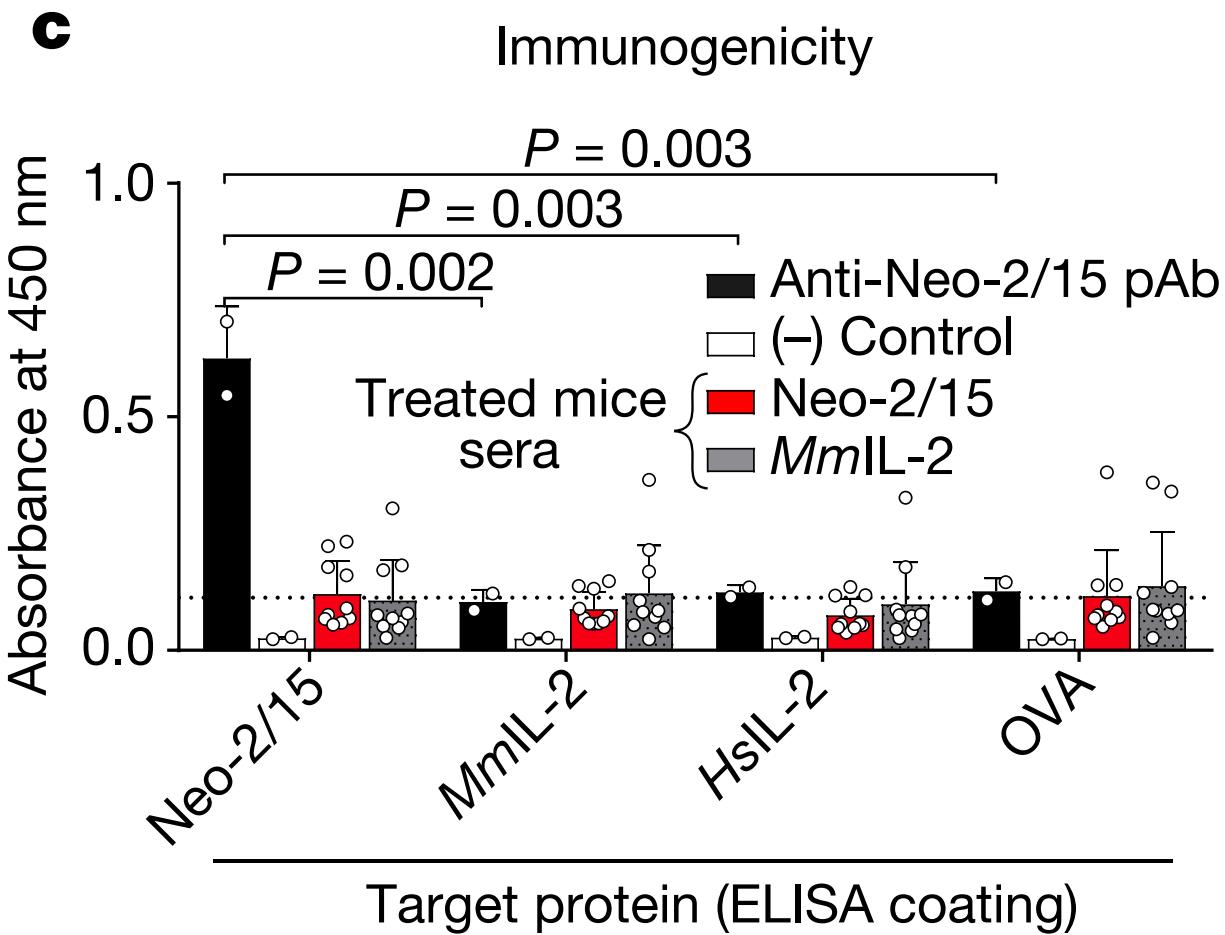
strong  
binding to  
beta-gamma

# Neo-2/15 has increased thermal stability

---

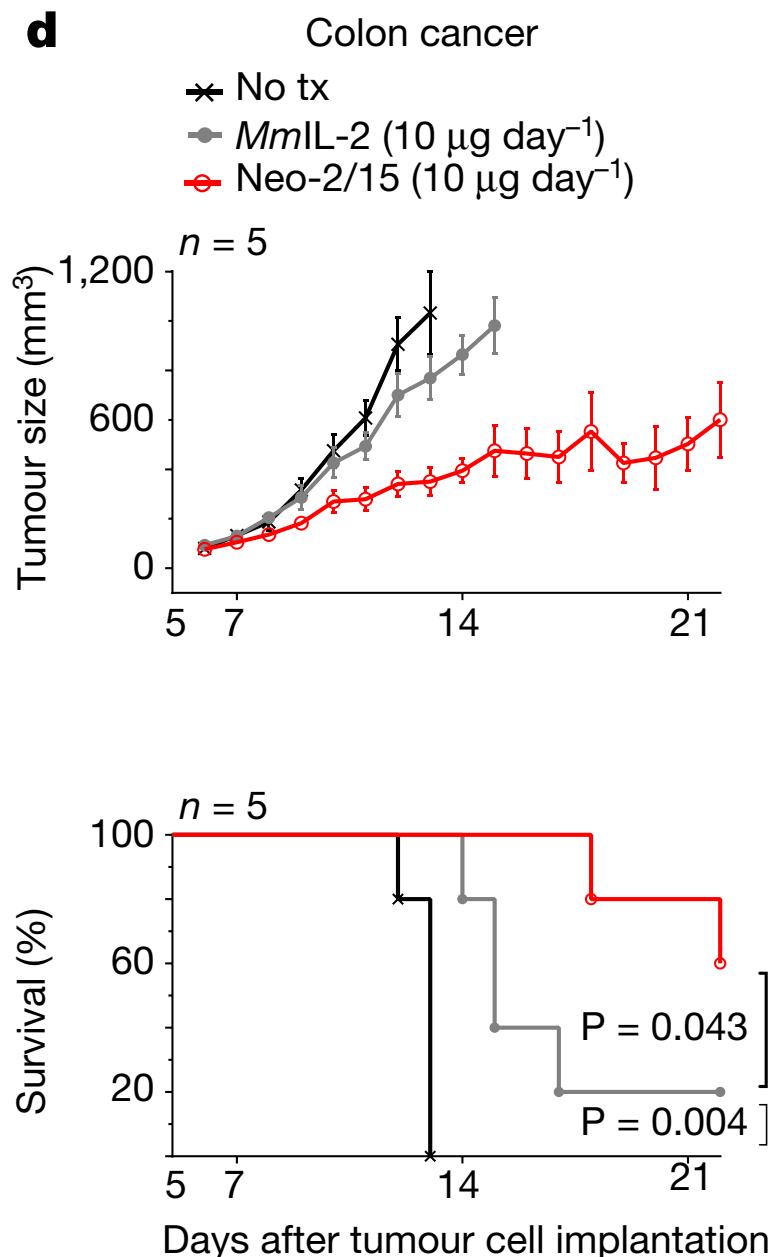


# Neo-2/15 shows limited immunogenicity



Anti-Neo-2/15 polyclonal antibody does not cross-react with human or mouse IL-2

# Neo-2/15 is more effective than mouse IL-2 in a colon cancer model



## In summary

---

- enormous progress in *de novo* design of protein structures
  - promise to create fine-tuned new architectures for many new functions
- 
- design of function is more challenging
  - some of these challenges can already be overcome by building known functional elements into *de novo* architectures with improved properties

