

Leave no amino acid unturned:
de novo design, directed evolution, and fitness
landscape exploitation leads to potential Influenza
therapeutics.

Tim Whitehead

Michigan State (via Baker Lab)

Collaborators



David



Sarel



Aaron Chevalier



Eva Strauch



Jacob Corn



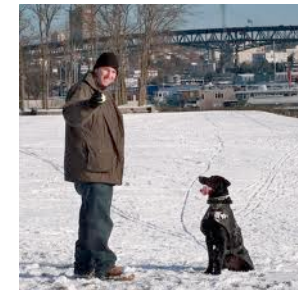
Ian Wilson



Damian Ekiert

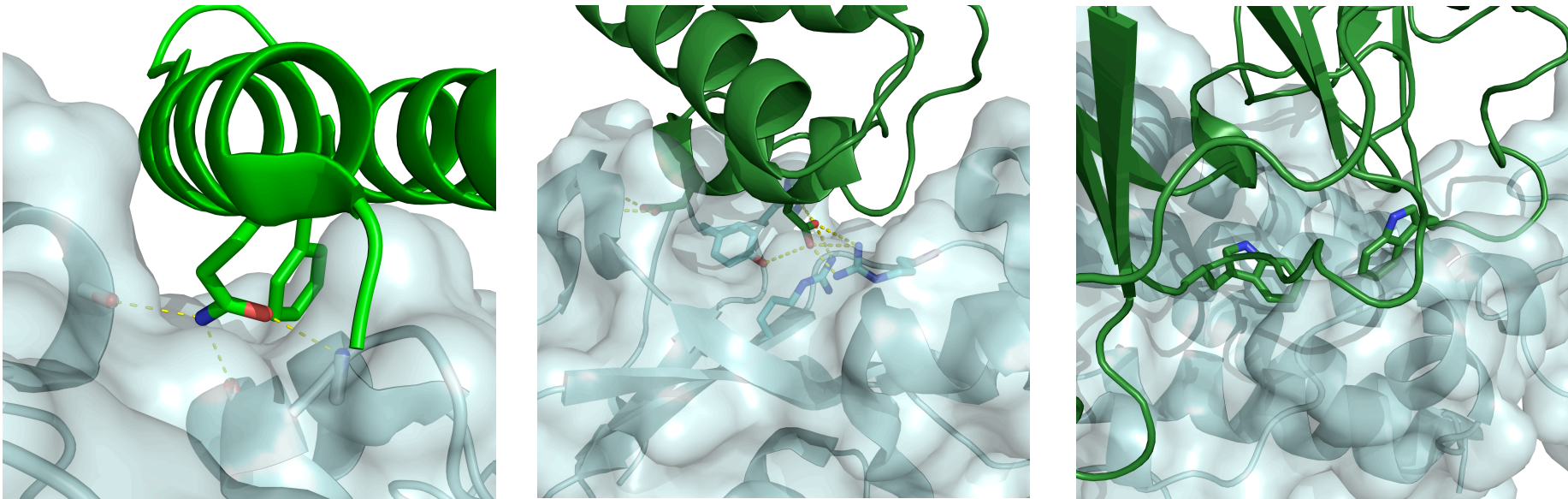


Cyrille Dreyfus



Yifan Song

Native protein interfaces come in all shapes and sizes: how do we design new interfaces?

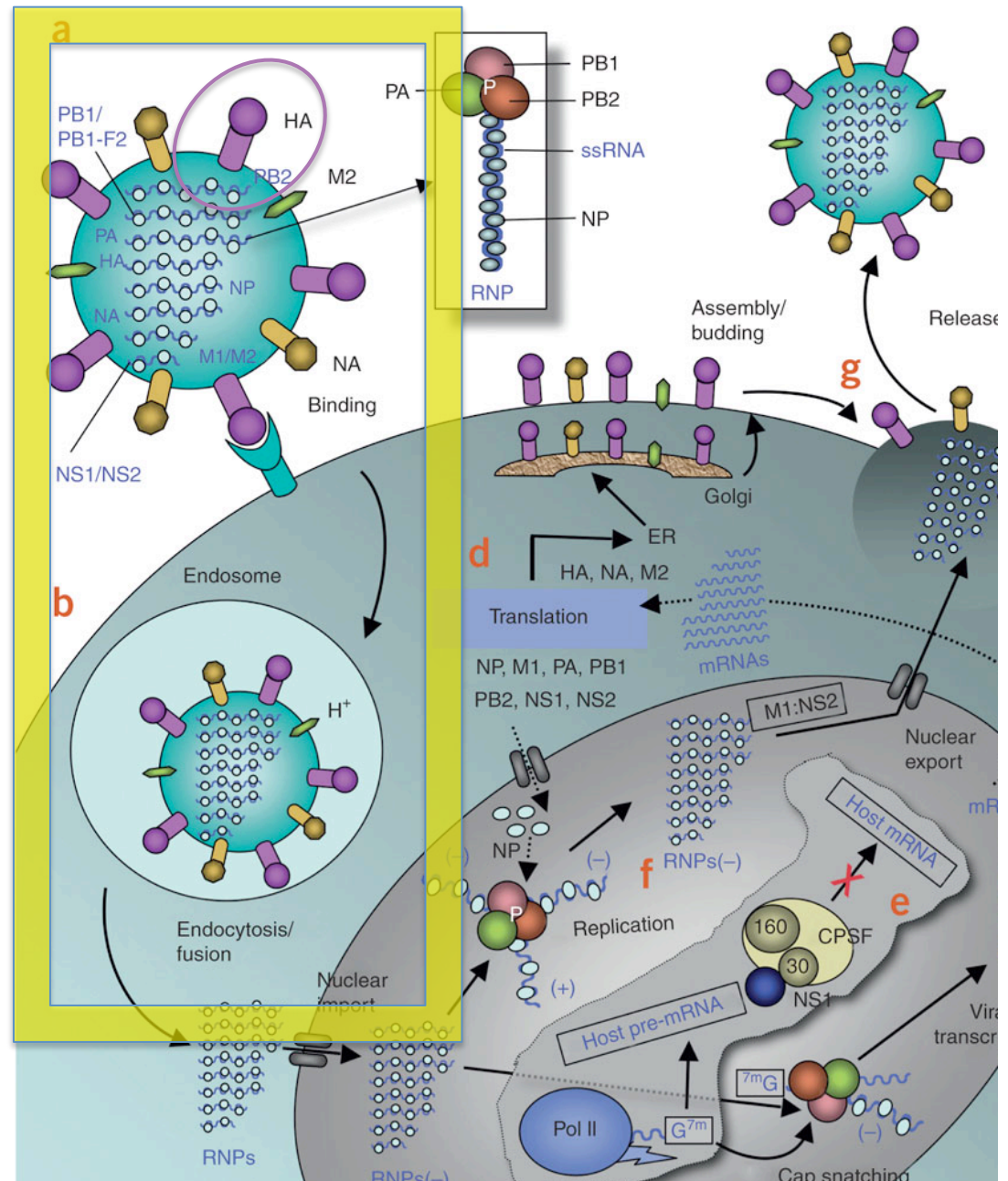


Problem statement: given a patch on a target protein surface, can one design a protein that interacts with high affinity and specificity to the targeted patch?

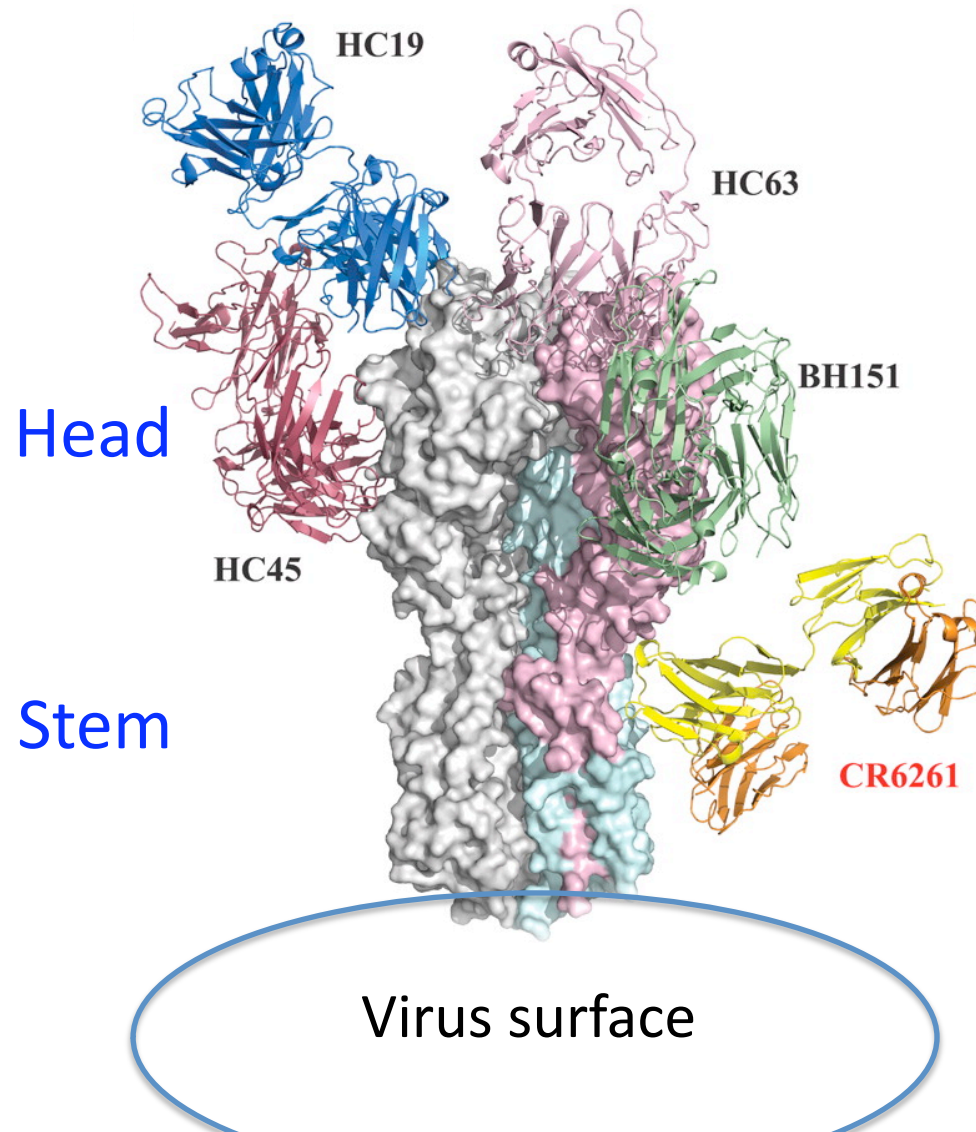
Influenza

- Pandemic strains capable of millions of deaths per year
- Vaccines are strain-specific – surface proteins mutate too rapidly to develop panflu vaccine
- Most recent H1N1 (Swine flu) isolates are resistant to Tamiflu (98% in US)*
- One of two major surface proteins is *hemagglutinin (HA)* (e.g. H1N1, H5N1)

*WHO Influenza A(H1N1) virus resistance to oseltamivir – 2008/2009 influenza season, northern hemisphere 18 March 2009



Human antibodies typically bind to the *head* of influenza hemagglutinin (HA) which mutates every year. Only CR6261 binds to the conserved *stem* region, *inhibiting* virus infectivity.

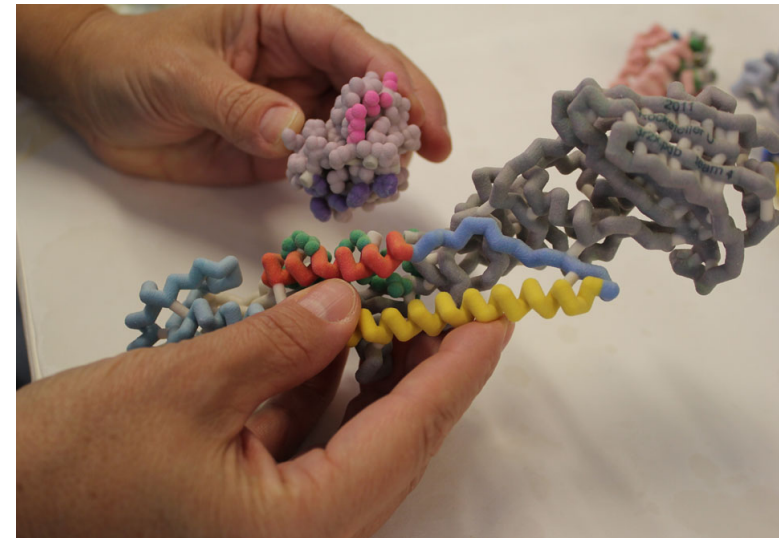
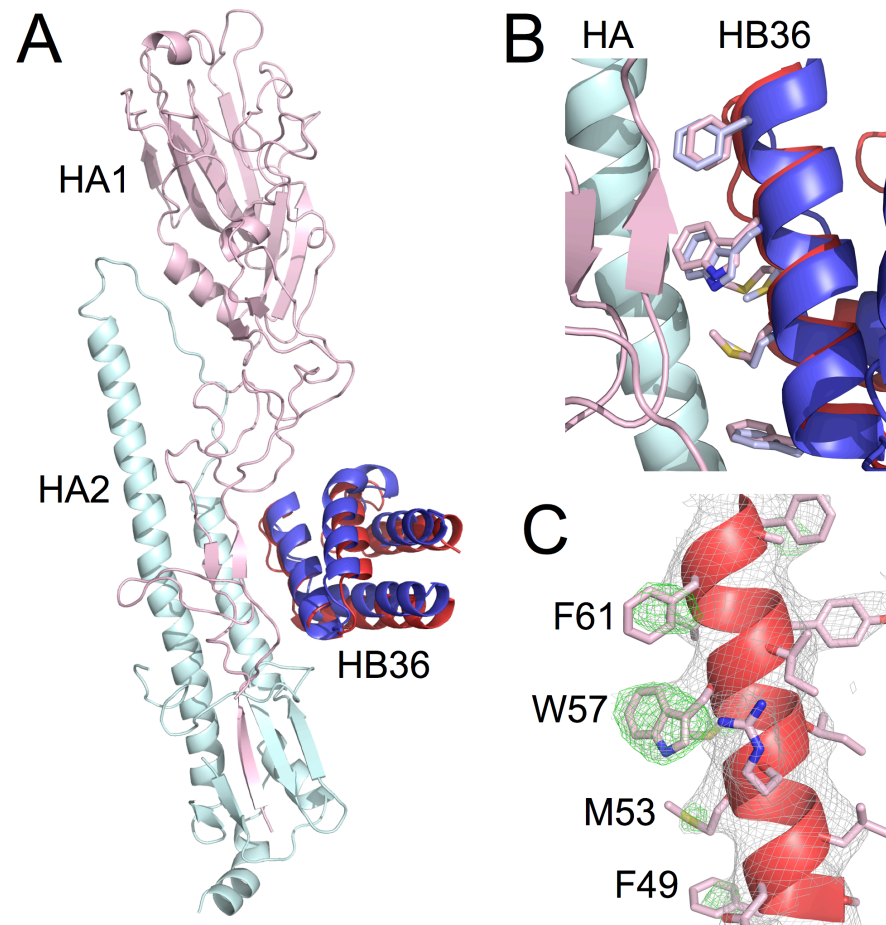


CR6261 will not be used as a frontline therapeutic

- Extremely high costs for antibody production (~\$100/g)
- Extremely high dosing required (5-10 mg/kg)

Our goal is to **design** small proteins targeting the *stem* region, inhibiting the virus

Two de novo proteins were designed that target an epitope conserved among group I influenza A virus



Courtesy <http://cbm.msoe.edu/>

Great! But.....

With Sarel Fleishman

...both initial hit rate (2-3%) and starting affinities ($\gg 1$ μM in vitro) are discouraging

How do we fix this?

If we better understood the physics underlying molecular recognition, our success rate and starting affinities would be much higher

Other factors can undoubtedly be underlying causes for failure (i.e. protein doesn't fold correctly, binding conception is wrong, assay is crap)

We can begin to understand the modeling deficiencies in the energy function by affinity maturation

Affinity maturation identifies deficiencies in modeling yet the data is sparse

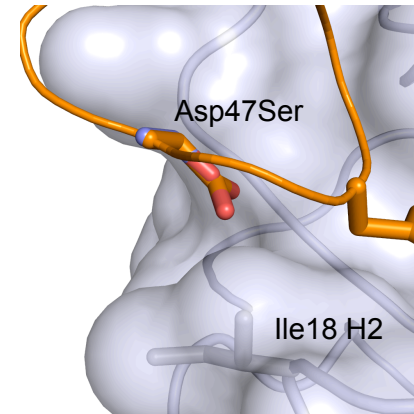
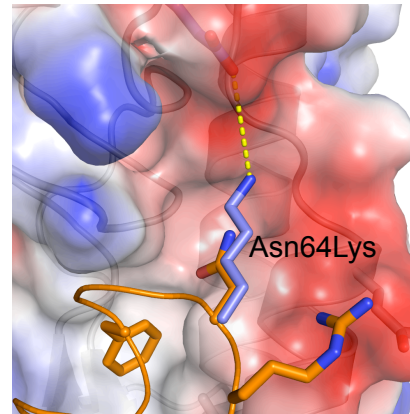
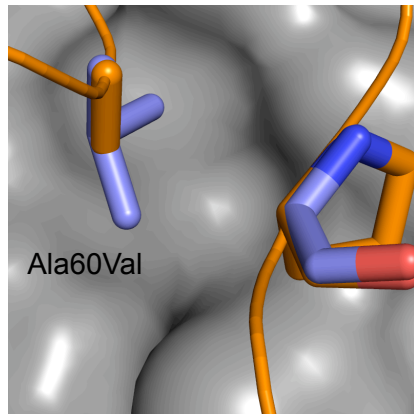
Design

Repulsive
Interactions

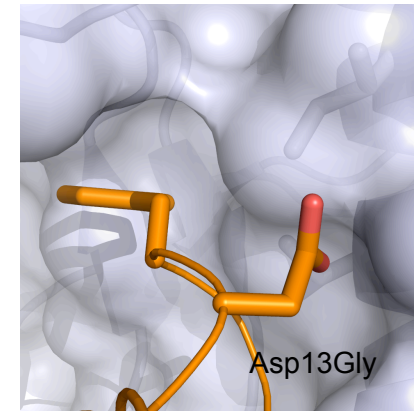
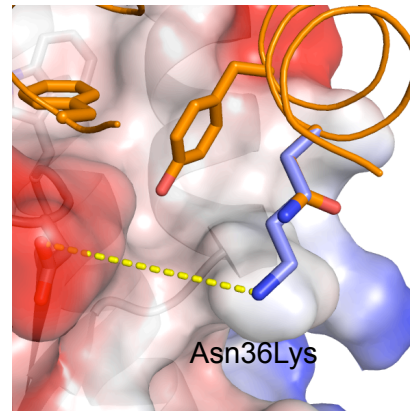
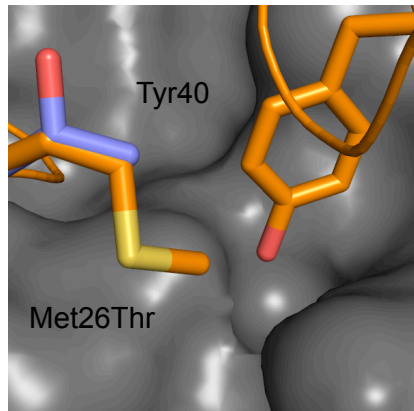
Electrostatics

Solvation

HB36

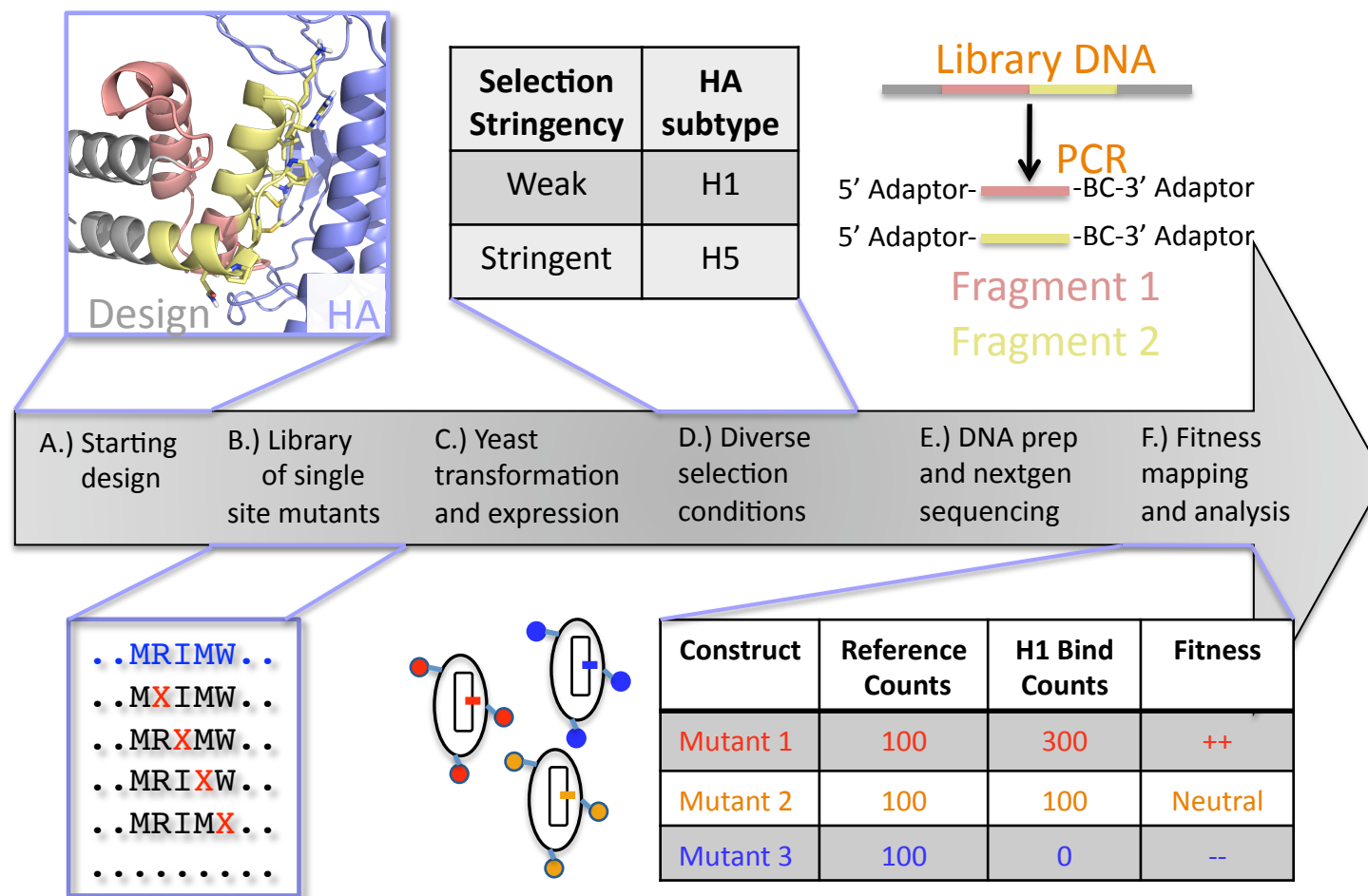


HB80



Can we develop an experimental method that determines the fitness of every single point mutant in a gene?

We extended deep mutational scanning procedure for use in yeast display and to cover entire genes

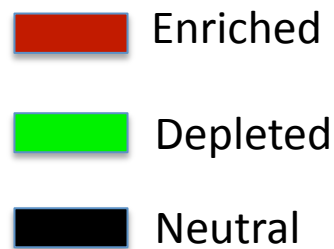
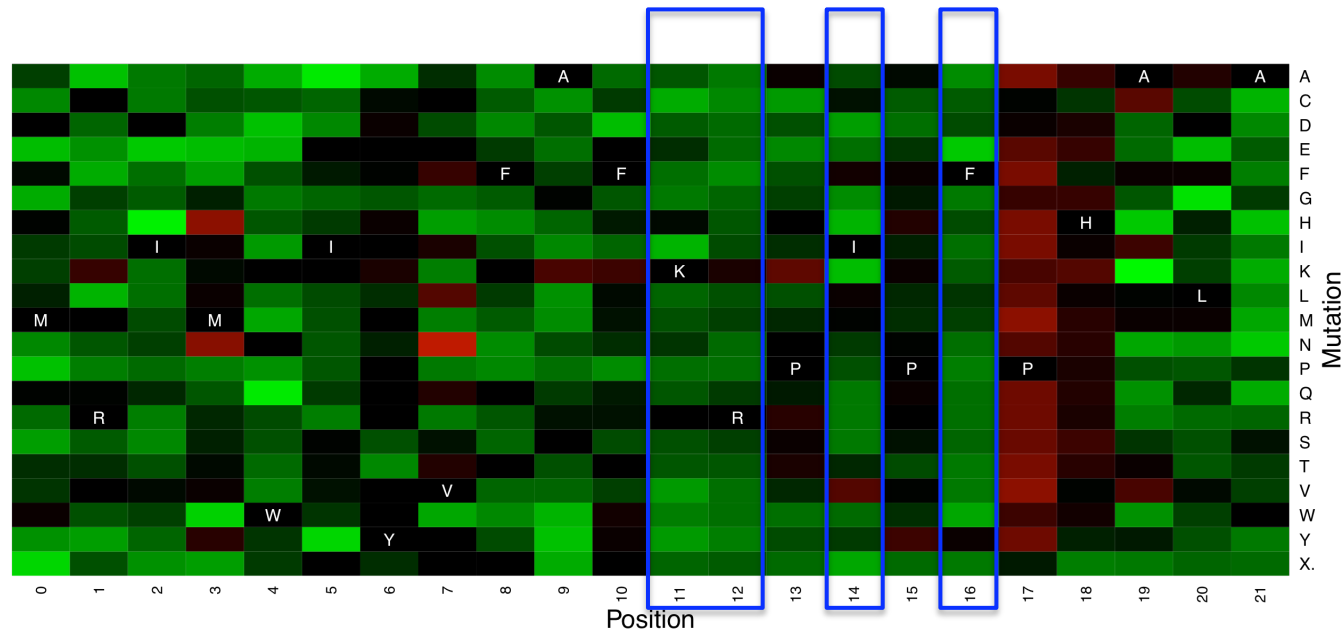


2 designs

30 selected populations

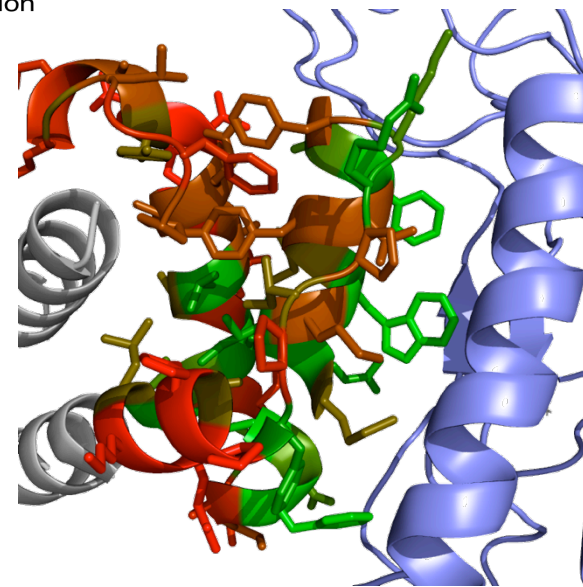
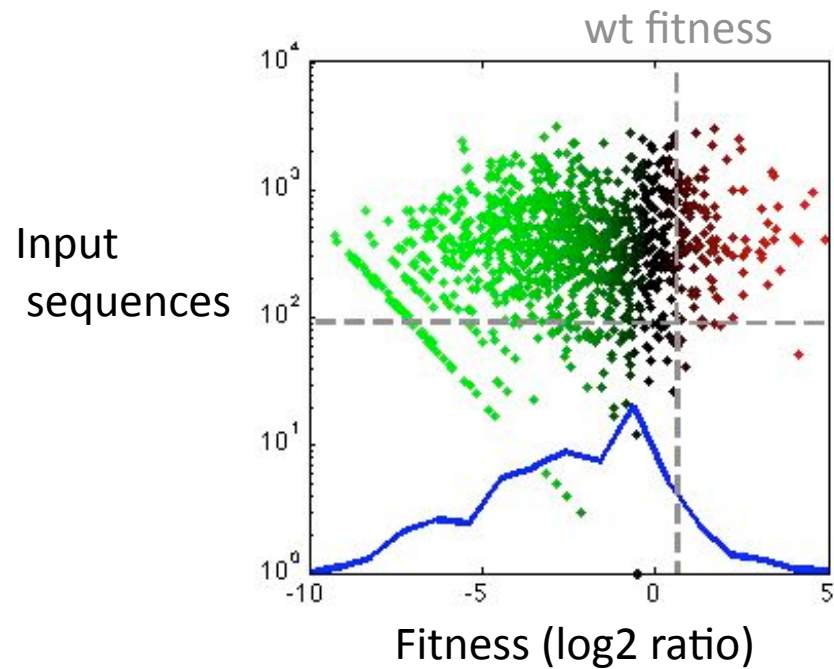
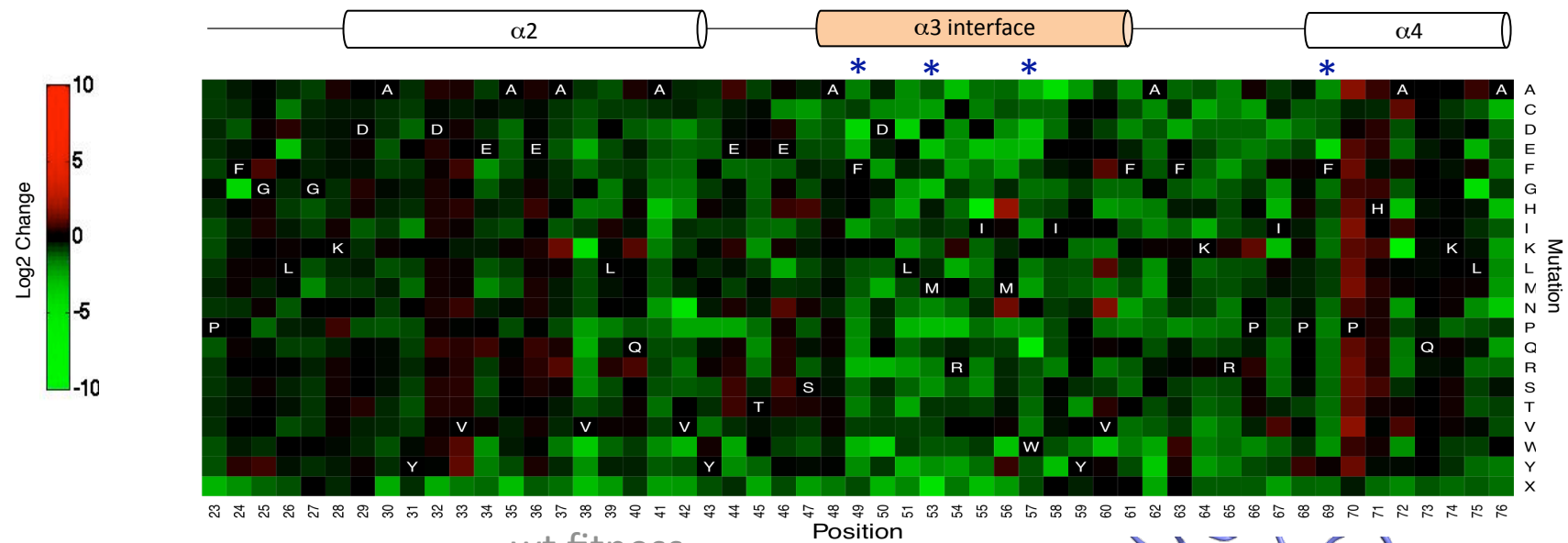
1.5 million reads median

Deep mutational scanning gives a fitness map for every single point mutation in a protein sequence



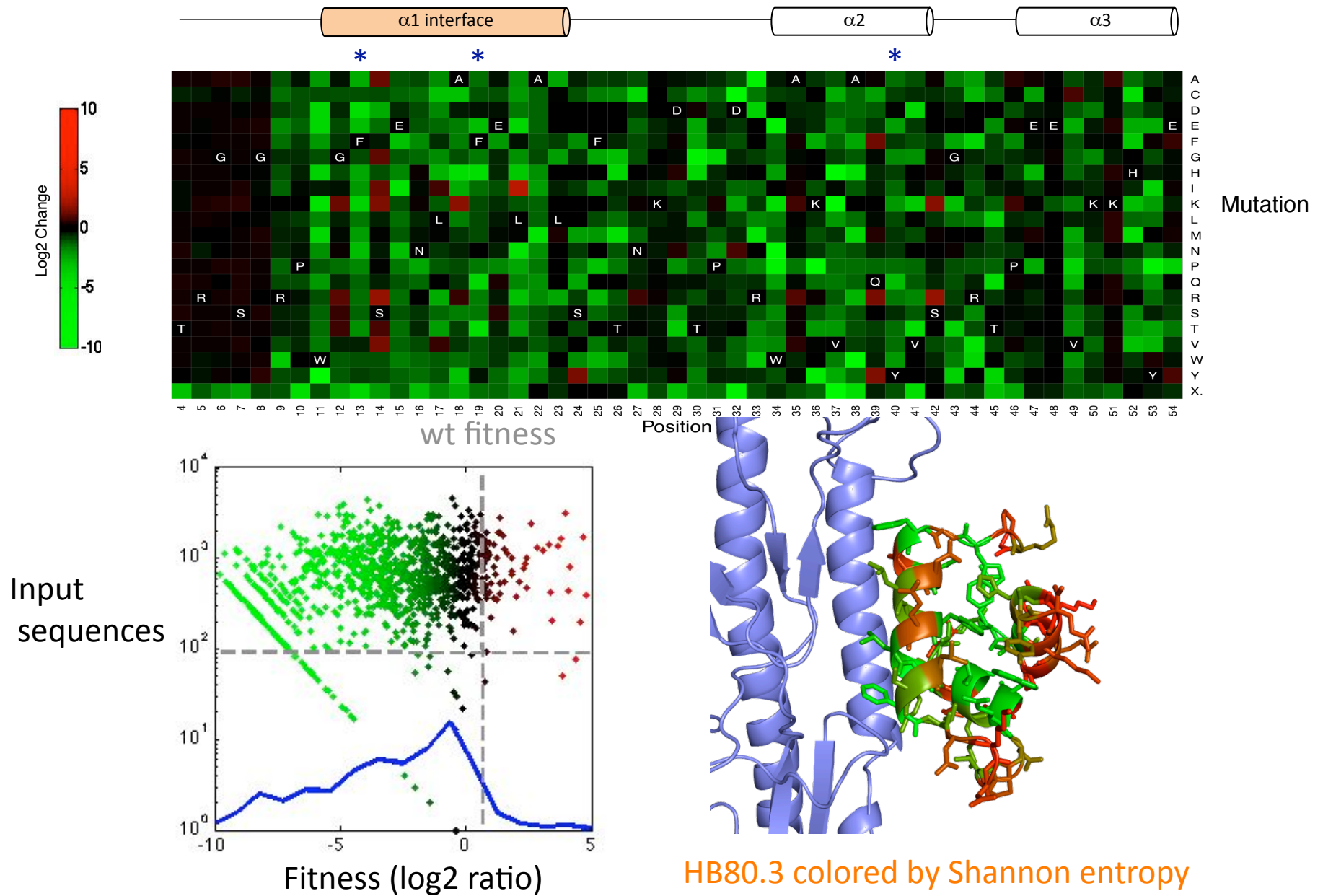
With Aaron Chevalier

Fitness landscape for HB36.4 design

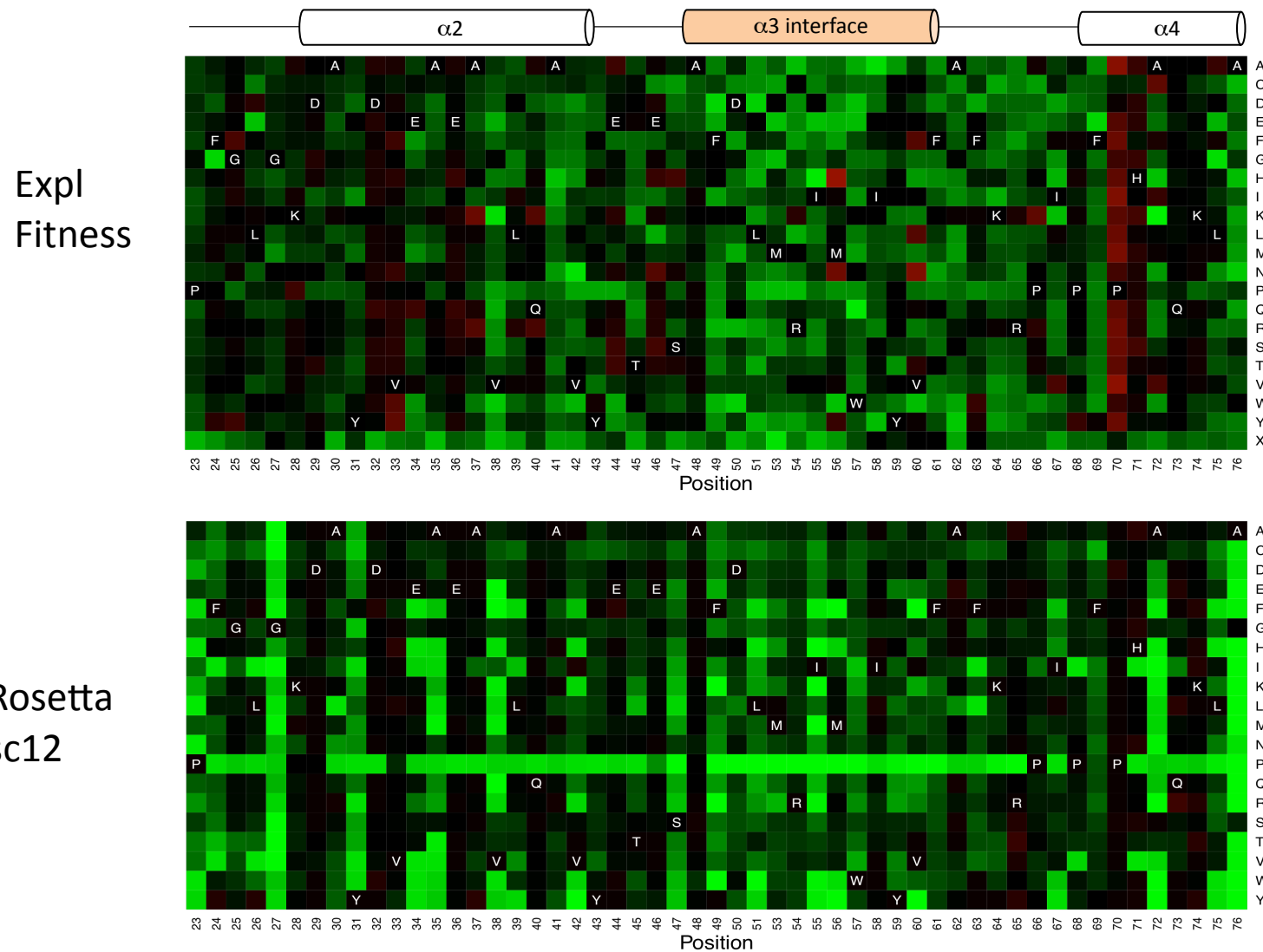


HB36.4 colored by Shannon entropy

Fitness landscape for HB80.3 design



Rosetta cannot identify beneficial substitutions in HB36



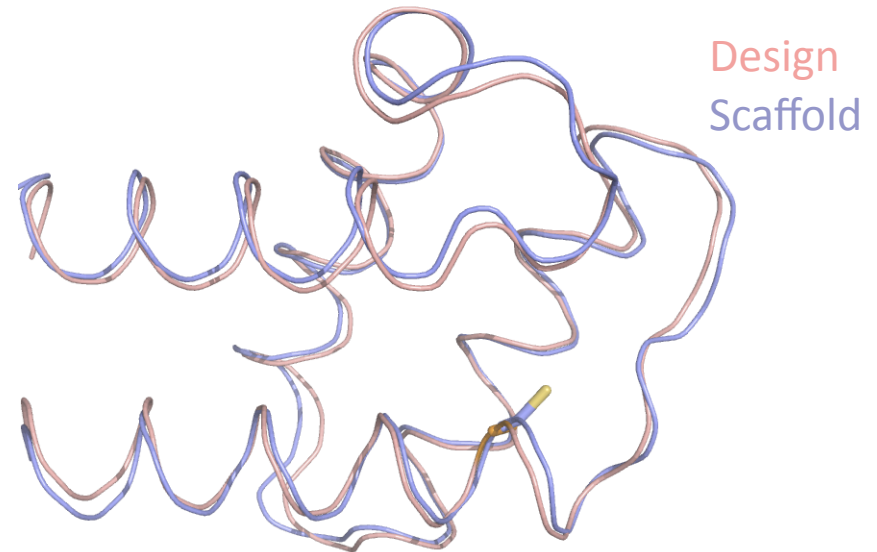
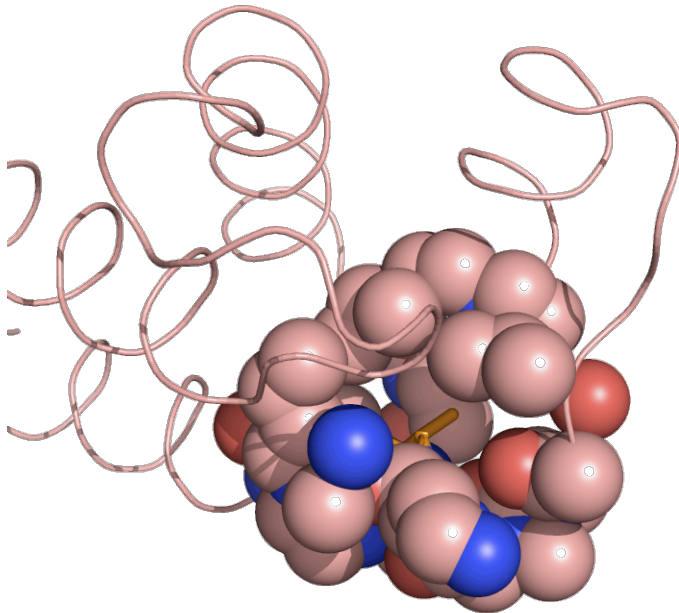
Lesson #1: don't mutate the core indiscriminately!

Deep sequencing found a
A72C core mutation as
beneficial which Rosetta
hates.

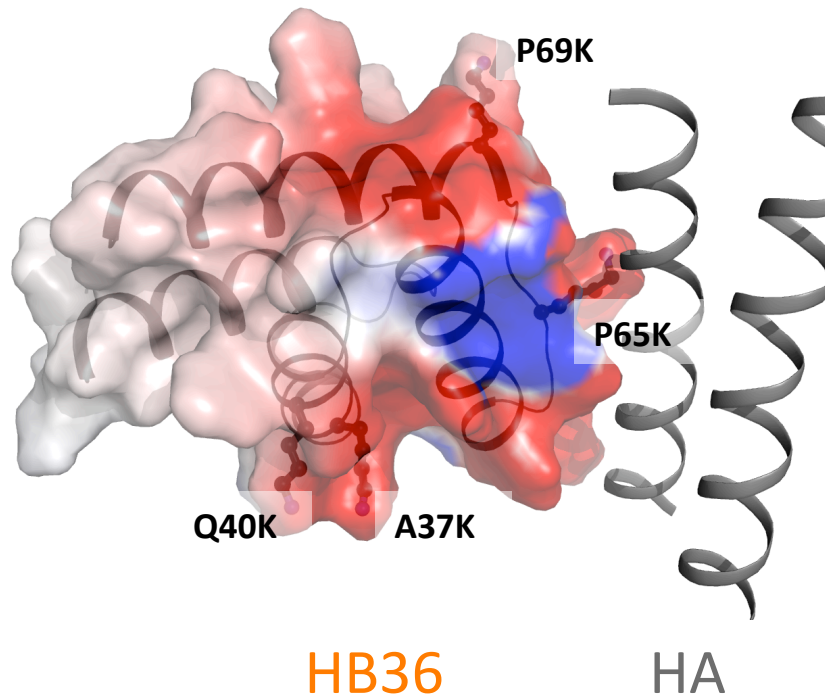
What was going on here?

A	A
C	C
D	D
E	E

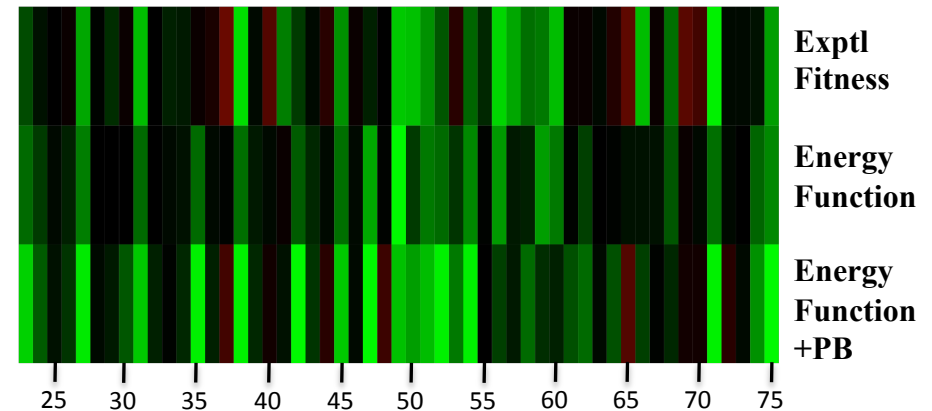
Design moves with
backrub and relax,
filling void left by cys
from scaffold



Lesson #2: electrostatics plays a much larger role in binding apolar complexes than we appreciated



All lysines substitutions in HB36

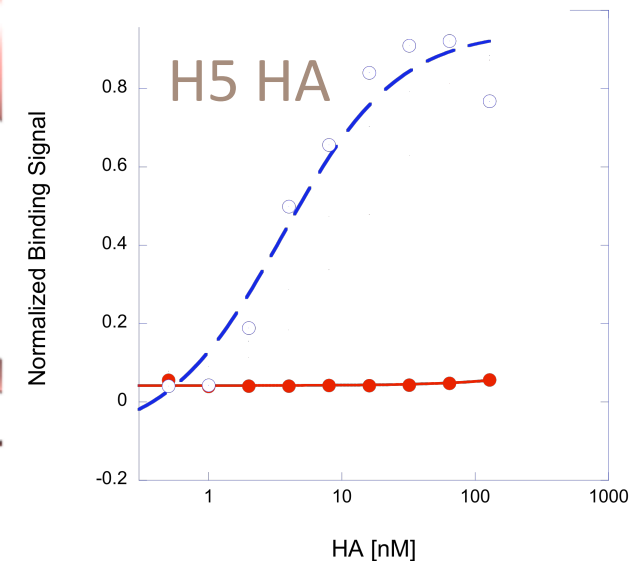
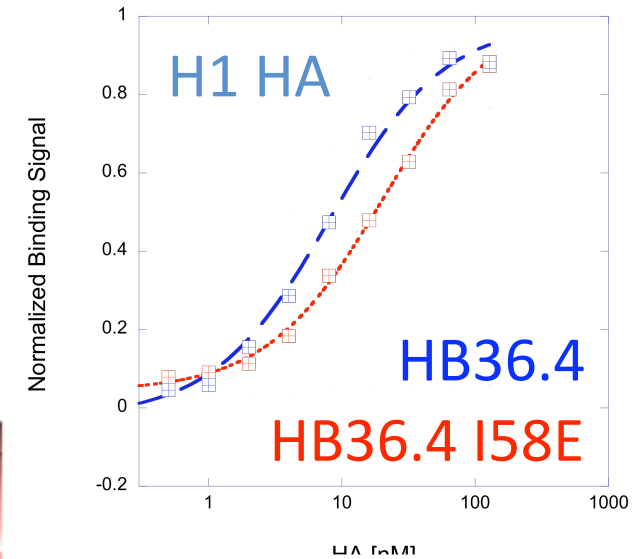
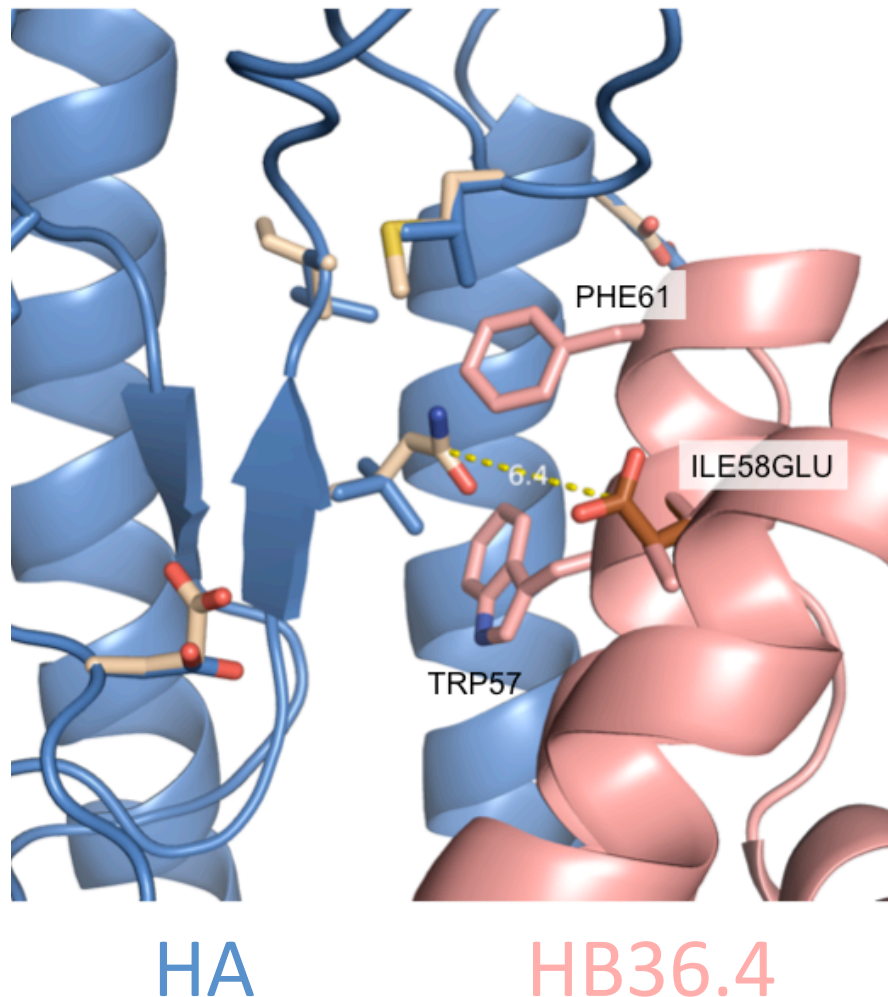


What can we do with these fitness landscapes besides learn how to better design?

- Diagnostics: engineering specificity
- Therapeutics: engineering affinity

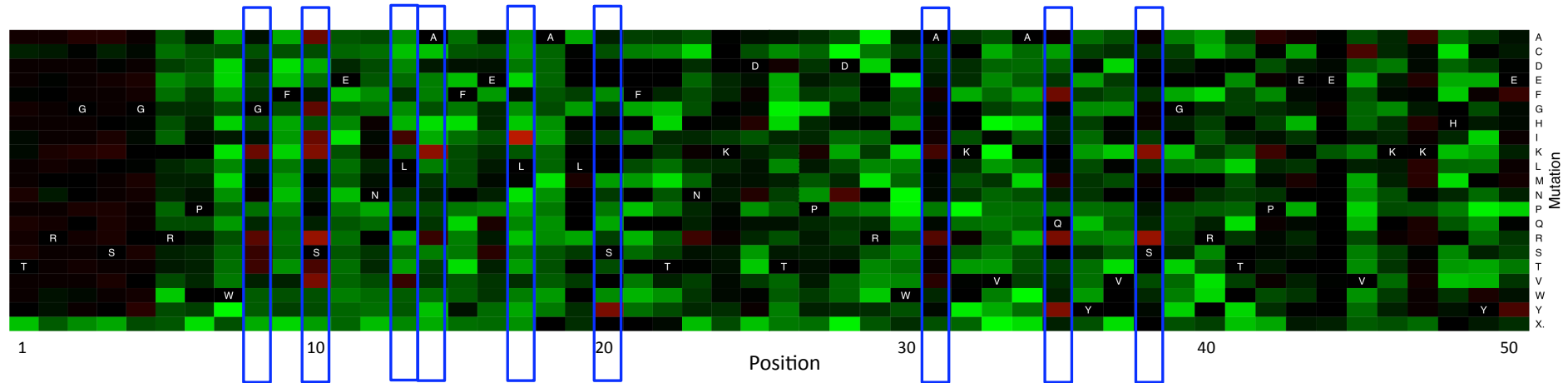
Engineering Specificity: Deep sequencing data shows us how to make H1-specific binders

Difference in HA epitope region between H1 and H5-specific subtypes



With Aaron Chevalier

Engineering affinity: beneficial substitutions are to a first approximation pairwise additive

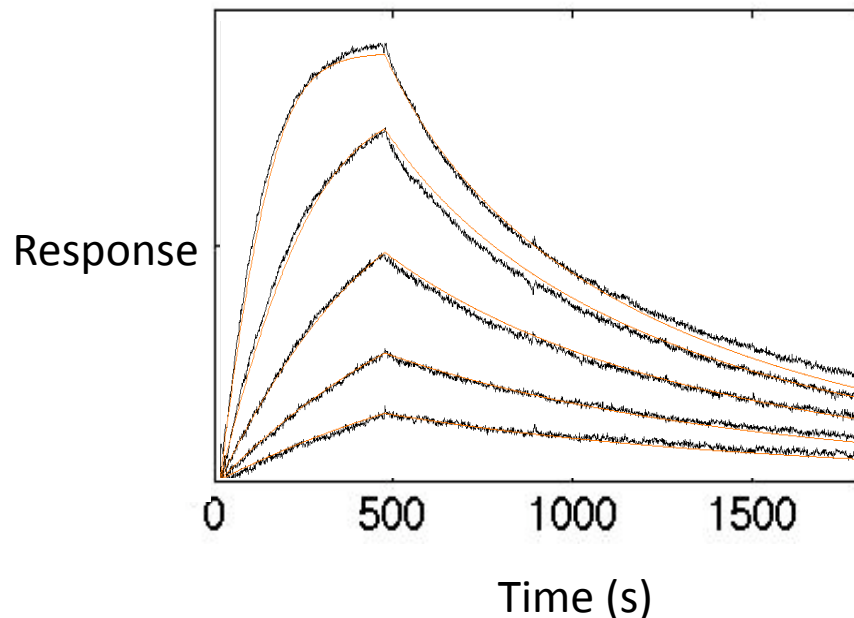


1. Create library where boxed positions can be either wt or enriched substitutions
2. Average 6 mutations per gene (range 0-9)
3. Theoretical library size $< 1 \times 10^6$
4. Stringent selection for 5 sorts using yeast display

HB80.3										
Position	12	14	17	18	21	24	35	39	42	
init. Library	K	R	I				K	Y	K	
	R	N	V	K	I	Y	R	R	R	
	G	S	L	A	L	S	A	Q	S	
	E			T			E	Stop	N	
	G			E			G	C		
							T	W		
best variant							K	H		
	K	S	I	A	I	S	K	Q	K	

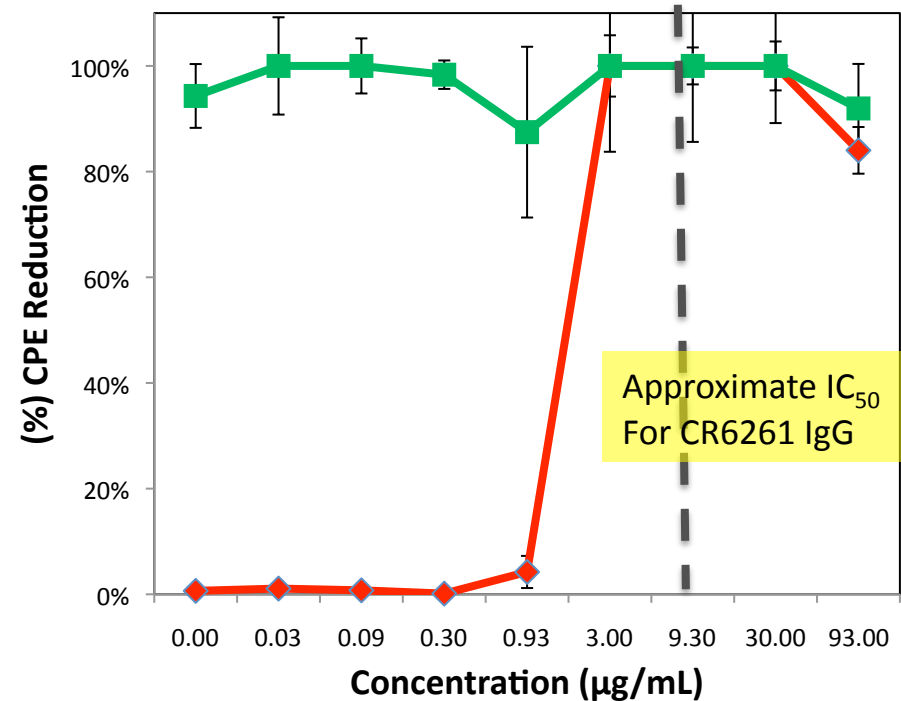
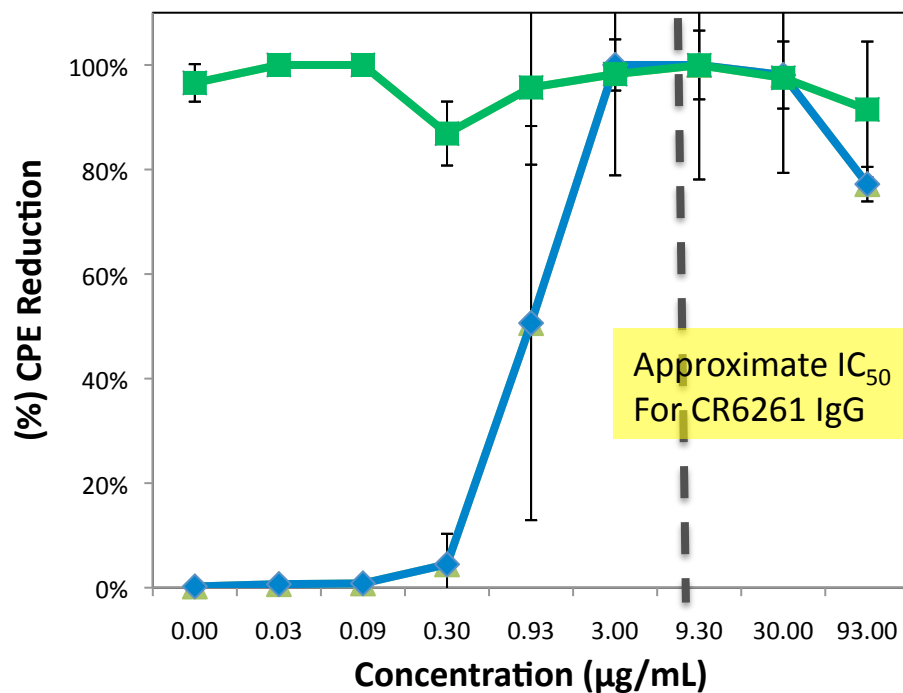
F-HB80.4 has 25-fold improved affinity and broadened specificity compared to its parent

Variant	K_d [nM]	$t_{1/2}$ [s]
F-HB80.3	18	36
F-HB80.4	0.6	550



	Strains	K_D (nM)	
		F-HB80.4	CR6261
Group 1	A/South Carolina/1/1918 (H1N1)	4.9	1.3
	A/Japan/305/1957 (H2N2)	16.7	n.d.
	A/Adachi/2/1957 (H2N2)	10.5	n.d.
	A/Vietnam/1203/2004 (H5N1)	1.4	1.2
	A/Indonesia/05/2005 (H5N1)	0.7	n.d.
	A/turkey/Massachusetts/3740/1965 (H6N2)	9.3	1.8
	A/turkey/Wisconsin/1/1966 (H9N2)	7.5	9
	A/duck/Alberta/60/1976 (H12N5)	430	-
	A/gull/Maryland/704/1977 (H13N6)	55	36
Group 2	A/black-headed gull/Sweden/4/1999 (H16N3)	540	52
	A/Hong Kong/1/1968 (H3N2)	-	-
	A/duck/Czechoslovakia/1956 (H4N6)	-	-
	A/Netherlands/219/2003 (H7N7)	-	-
	A/chicken/Germany/N/1949 (H10N7)	-	-
	A/mallard duck/Astrakhan/263/1982 (H14N5)	-	-
	A/shearwater/West Australia/2576/1979 (H15N8)	-	-

F-HB80.4 neutralizes Influenza virus at doses 10-fold less than the best-in-class antibody!



Outlook

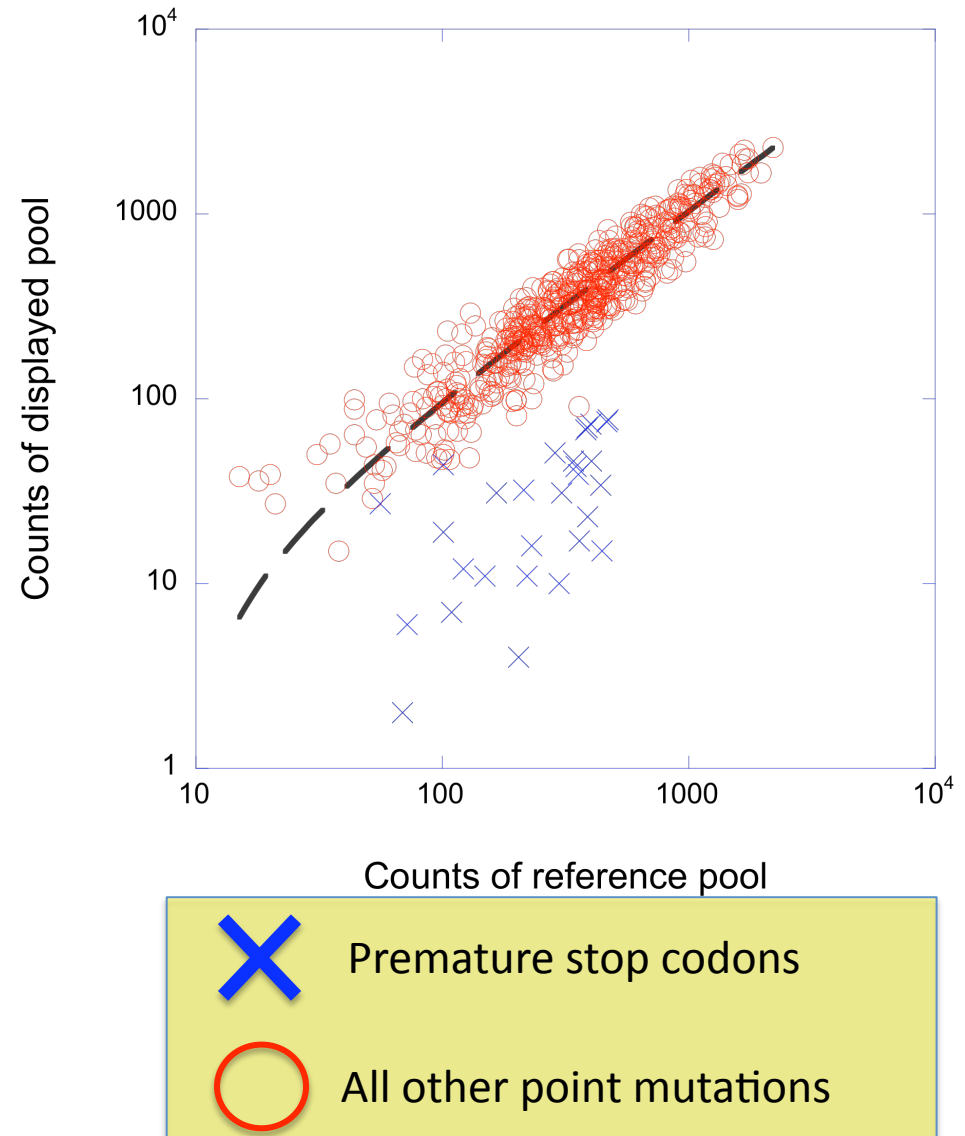
From conception of a small protein that binds a specific conserved epitope on Influenza and neutralizes the virus to a lead molecule took:

2 years, 2 postdocs, 2 grad students, 2 PI's

In 20 years, can we do the same in 2 weeks?

Surprise/Problem #1

- “yeast polypeptide fusion surface display predicts thermal stability”¹
- “proteins are subject to quality control in the ER, wherein chaperones and foldases ensure the fidelity of the secreted protein species”²



1. J Mol Biol 292, 949 (1999); 2. Handbook of therapeutic monoclonal antibodies” (2009)