

*Enabling rapid liquid and freeze-dried
formulation design for the manufacture and
delivery of novel biopharmaceuticals*

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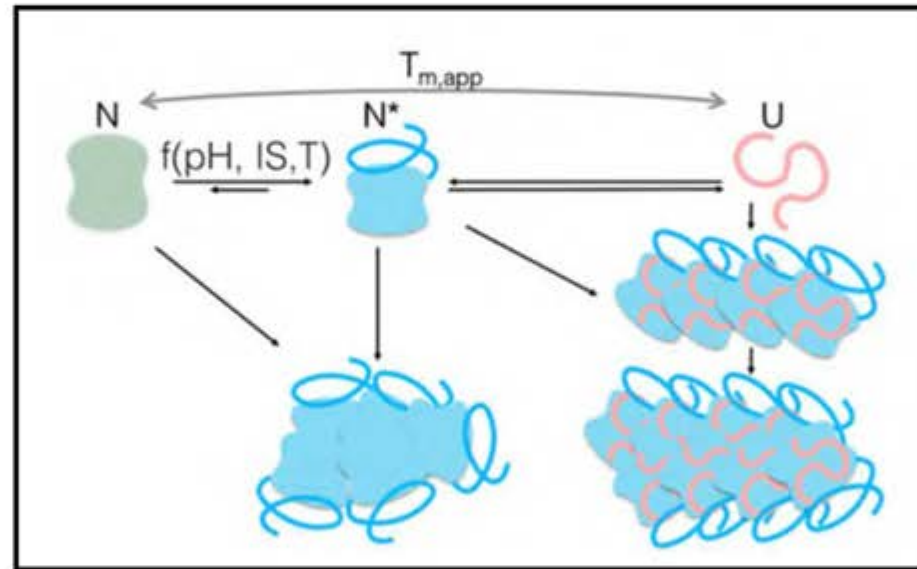
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University of Manchester



- Biological medicines account for >25% of all new drug approvals: 8 of the top 10 selling drugs
- Biopharmaceuticals market is rapidly growing with reported sales of £197 billion in 2016 (compared with total drug market of £816 billion)
- Next generation therapies are increasingly complex and engineered for biological activity at the expense of physical and chemical stability (eg protein fusions, fragments, conjugates with small drug molecules)
- Formulation development of biopharmaceuticals:
 1. Dosage formulations fixed quickly - in time for clinical trials. Not much material is available. Shelf life over 2 years not known until mid-way through clinical trials.
 2. Formulations require stability, potency, and ease of delivery to patient
 3. Many therapeutics require high concentrations which leads to increased physical degradation, poor rheological properties, and phase separation

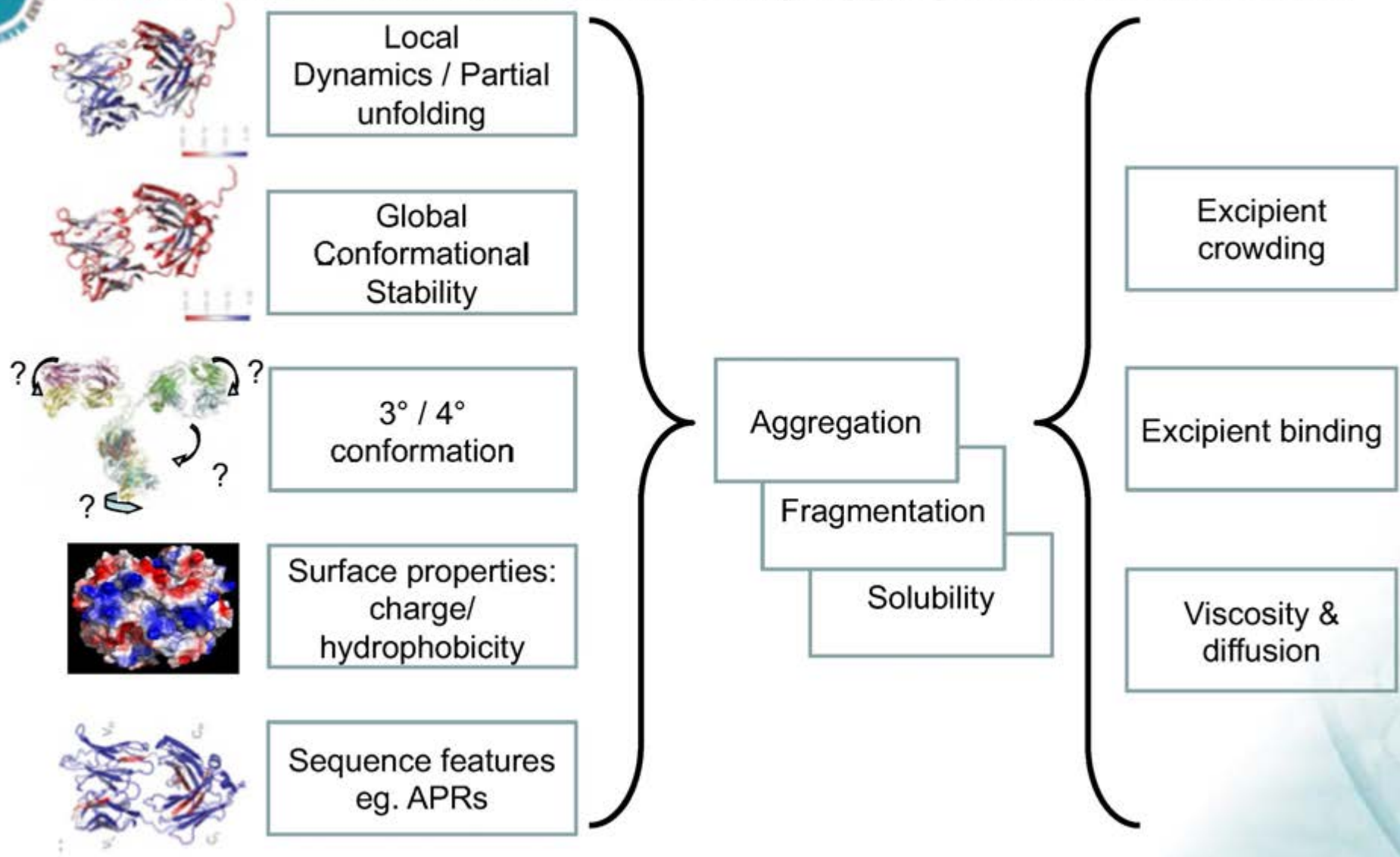
Protein aggregation

- Predicting and controlling aggregation is an outstanding challenge:
 1. Key intermediates are transient, have low populations, and are difficult to isolate / study
 2. Multiple mechanisms for aggregation, depend on protein and environment (solvent properties, temperature)



- Rapid experimental screens are too indirect:
 1. Unfolding temperature or free energy, colloidal stability (eg aggregation temperatures and protein-protein interaction measurements)
 2. Accelerated (eg. high T) aggregation assumes Arrhenius-type extrapolations

Aim1: Understand factors affecting aggregation in formulation



Aim2: How can we predict better formulations?

Do conformational and colloidal stabilities correlate to aggregation rates?

Does forced degradation at high temperature predict shelf-life?


Can alternative methods be developed for predicting aggregation rates?

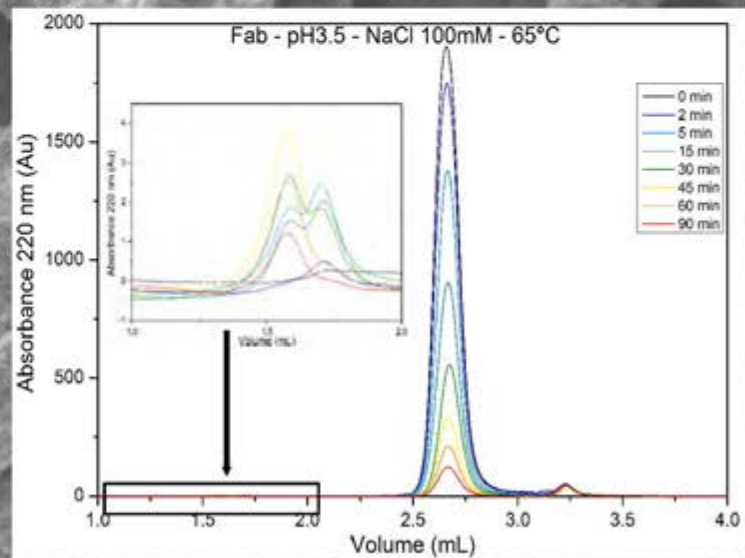
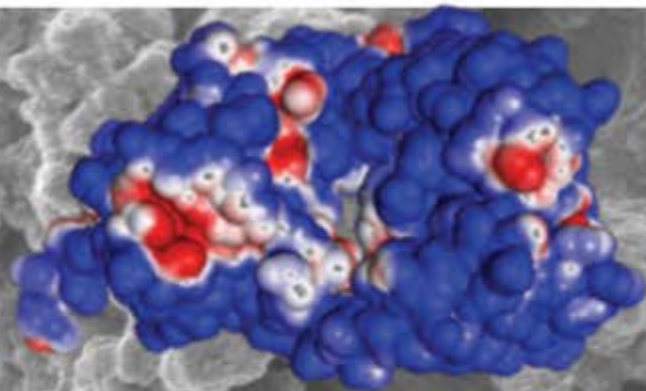
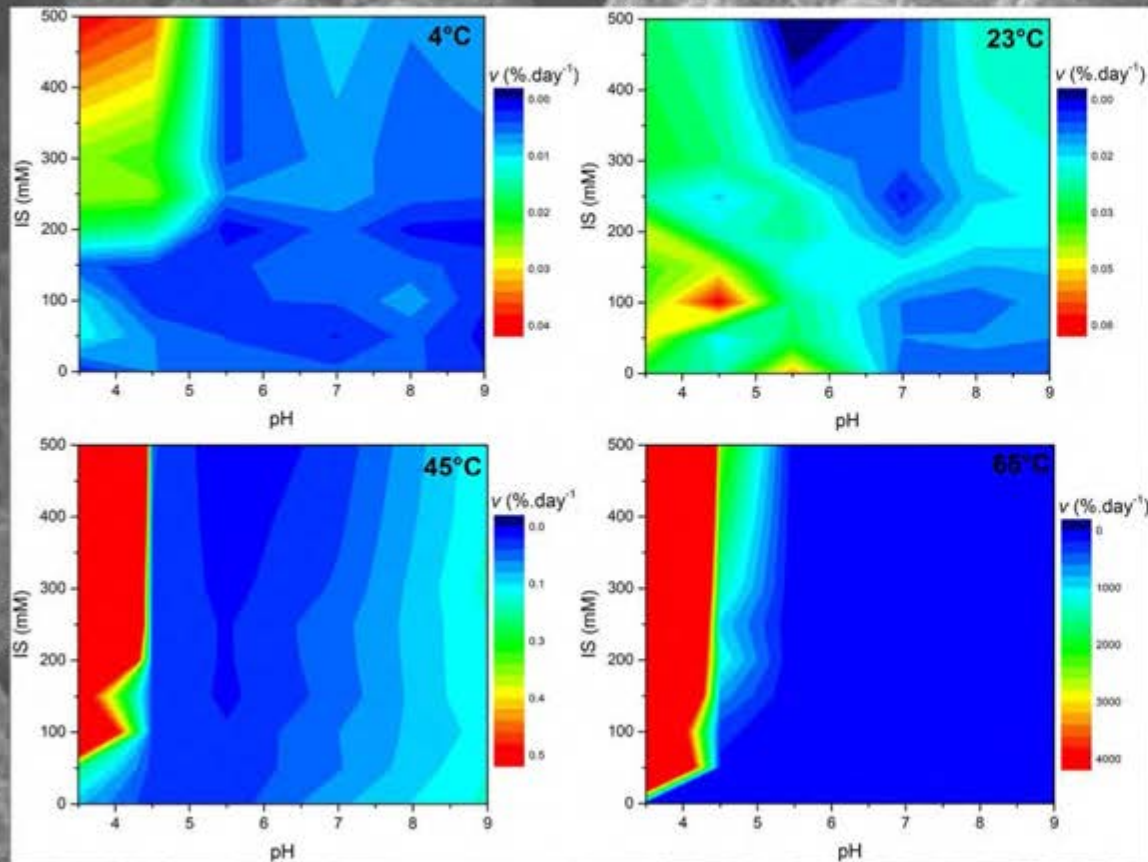
Aim3: How can we engineer based on predictions?

Can we engineer lower aggregation rates?

Can we develop novel (GRAS-based) excipients?



- O1. Use high-throughput automation to generate a large experimental formulation dataset for protein:excipient combinations, that will include aggregation kinetics, conformational stability, colloidal stability, phase behaviour, and rheology measurements.
 - O2. Molecular informatics and modelling will improve predictability of formulation attributes and excipient effects
 - O3. Analytical advances will enable earlier, more sensitive, and lower-volume assessments of formulated protein degradation kinetics.
- 



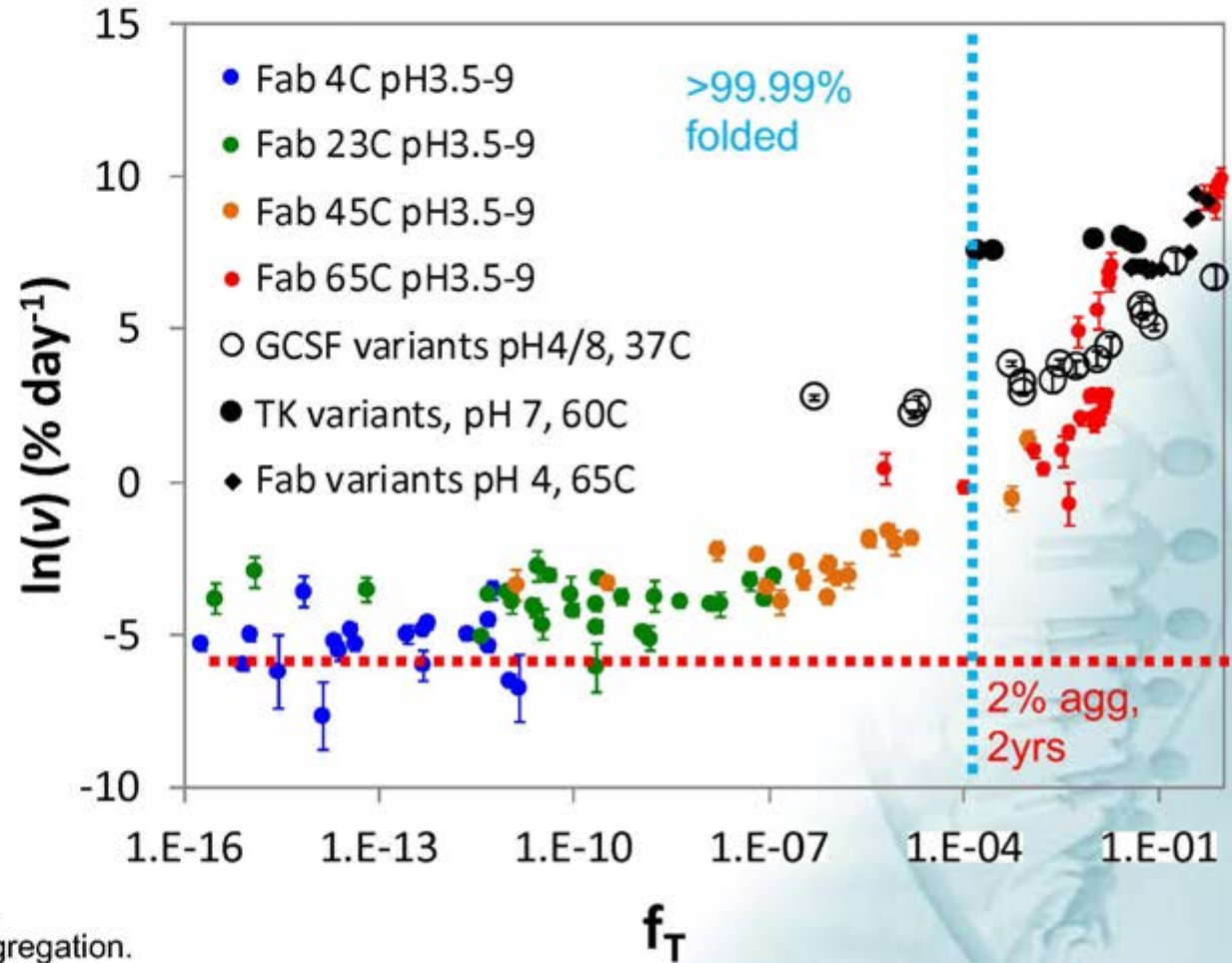
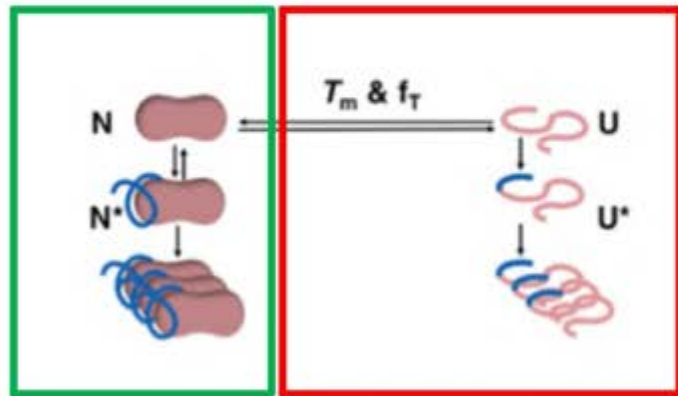
- Kinetics of native monomer loss determined for >1 year
- Range of pH, incubation T, and ionic strength

Kinetics at low T_{inc} often dont correlate with melting temperature

Where $T_{inc} \ll T_m$, fraction unfolded is $\ll 0.0001$:

Global unfolding (and hence T_m) is not relevant

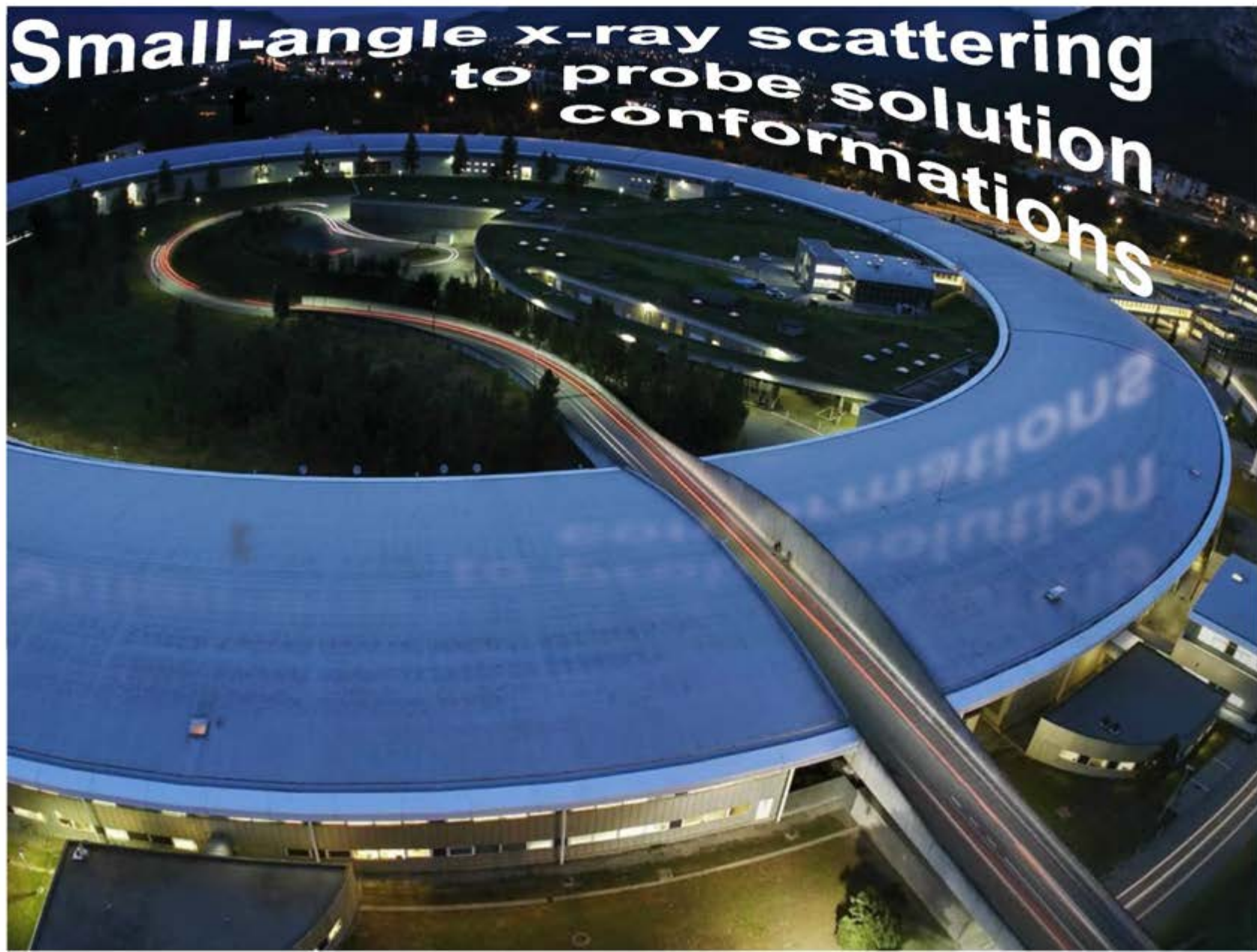
Native ensemble dynamics & colloidal stability control aggregation kinetics.

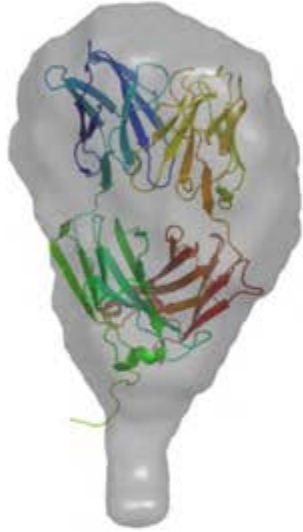


Zhang et al. (2018) Molecular Pharmaceutics. 15, 3079-3092
 Robinson et al. (2018) Molecular Pharmaceutics. 15, 256-267.
 Chakroun et al. (2016) Molecular Pharmaceutics. 13, 307-319.
 See also Roberts (2013) – review on non-Arrhenius protein aggregation.

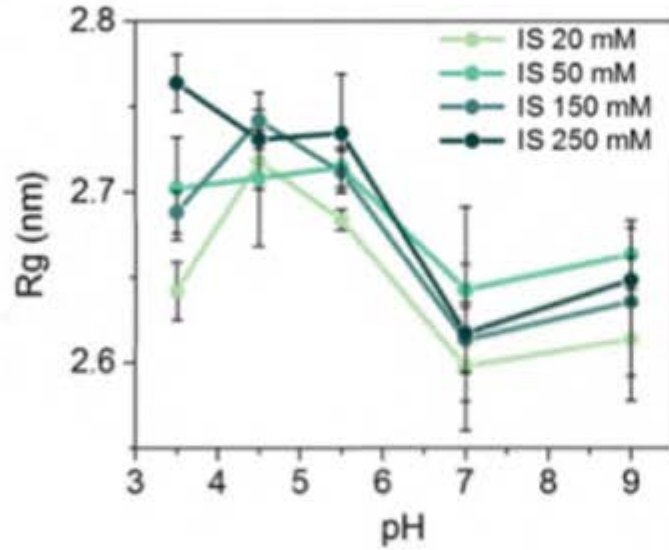


Small-angle x-ray scattering to probe solution conformations

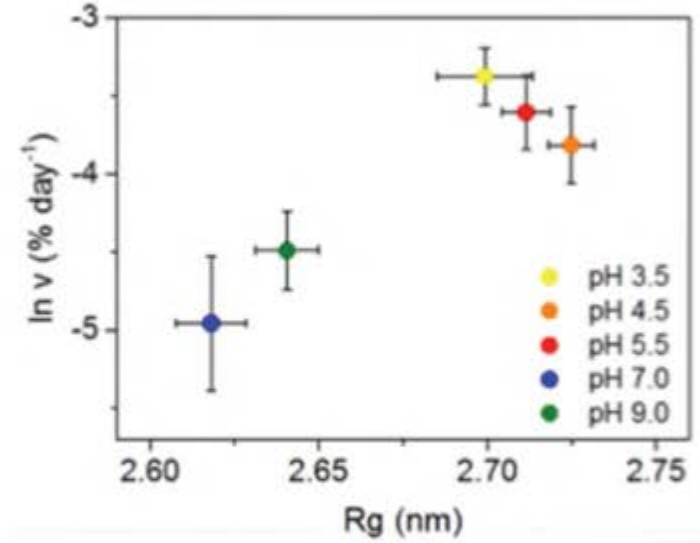




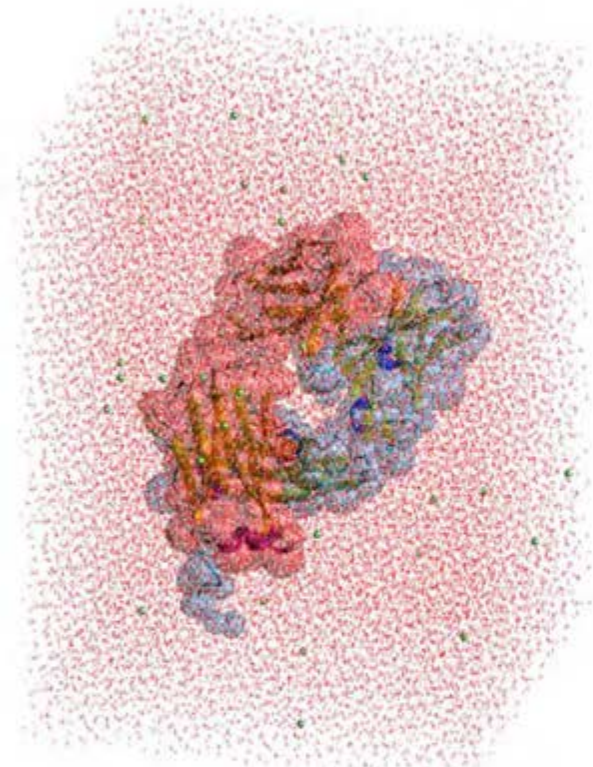
pH-dependent conformation



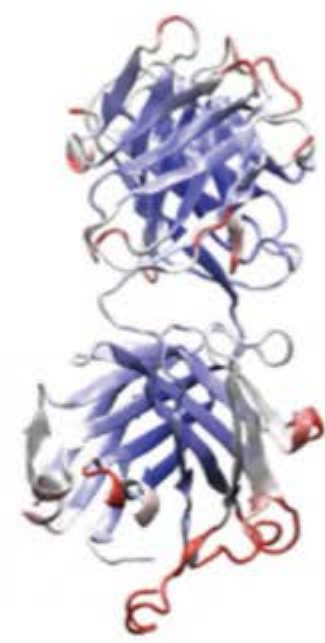
Aggregation kinetics correlate



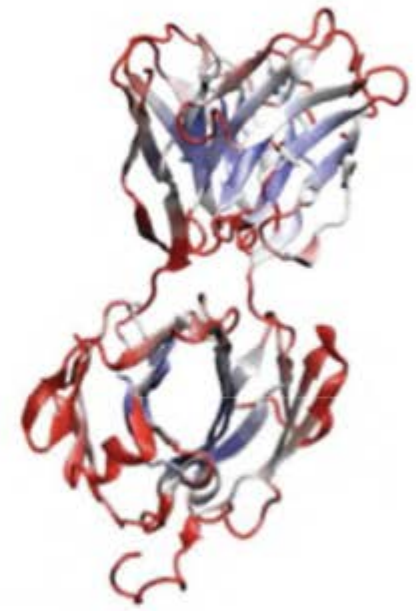
- Conformational change with pH correlates with aggregation kinetics, at 23 °C



Equilibrium RMSF (300K)
- pH7, 25°C, 50ns, 50mM IS
- pH3.5, 25°C, 50ns, 50 mM IS
- OPLS-AA/L force field & SPC/E water
- Triplicated



pH 7

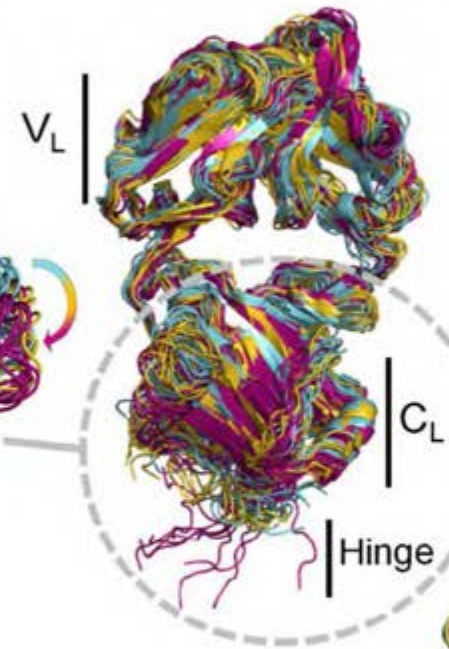
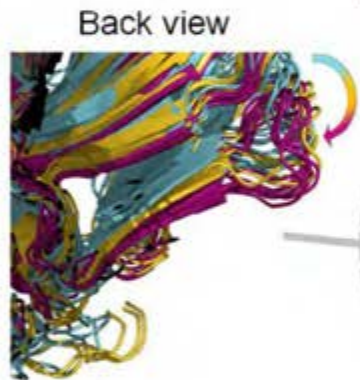


pH 3.5

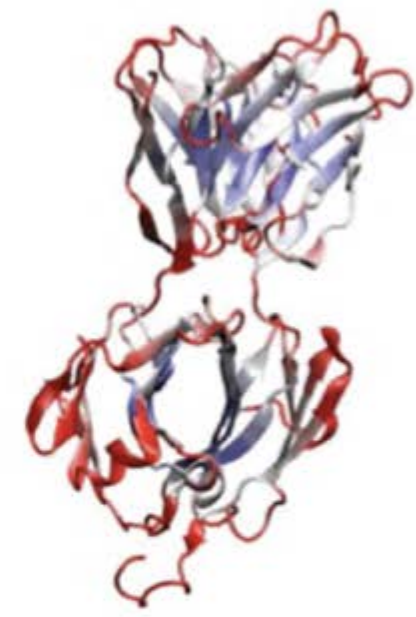
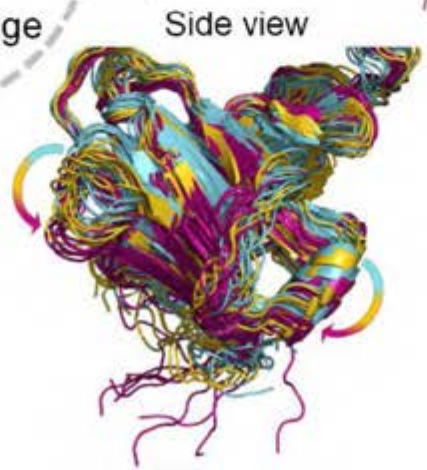
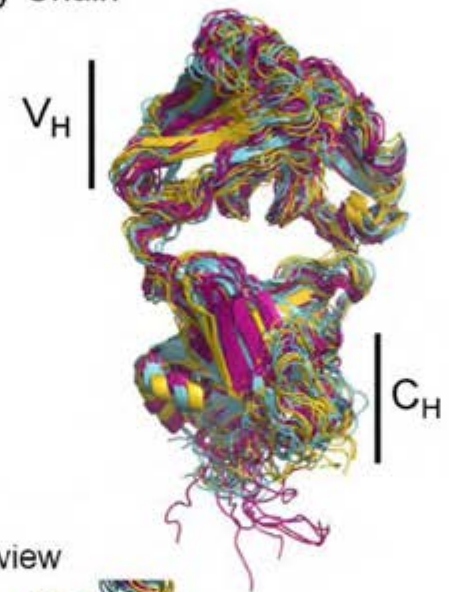
CL domain displacement

A Light Chain

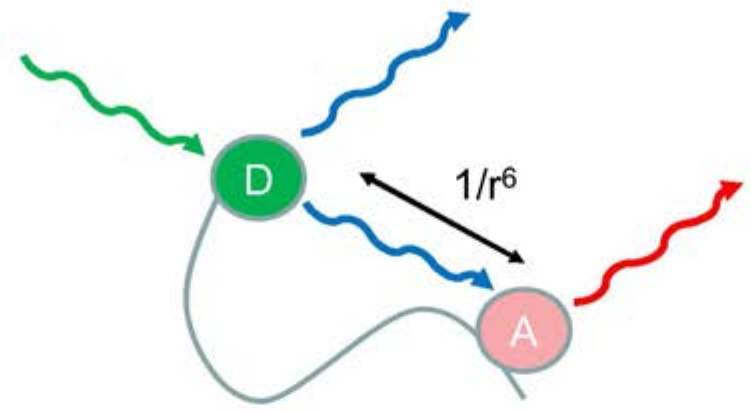
- pH 7.0
- pH 5.5
- pH 3.5



B Heavy Chain



CL domain displacement

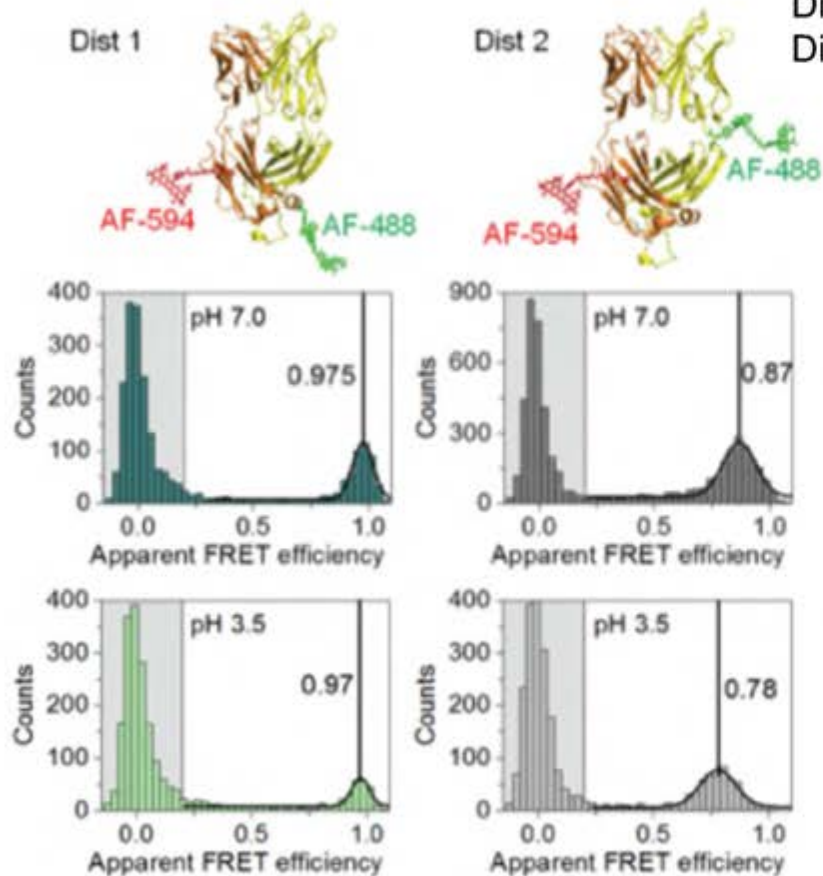


Single-molecule fluorescence

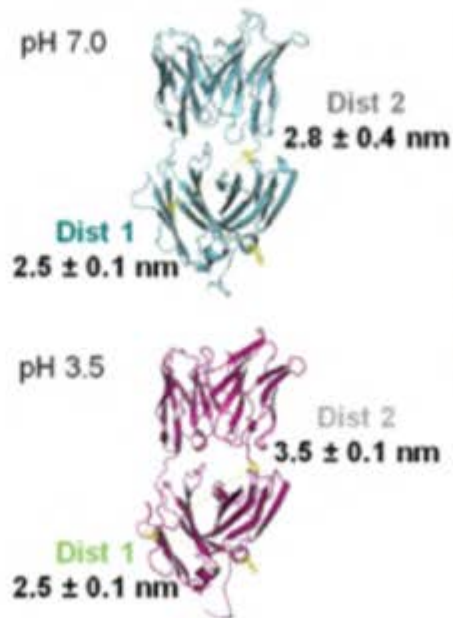


smFRET analysis of pH-dependent Fab conformations

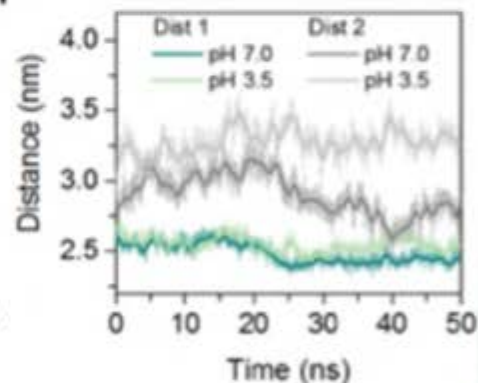
A smFRET



B SAXS

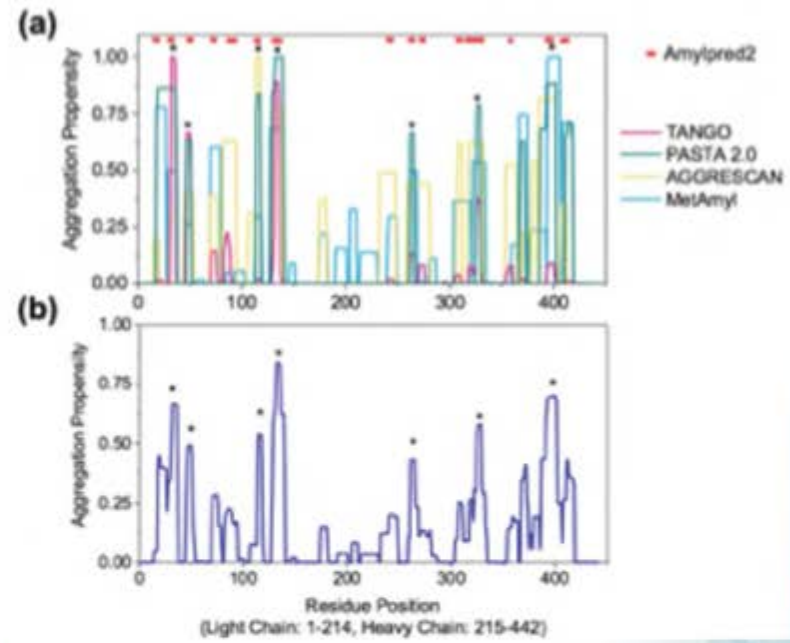
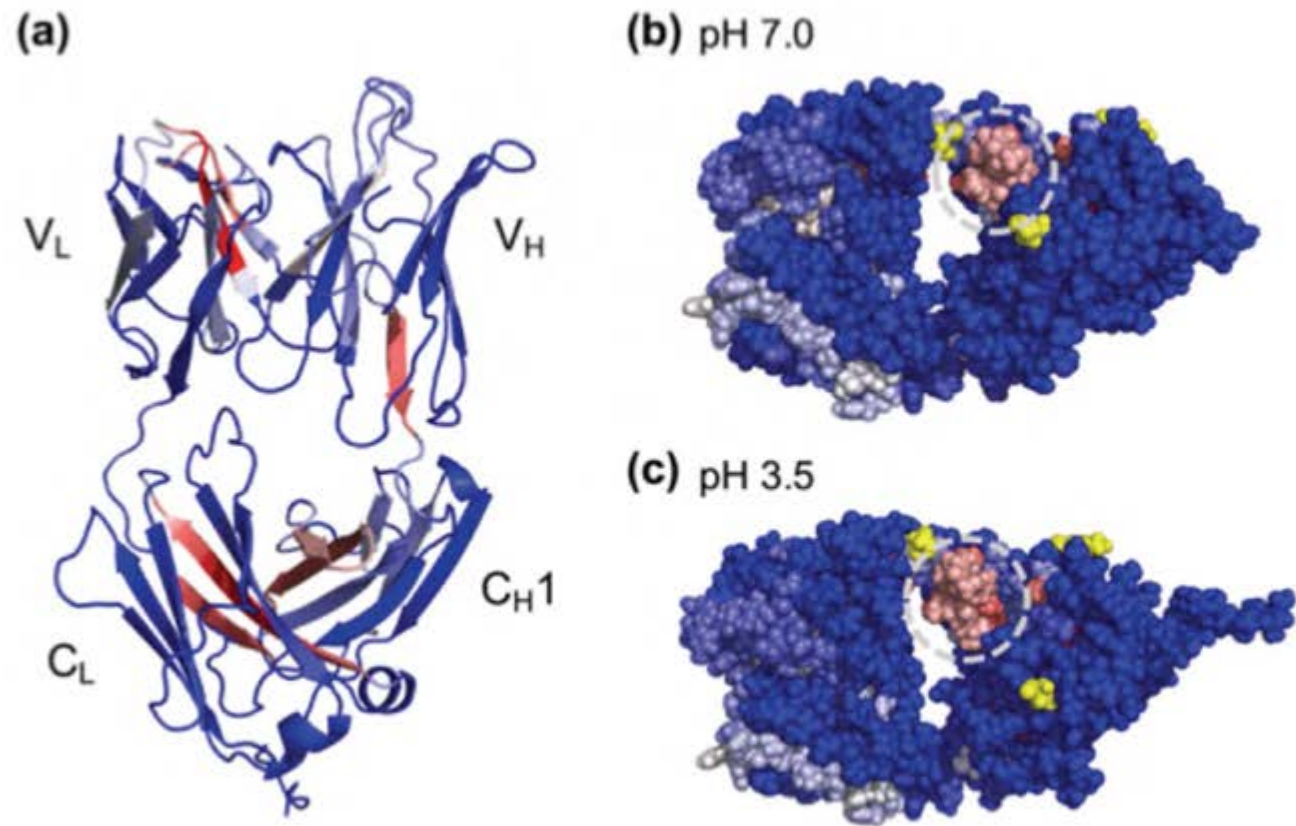


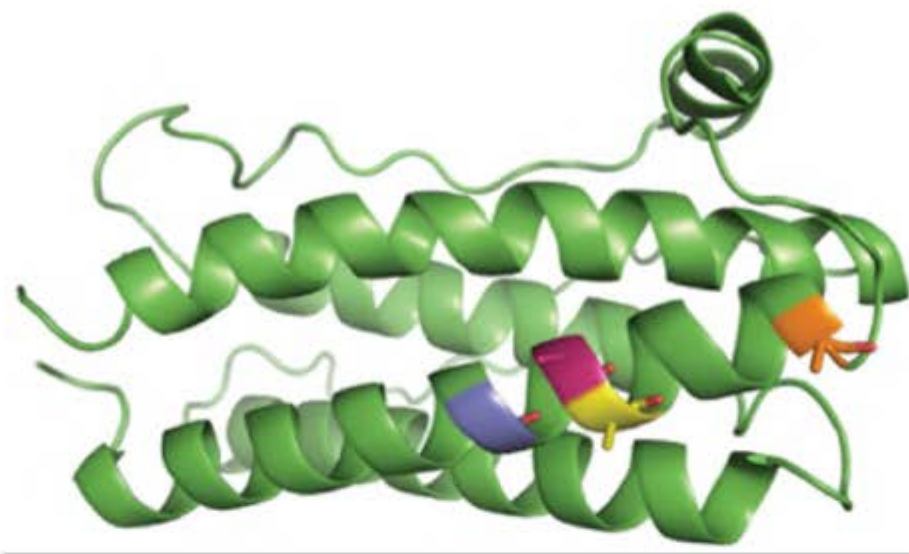
C MD simulations



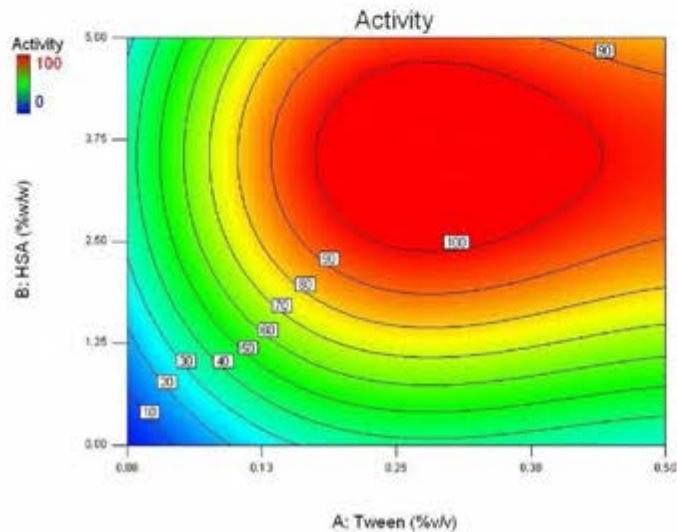
smFRET, SAXS and MD reveal the same dynamics and conformational shift with pH

Best-fit SAXS structures reveal APR exposure at low pH



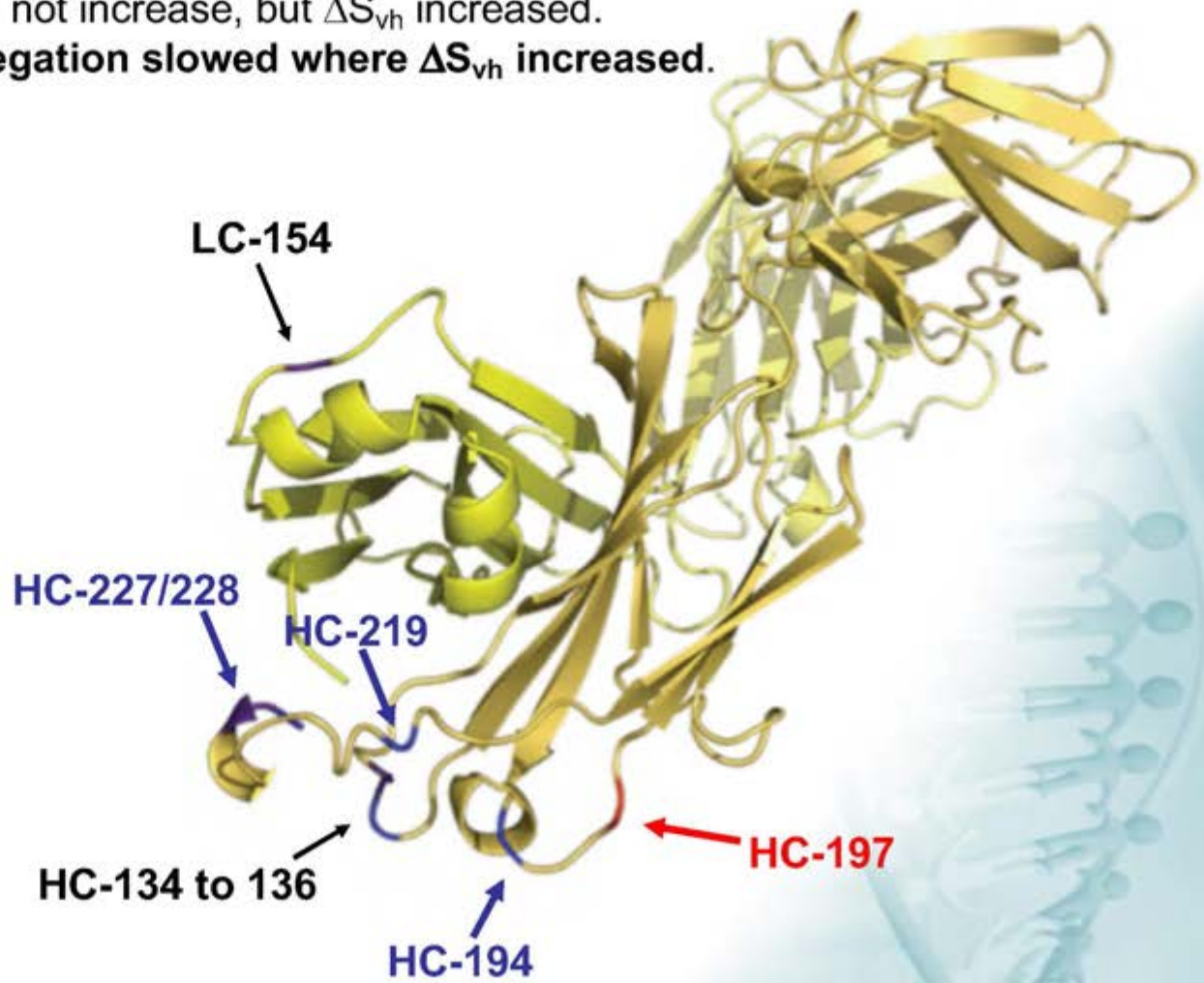
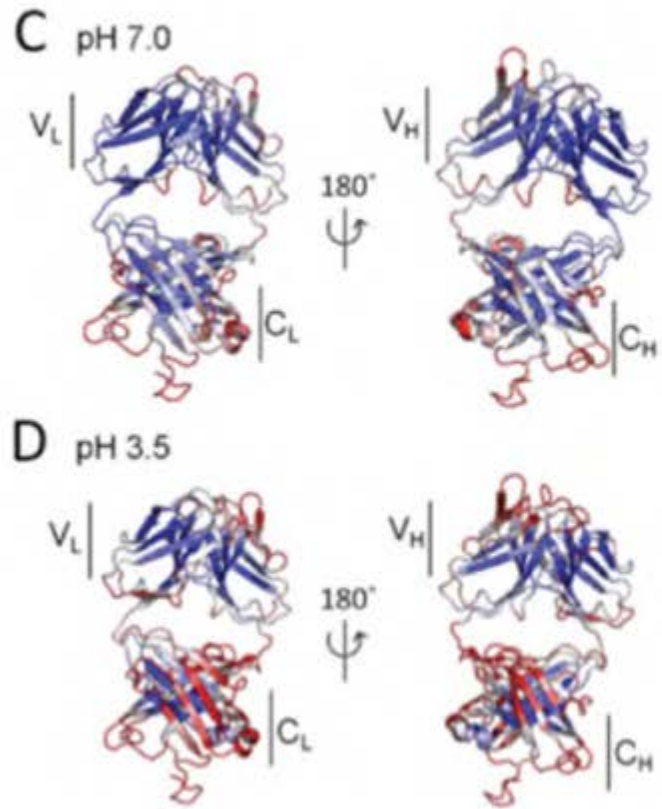


Protein engineering and formulation



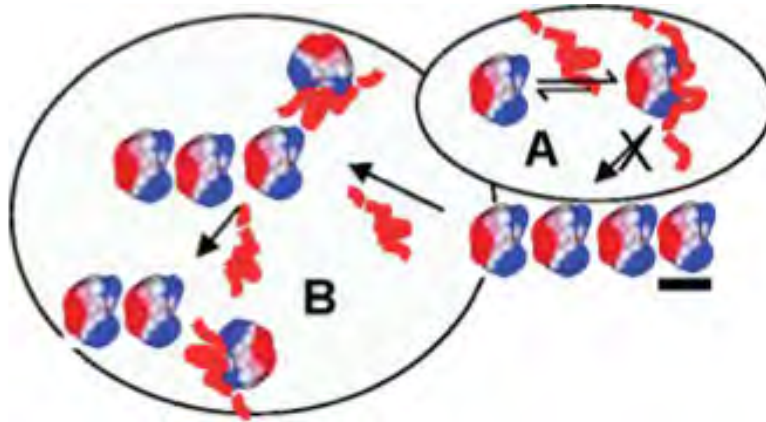
MD-guided protein engineering to slow aggregation

T_m did not increase, but ΔS_{vh} increased.
Aggregation slowed where ΔS_{vh} increased.



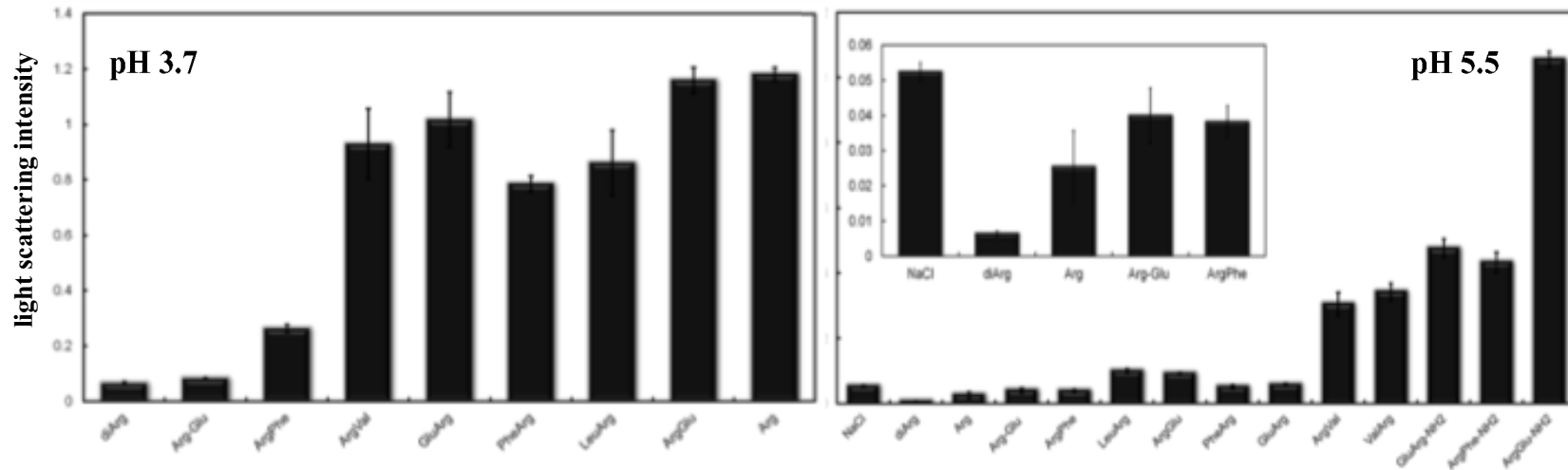
Designed 12 stabilising and 5 destabilising mutations using ROSETTA at flexible sites.

Dipeptides as novel excipients



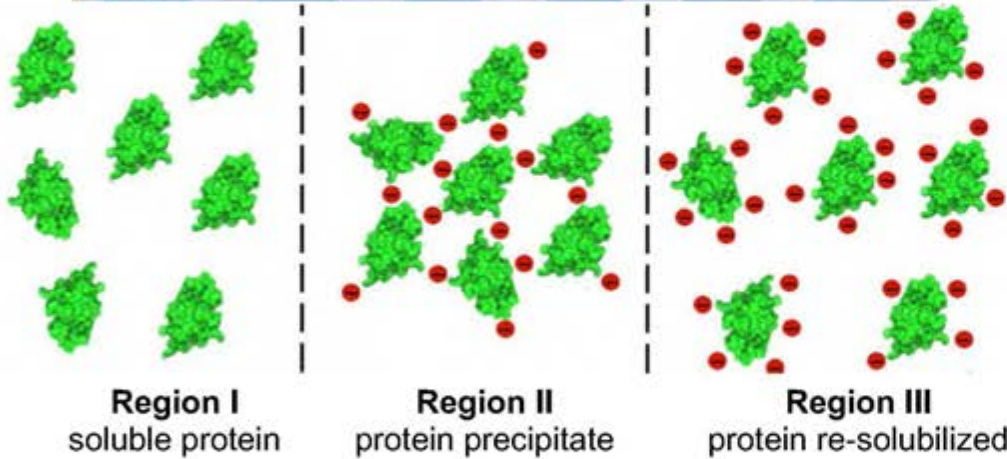
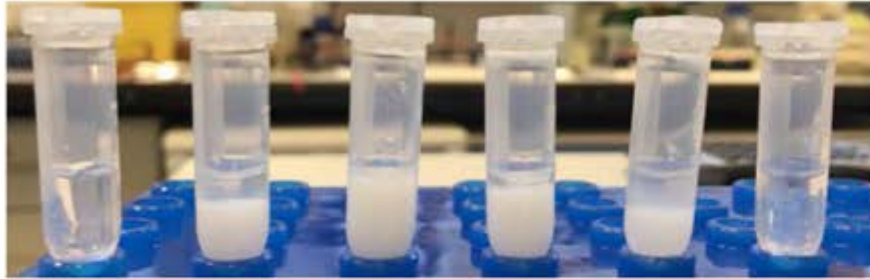
• At isoelectric pH, diArg is most effective at reducing insulin self association versus all other additives reflecting ability to neutralize electrostatic attraction.

• At pH 3.7, diArg, ArgPhe and mixtures of Arg and Glu equally effective at neutralizing hydrophobic interactions between insulin molecules

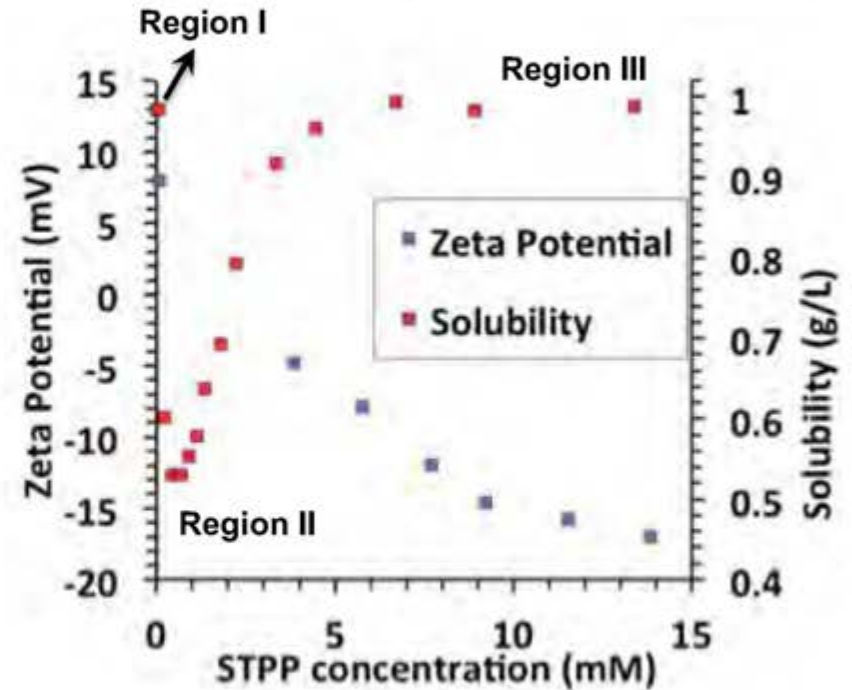
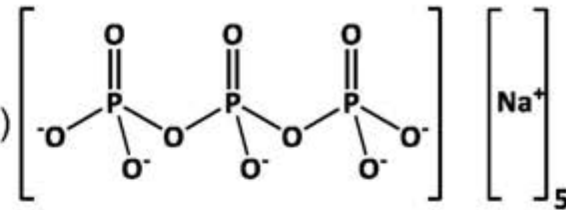


STPP for tuning protein phase behaviour

Increasing STPP concentration

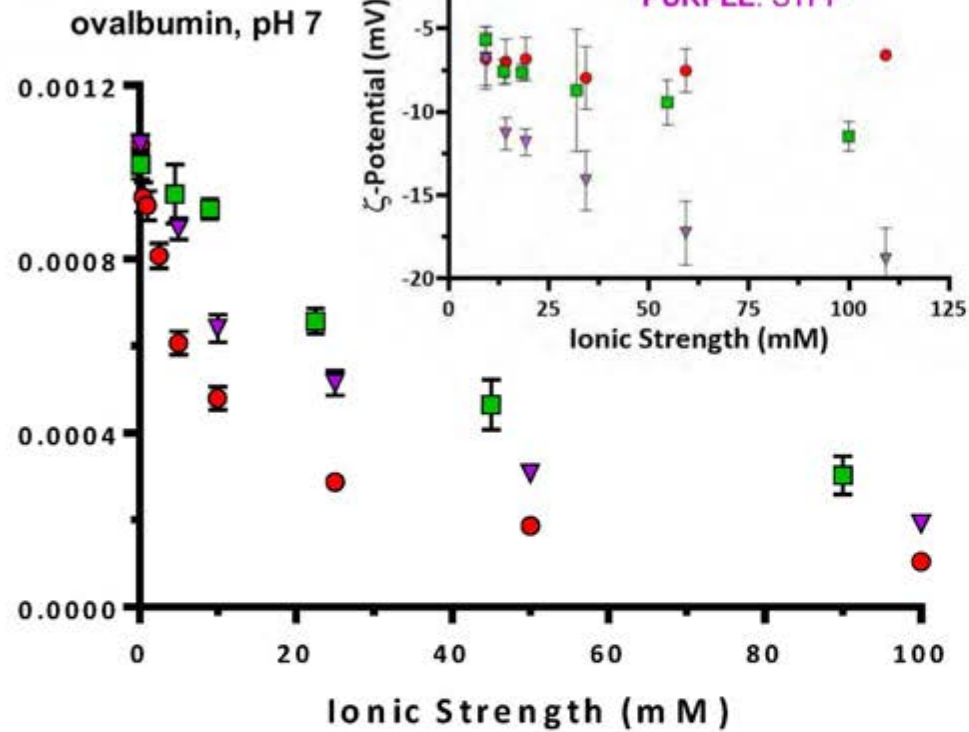


sodium triphosphate (STPP)



Lysozyme, pH 8

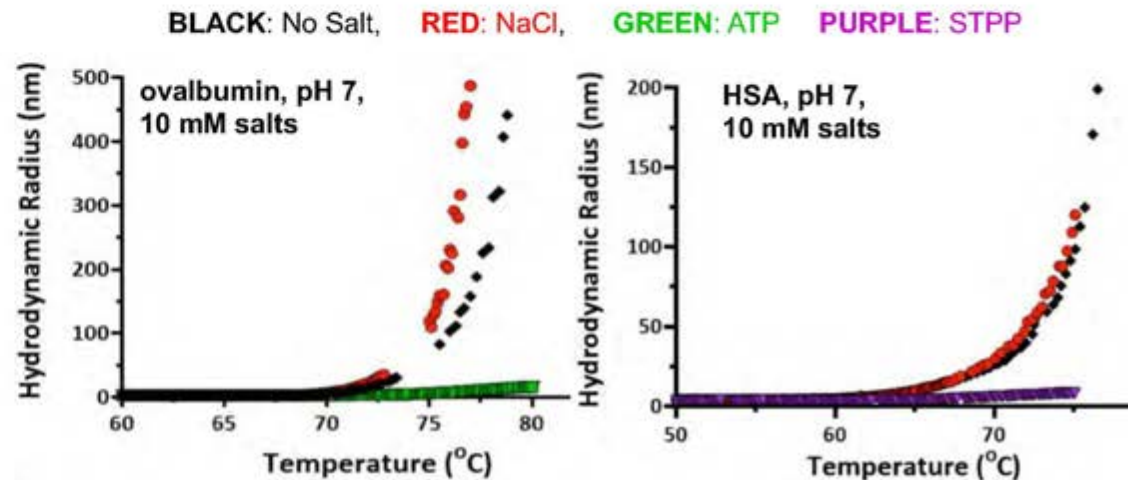
- Reentrant condensation phenomena observed with positively charged proteins and STPP (or ATP)
- Initial [STPP] causes protein precipitation through forming ion bridges across proteins
- Resolubilization at higher [STPP] due to overcharging protein
- STPP can direct formation of reversible colloidal gels with glassy dynamics for formulating proteins



- ATP and TPP prevent thermal-induced aggregation of negatively charged proteins.
- Stabilization is due to reduction in aggregate growth rates through electrostatic stabilization

ATP and STPP prevent protein aggregate growth

- Multivalent anions ATP and TPP supercharge negatively charged proteins through ion binding
- Supercharging protein with ATP or TPP increases protein-protein repulsion and colloidal stability as reflected by increase in B_{22} values at fixed IS

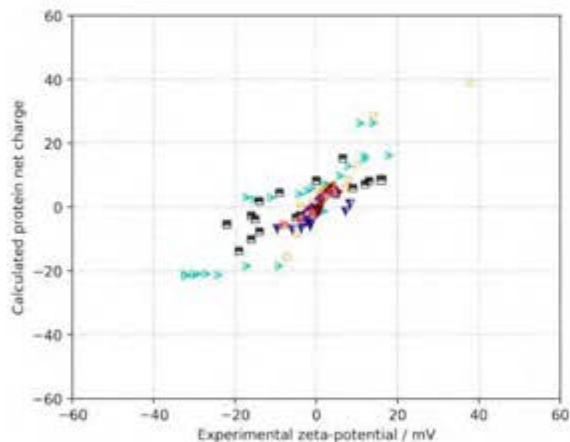


Polyvalent anion binding and solubility

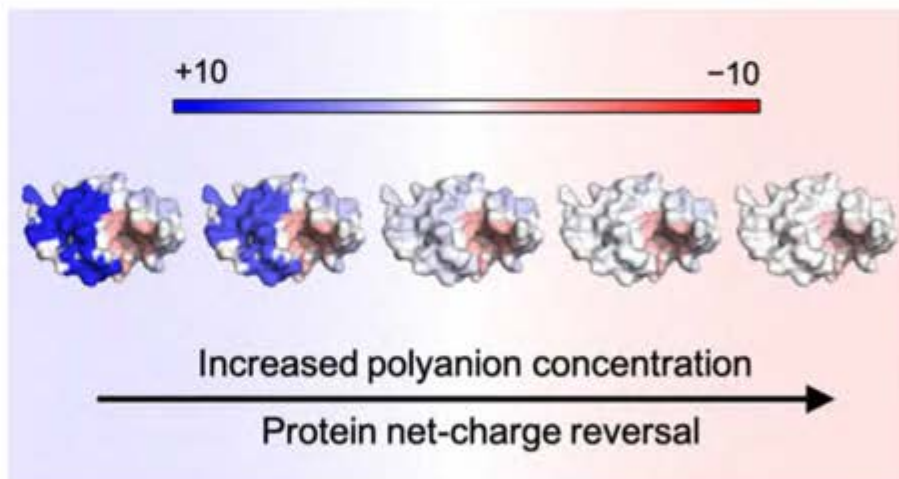
A model (not a web tool), following on from Bye & Curtis:

Controlling Phase Separation of Lysozyme with Polyvalent Anions

Jordan W. Bye[✉] and Robin A. Curtis*

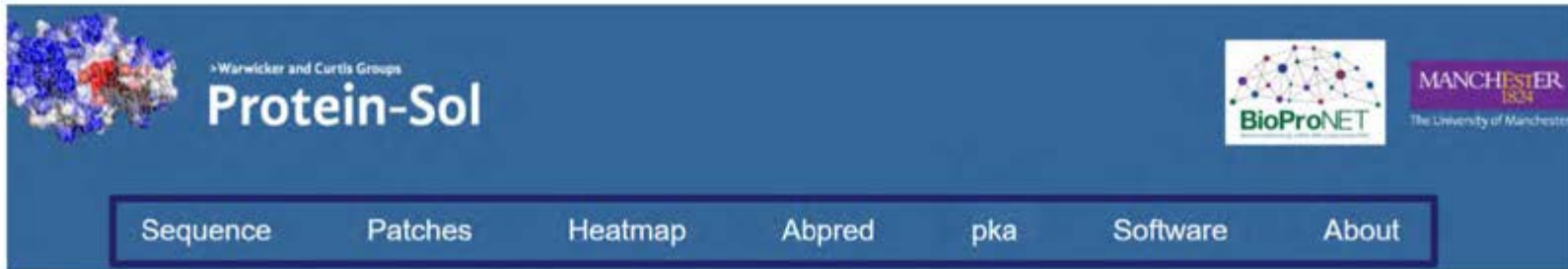


model for anion
binding fits
measured zeta potl



model applied to human proteins
suggests that over-charging and
potential solubilisation is general

Solubility prediction server



>Warwick and Curtis Groups
Protein-Sol

BioProNET
The University of Manchester

Sequence Patches Heatmap Abpred pKa Software About

Sequence Prediction

The protein-sol software will take a single amino acid sequence and return the result of a set of solubility prediction calculations, compared to a solubility database.

Please enter a single sequence of single letter amino acid codes in the FASTA format.

For example

```
> P00547
```

```
MVKVYAPASSANMSVGFVDVLGAAVTPVDGALLGDVVTVVEAAETFSLNNLGRFADKLPSEPRENIVYQCWERFCQELGKQI  
PVAMTLEKNMPIGSGLGSSACSVVAALMAMNEHCGKPLNDTRLLALMGELEGRISGSIHYDNVAPCFLGGMQLMIEENDI  
TSCQVPGEDFWLWVLA YPCTKVSTAFARA TLPAQVRRDDCTAHGRHTAGETHACYSDOPELAAKLMKDVTAEFVPERITLP
```

Originally (2017): Sequence-based solubility prediction, based *E. coli* data
Currently (2020), also: Patches, Heatmap, Abpred, pKa
In development (2020): Excipient predictions (for next meeting)

Patches: with Fab fragment

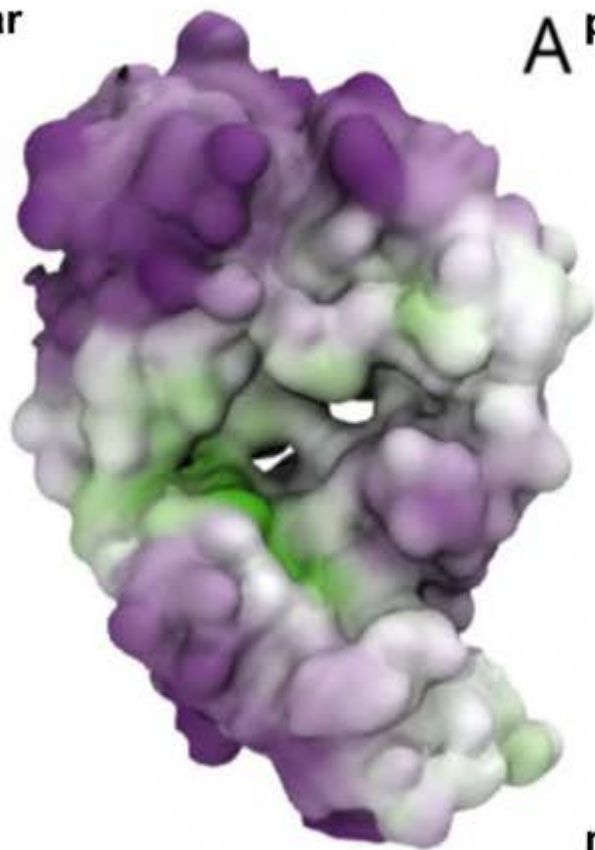
non-polar

2.3

1.45

0.6

polar
polarity



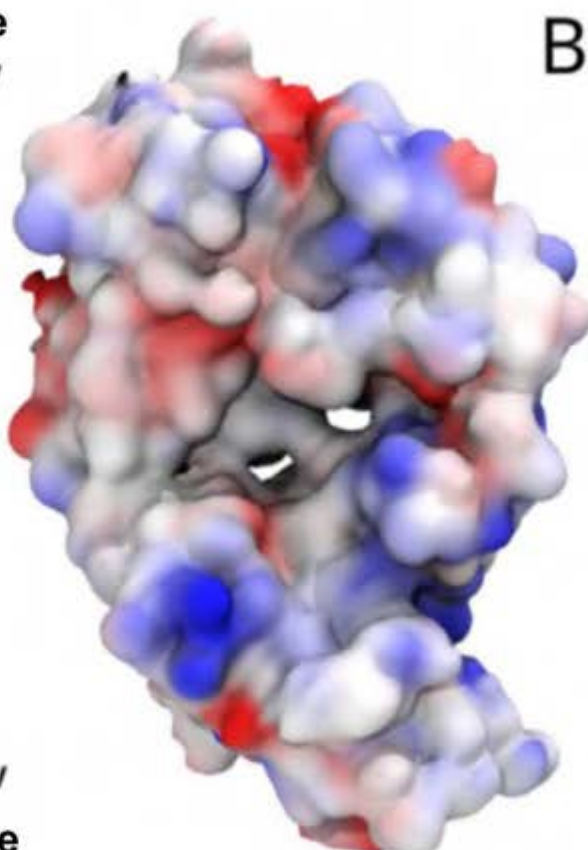
A positive

75mV

0mV

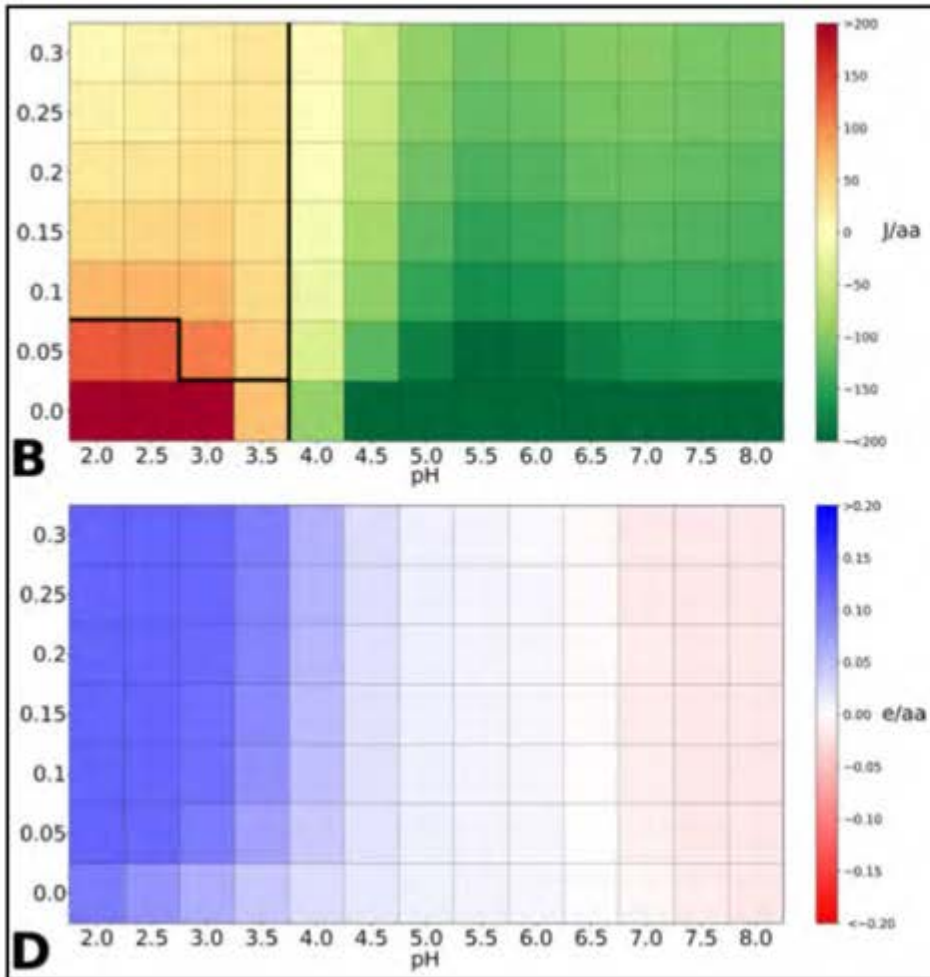
-75mV

negative
charge



B

Heatmap: pH and ionic strength

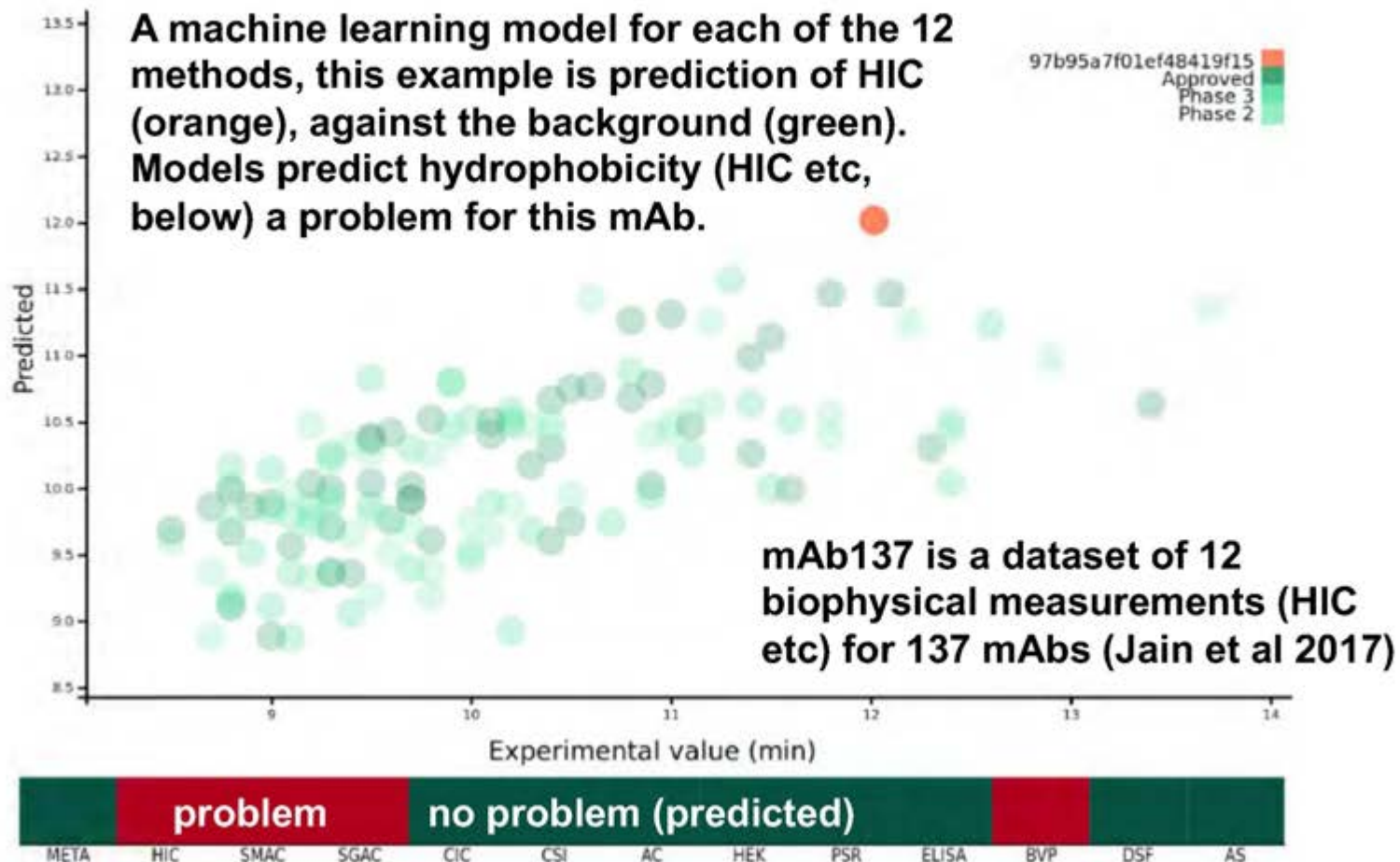


Predicted pH and ionic strength dependence of folded state stability (upper panel), and net charge (lower panel).

Upper panel shows 'phase diagram' fit to experimental data.

Method can be used to identify regions that lower stability in conditions such as a low pH of bioprocessing. These regions can then be engineered out, e.g. switching Asp/ Glu for Asn/Gln.

AbPred: ML models from mAb137





Take-home messages

Aim1: Understand factors affecting aggregation in formulation

- local dynamics/unfolding
- exposure of aggregation hotspots (APRs)
- colloidal stability (net charge)
- Optimize formulations to increase T_m , decrease dynamics and increase net charge

Aim2: How can we predict better formulations?

- Local dynamics and aggregation hotspots can be predicted computationally
- Excipient interactions can be predicted by molecular docking or LCMS
- Net charge and effect of charged excipient binding can be predicted computationally
- see webserver at protein-sol.manchester.ac.uk

Aim3: How can we engineer based on predictions?

- Mutations that suppress dynamics sometimes decrease aggregation kinetics
- Novel (GRAS-based) excipients based on dipeptides, and supercharging with STPP



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Cheng Zhang
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Mathew Robinson
Victoria Wood
Akash Pandya
Nuria Codina
Maariyah Samad

where are they now?

- Biopharm Services
- Lonza
- Merus (NL)
- IPSEN
- **PDRA UCL**
- GSK
- Lonza
- GSK
- **PhD UCL**
- **NeoLeukin (Seattle, USA)**
- O4 Research

Collaborators:

SAXS & AUC
GCSF
Mass Spec
Fabs
IgGs
Molecular docking

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- Jordan Bye – University of Manchester
- Max Hebditch – University of Manchester

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- Akash Pandya - UCL
- Nikita Vekaria – University of Manchester
- Jas Kalayan – University of Manchester

EPSRC EP/I033270/1 : CIM
EP/L015218/1 : CDT
EP/K005030/1 : Equipment
EP/M028100/1 : Equipment
EP/N025105/1 : Formulation
EP/P006485/1 : FTHM Hub

BRIC PhD: BB/J003824/1
BRIC: BB/I017119/1
BRIC: BB/K011162/1
BRIC: BB/E005942/1

