

Comparison of protein H exchange data with local fluctuations observed in perfectly funneled structure based models

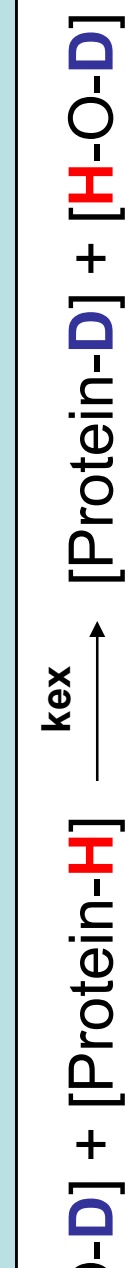
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ABSTRACT

The principle of minimal frustration assumes that the overall landscape of proteins has evolved to be a rough funnel. When energetic frustration is low enough, topology alone becomes the key factor governing the folding. The structures of transition state ensembles, folding intermediates, and the mechanisms of dimerization, and domain swapping have been well predicted by models where frustration has been removed and topological information about the native state is the sole input. In this work we compare the protein motions simulated using perfectly funneled structure-based models, and amide hydrogen exchange (HDX) measurements under native conditions that report on the thermodynamics and kinetics of protein structures.

1 HDX - local unfolding model



$$PF = \frac{k_{cl}}{k_{ex}} \quad k_{ex} = \frac{k_{op}}{k_{op} + k_{cl} + k_{cl}} \quad PF = 1 + \frac{k_{cl} \cdot k_{cl}}{k_{op}}$$

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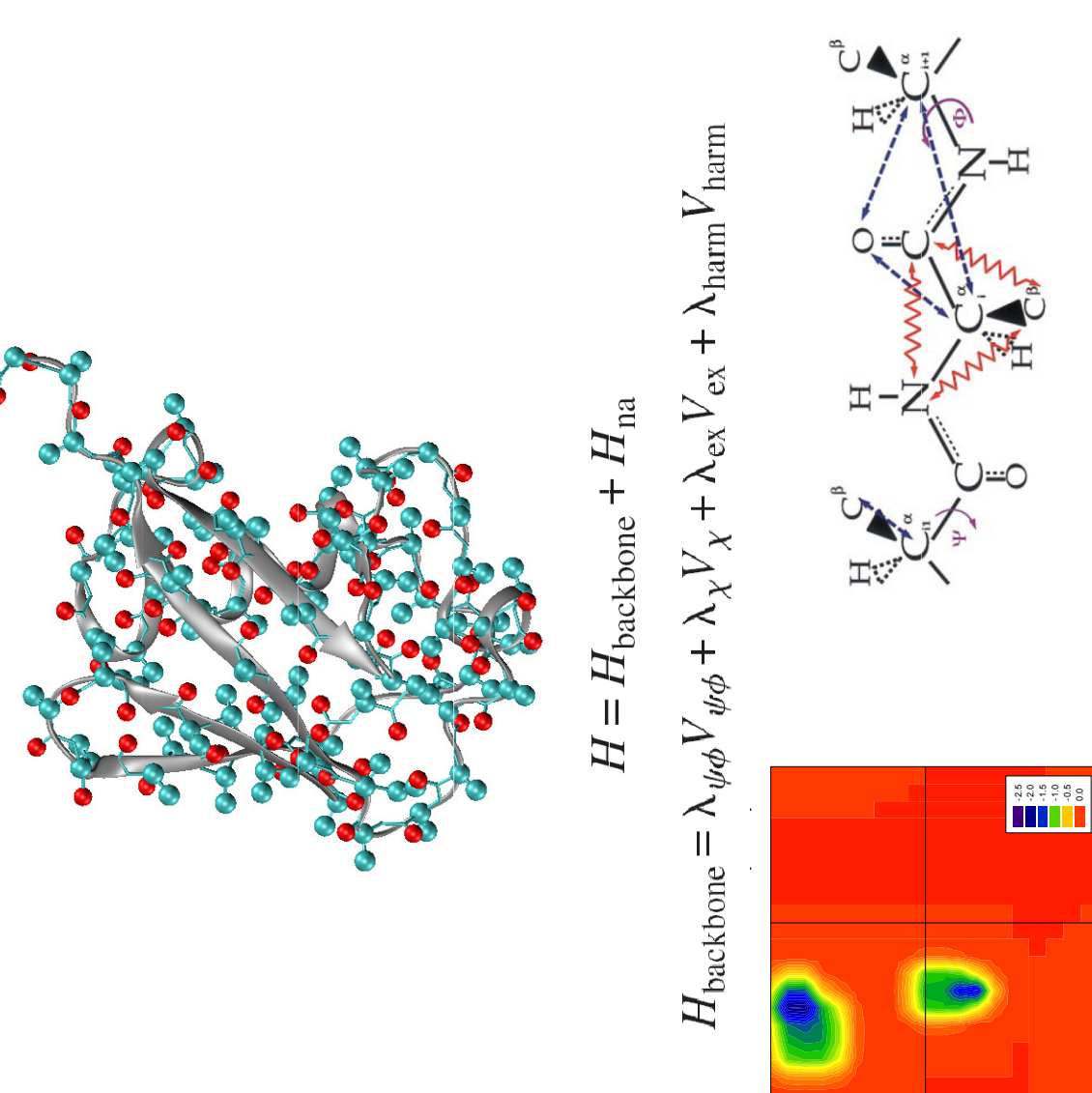
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2 Structure based model

Coarse grain model: only C α , C β , and O



Attractive forces only between residues that make contact in the native structure

Homogeneous contact potential

Non additivity: forces depend on the structural context. The weight of a pair interaction is modulated by the presence of other contacts (P dependent).

$$H_{ij} = -\sum_k \frac{1}{r_{ijk}} \left(\frac{r_{ijk}}{r_{ijk}^0} \right)^{12} \left(\frac{r_{ijk}}{r_{ijk}^0} \right)^6$$

$$E_i = \sum_j \epsilon_{ij}(r_{ij}) = -\sum_j \left(\frac{1}{r_{ij}} \right) \left(\frac{r_{ij}}{r_{ij}^0} \right)^{12} \left(\frac{r_{ij}}{r_{ij}^0} \right)^6$$

$$a = \frac{1}{8N} \left(\sum_i \sum_j \gamma_i \gamma_j (r_{ij} - r_{ij}^0)^2 \right)$$

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3 Method

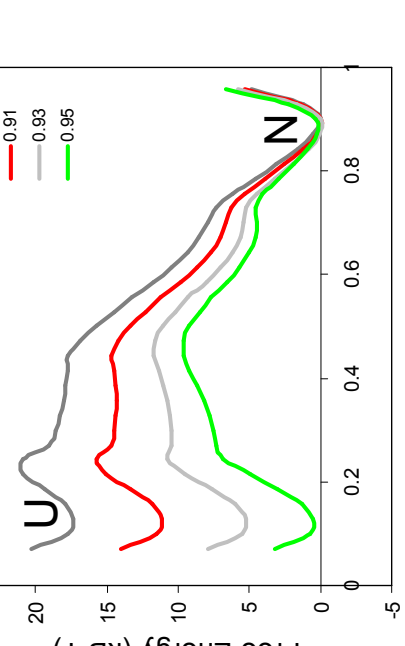
MD simulation - Umbrella sampling Q_w bias

$$Q_w = \frac{2}{(N-1)(N-2)} \sum_{i,j=1}^{N-1} \exp \left(-\frac{(Q_i - Q_j)^2}{2\sigma_i^2} \right)$$

Histogram analysis (WHAM) – Probability of sub states

$$F_{\text{WHAM}} = -k_B T \ln P_{\text{WHAM}}$$

Temperature Screening



Calculate the probability of the open and closed conformations for each residue from evaluation of local order parameters

Prediction of $\ln Pf$ assuming EX2 conditions

$$\ln Pf_{\text{pred}} = \ln \left(1 + \frac{k_{cl}}{k_{op}} \right) = \ln \left(1 + \frac{P_c}{P_o} \right)$$

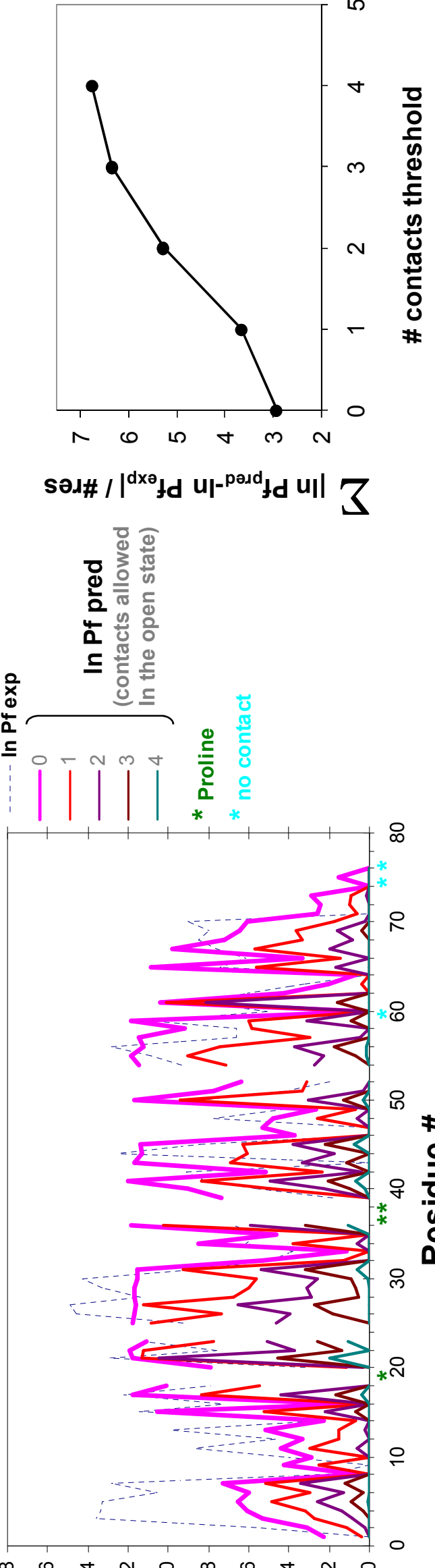
4 Local order parameters used for the structural definition of the open and closed conformations

Accessibility

The accessibility was evaluated by the number of contacts (Qi) of each residue

$$Q_i = \sum_j q_{ij} \quad q_{ij} = 0.5(1 + \tanh(5(r_{ij} - r_{\text{cutoff}}))) \quad r_{\text{cutoff}} = 6.5 \text{ \AA}$$

Residues with a number of contacts below a certain threshold were considered to be in the open state



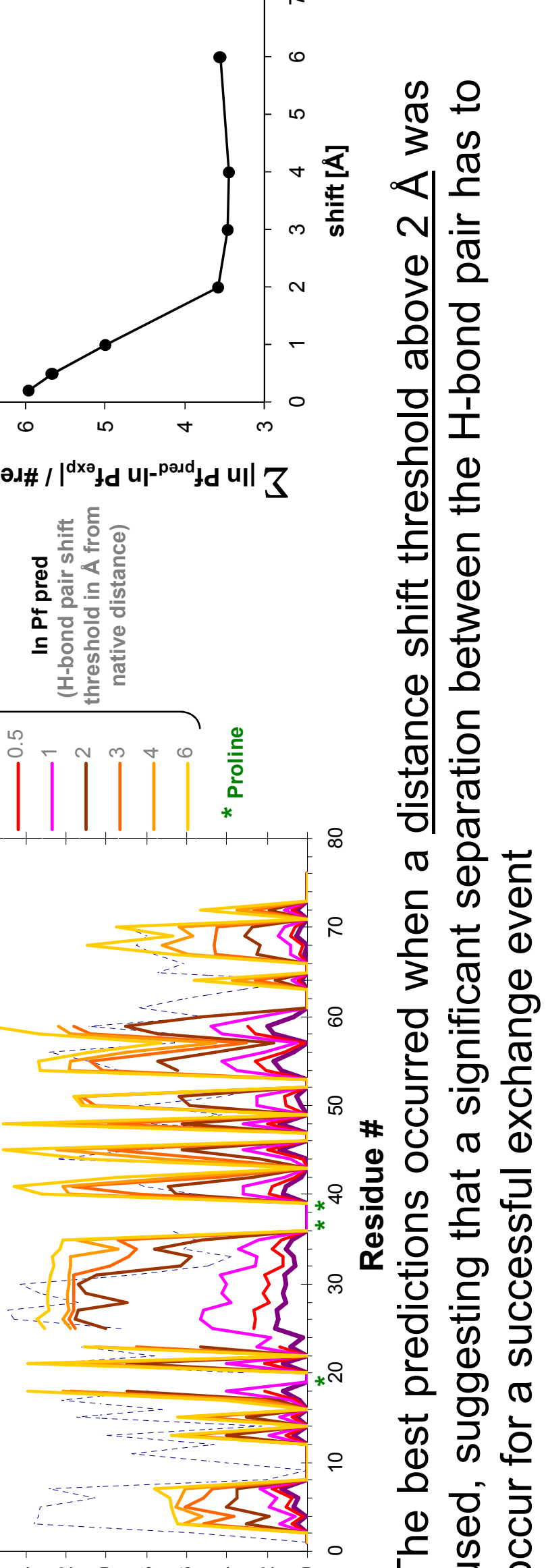
The best predictions occurred when 0 or 1 contact were allowed in the open state. This suggests that an extensive local destabilization is needed for an exchange event to occur.

H-bonding

Residues H-bonded in the native structure were considered to preserve their H-bonding state along the simulations if their pairwise distance was below a certain threshold. In this case they were considered to be in the closed state.

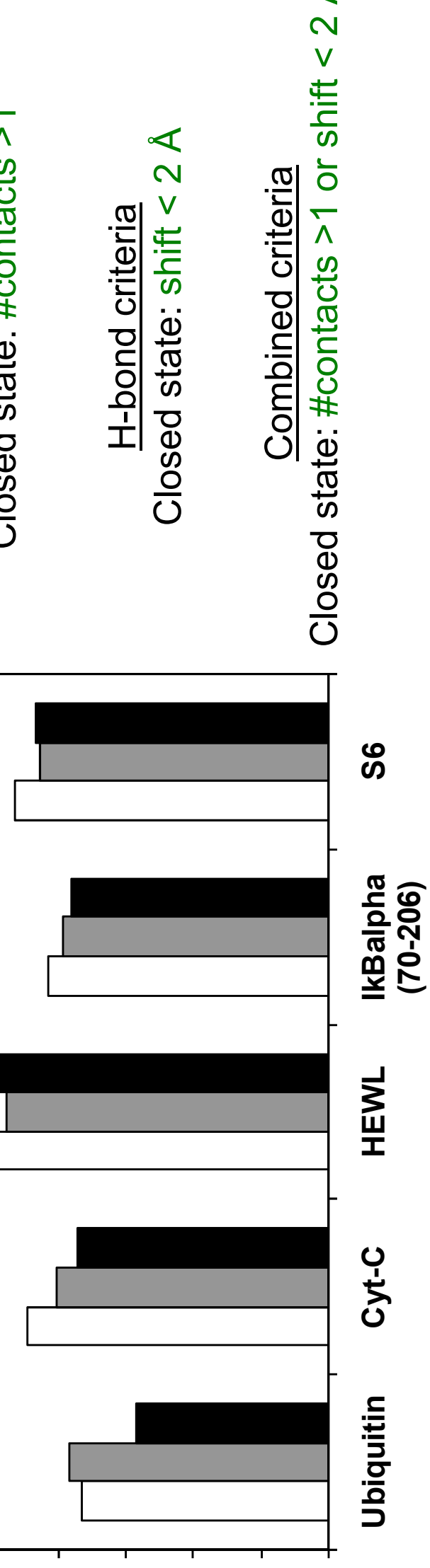
$$\text{threshold} = r_{ijN} + \text{shift}$$

r_{ijN} = native distance between residues i and j



The best predictions occurred when a distance shift threshold above 2 Å was used, suggesting that a significant separation between the H-bond pair has to occur for a successful exchange event

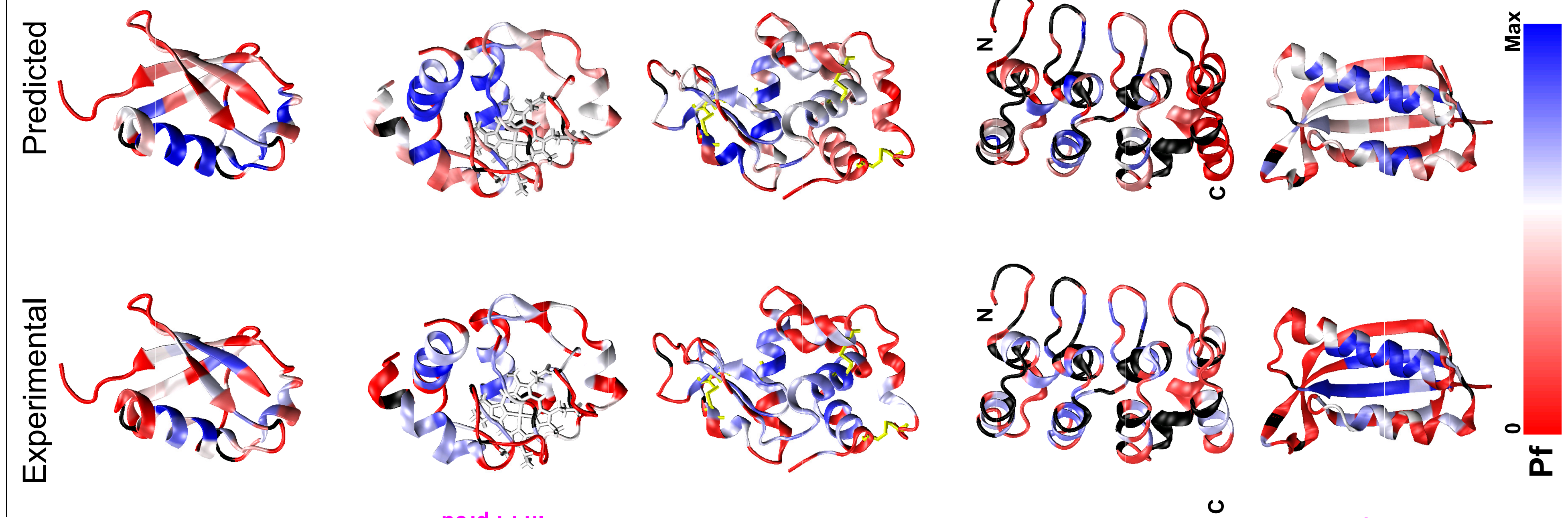
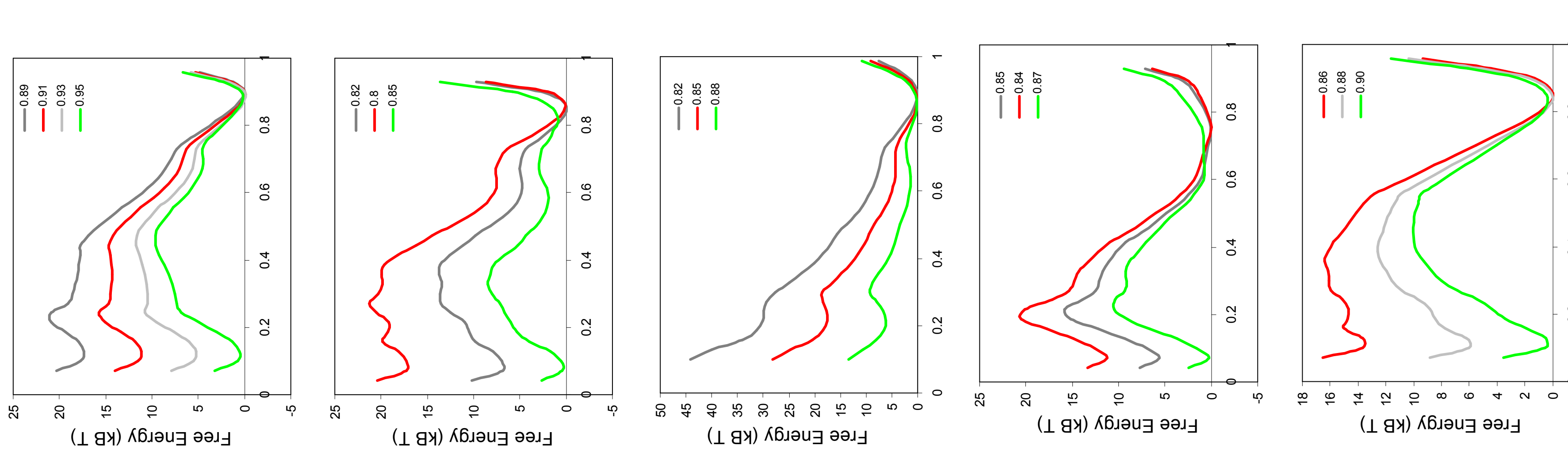
Most predictions improves when the accessibility and H-bonding criteria are combined



5

Topology is a dominant factor that determines HDX kinetics

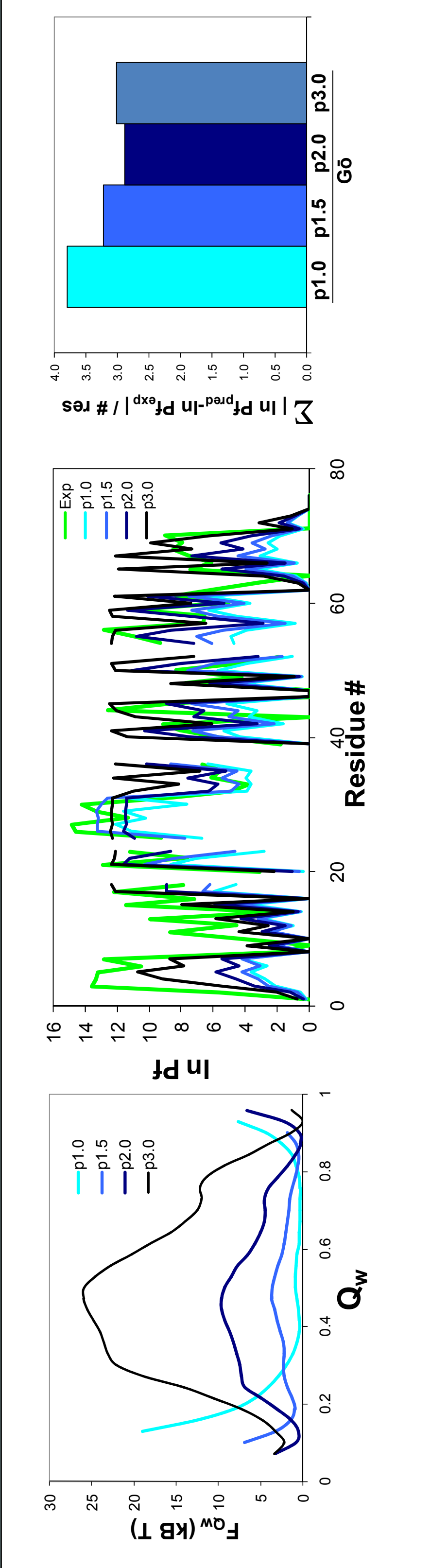
Red: Experimental temperature (matches the real global stability under the HDX conditions)
Green: Sampling temperature (T)



CONCLUSIONS

In this work we developed a method for the interpretation of HDX protection factors from perfectly funneled structure based model simulations. The agreement between the predictions and experiments suggests that topology has a dominant role in determining the exchange kinetics. The comparison of different criteria for structurally defining the open and closed states indicates that both accessibility and H-bonding properties should be taken into account in the analysis. The improvement in the predictions produced by the inclusion of non additive forces highlights the relevance of many body interactions in protein dynamics. The method developed will be useful for the interpretation of the exchange process, the parametrization of effective interactions between residues and fine tuning of the energetic terms in molecular dynamic simulations, the analysis of protein motions, and the characterization of the native and partially folded ensembles.

6 Non additive interactions improves HDX predictions



7 Dependence of HDX predictions with the temperature

