

Reporting Summary

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Statistics

For all statistical analyses, confirm that the following items are present in the figure legend, table legend, main text, or Methods section.

n/a Confirmed

- ☒ ☐ The exact sample size (n) for each experimental group/condition, given as a discrete number and unit of measurement
- ☒ ☐ A statement on whether measurements were taken from distinct samples or whether the same sample was measured repeatedly
- ☒ ☐ The statistical test(s) used AND whether they are one- or two-sided
Only common tests should be described solely by name; describe more complex techniques in the Methods section.
- ☒ ☐ A description of all covariates tested
- ☒ ☐ A description of any assumptions or corrections, such as tests of normality and adjustment for multiple comparisons
- ☒ ☐ A full description of the statistical parameters including central tendency (e.g. means) or other basic estimates (e.g. regression coefficient) AND variation (e.g. standard deviation) or associated estimates of uncertainty (e.g. confidence intervals)
- ☒ ☐ For null hypothesis testing, the test statistic (e.g. F , t , r) with confidence intervals, effect sizes, degrees of freedom and P value noted
Give P values as exact values whenever suitable.
- ☒ ☐ For Bayesian analysis, information on the choice of priors and Markov chain Monte Carlo settings
- ☒ ☐ For hierarchical and complex designs, identification of the appropriate level for tests and full reporting of outcomes
- ☒ ☐ Estimates of effect sizes (e.g. Cohen's d , Pearson's r), indicating how they were calculated

Our web collection on [statistics for biologists](#) contains articles on many of the points above.

Software and code

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Data collection We have provided a full code availability statement in the manuscript

Data analysis We have provided a full code availability statement in the manuscript

For manuscripts utilizing custom algorithms or software that are central to the research but not yet described in published literature, software must be made available to editors and reviewers. We strongly encourage code deposition in a community repository (e.g. GitHub). See the Nature Portfolio [guidelines for submitting code & software](#) for further information.

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All manuscripts must include a [data availability statement](#). This statement should provide the following information, where applicable:

- Accession codes, unique identifiers, or web links for publicly available datasets
- A description of any restrictions on data availability
- For clinical datasets or third party data, please ensure that the statement adheres to our [policy](#)

The implementation of the coarse-grained HPS model can be downloaded from <https://github.com/wzhenglab/ntail>. Other data supporting the findings of this study are available from the corresponding authors upon reasonable request.

Research involving human participants, their data, or biological material

Policy information about studies with [human participants or human data](#). See also policy information about [sex, gender \(identity/presentation\), and sexual orientation](#) and [race, ethnicity and racism](#).

Reporting on sex and gender	n/a
Reporting on race, ethnicity, or other socially relevant groupings	n/a
Population characteristics	n/a
Recruitment	n/a
Ethics oversight	n/a

Note that full information on the approval of the study protocol must also be provided in the manuscript.

Field-specific reporting

Please select the one below that is the best fit for your research. If you are not sure, read the appropriate sections before making your selection.

☒ Life sciences ☐ Behavioural & social sciences ☐ Ecological, evolutionary & environmental sciences

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Life sciences study design

All studies must disclose on these points even when the disclosure is negative.

Sample size	n/a
Data exclusions	No data were excluded
Replication	All data, including replicates, are included and discussed in manuscript and supplementary information
Randomization	n/a
Blinding	n/a

Reporting for specific materials, systems and methods

We require information from authors about some types of materials, experimental systems and methods used in many studies. Here, indicate whether each material, system or method listed is relevant to your study. If you are not sure if a list item applies to your research, read the appropriate section before selecting a response.

Materials & experimental systems

n/a	Involved in the study
<input checked="" type="checkbox"/>	<input type="checkbox"/> Antibodies
<input checked="" type="checkbox"/>	<input type="checkbox"/> Eukaryotic cell lines
<input checked="" type="checkbox"/>	<input type="checkbox"/> Palaeontology and archaeology
<input checked="" type="checkbox"/>	<input type="checkbox"/> Animals and other organisms
<input checked="" type="checkbox"/>	<input type="checkbox"/> Clinical data
<input checked="" type="checkbox"/>	<input type="checkbox"/> Dual use research of concern
<input checked="" type="checkbox"/>	<input type="checkbox"/> Plants

Methods

n/a	Involved in the study
<input checked="" type="checkbox"/>	<input type="checkbox"/> ChIP-seq
<input checked="" type="checkbox"/>	<input type="checkbox"/> Flow cytometry
<input checked="" type="checkbox"/>	<input type="checkbox"/> MRI-based neuroimaging

Plants

Seed stocks

n/a

Novel plant genotypes

n/a

Authentication

n/a

Transient Non-local Interactions Dominate the Dynamics of Measles Virus N_{TAIL}

Corresponding Author: Dr Sara Vaiana

This file contains all reviewer reports in order by version, followed by all author rebuttals in order by version.

Version 0:

Reviewer comments:

Reviewer #1

(Remarks to the Author)

This paper is a well-executed exploration of how transient non-local interactions shape the behavior of the measles virus N_{TAIL} domain. The authors do an excellent job combining experimental techniques with computational modeling to uncover previously unknown interactions that influence the structure and function of this intrinsically disordered protein. Their findings not only deepen our understanding of N_{TAIL}'s structural properties but also shed light on its crucial role in viral transcription and replication.

The paper is surely well written, the conclusions are sound and well explained. The use of photo-induced electron transfer (PET) experiments allows for an unusually precise look at how different regions of N_{TAIL} interact in real time—an approach that overcomes many of the challenges associated with studying intrinsically disordered proteins. The authors go a step further by validating their experimental results with a combination of coarse-grained and all-atom molecular dynamics simulations, carefully ensuring that their computational models align with real-world data. Their attention to force field validation and systematic cross-checking with experimental results makes this an exceptionally robust and convincing study.

Beyond just the methods, the biological significance of the findings is also appealing. The discovery that N_{TAIL}'s conformational behavior is shaped by non-local electrostatic interactions provides a new perspective on how this domain regulates binding to the phosphoprotein X-domain (PXD) and, ultimately, the recruitment of the viral polymerase. The way in which specific charged residues influence the structural preferences of N_{TAIL}—and how this is further supported by coevolutionary analysis—adds another layer of importance to the work.

I can only recommend acceptance

Reviewer #2

(Remarks to the Author)

In this work, the authors present a compelling and impactful set of experiments and simulations to determine the role of flanking regions in the regulation of N_{tail}. They conclude that transient interactions driven by electrostatics in the flanking regions have a significant effect on N_{tail} conformations and thus function. The PET experiments show clear heteropolymeric behavior, with charge interactions in two regions of the protein having opposing effect. Both the coarse grained and all atom simulations bring complementary information to understand the protein dynamics. In general, I found the overall work and conclusions to be very robust. I had only fairly minor suggestions or questions.

There appear to be several steps in converting between simulation and PET experiments, and so plotting the ratio of the different rates is important. In Figure 2B, the ratio is between the reference condition and different buffer conditions was used, however a more relevant comparison might be the differences between PET constructs. Without this, it's hard to determine how well the simulations reproduce either the overall trend or the deviations relative to the trend as in figure 1C. Can you plot something like Figure 1C for both types of simulations? Even if it's somewhat different, e.g. the appropriate distance distribution between relative amino acids even if they aren't mutated in the simulations? That is, comparisons like Figure S22, but for all of the constructs seem very important for validating the comparison between experiment and simulations.

The analysis on the four conformations in the all atom simulations is quite nice, but it would also be nice to see a similar analysis averaged across the whole simulation. Contact maps or analysis like figure 2A would both be very useful.

It would be nice to include a small figure of charge density along the sequence, that is the diagonal of Fig 2 top is important enough to take a bit more space.

I don't understand the charge state of D and E residues at pH 4 in the coarse grained simulation. Are they chosen randomly to be charged or not charged and then fixed at that charge, given fractional charges, or stochastically changed throughout the simulation. The above plot could include the simulation charge distribution both at physiological pH and at pH 4.

It would be nice if figure 2C could include comparisons to experiment for the salt condition as well, although I'm unclear if that experiment has been done. Perhaps it is a different ratio. The presentation is actually a bit confusing, the dots in figure 2B are hard to see, but clearer than the lines in Figure 2C, since the 2C lines are only comparisons to a single condition.

The calculated bare rates for quenching as a function of pH are the same within experimental error, but still almost 20% different. Is it possible to get more accuracy on these measurements? It might be useful to correct the comparisons between the protein PET results.

The all atom simulations were (understandably) only done at one pH, and if I follow the logic correctly, the argument is that since the "physiological" PET rates being so very different between C3W2 and C4W2 is recapitulated in the all atom model, then they are relevant. However, the coarse-grained model were also able to reproduce that difference between rates, but unable to capture the changes due to the pH shift. This is all fine, but from a writing perspective, it makes it unclear whether the all atom have really succeeded in reproducing the experimental data. I wonder, simply from a story-telling perspective, if it would better to focus on the increased sequence information given by the all atom rather than the limitations of the coarse grained model.

Reviewer #3

(Remarks to the Author)

In this study, the authors explored how certain internal interactions within the measles virus nucleoprotein (NTAIL) affect its behavior and role, especially in its connection with the polymerase complex, which helps the virus replicate and transcribe. They used experimental and computational methods and discovered that short-lived interactions between separate parts of NTAIL are crucial in shaping its behavior and interactions with other proteins, influencing the virus's ability to replicate and transcribe. The findings suggest these types of interactions might also be important in other similar proteins, although this was not directly investigated or substantiated.

Overall the paper provides a thorough and valuable analysis that contributes significantly to our understanding of these proteins and should definitely be published.

Concerns:

While I have no major concerns about the scientific value or quality of the research, the presentation could be clearer and less technical. I honestly had a hard time following many parts of the text and managing the large number of supplementary figures and tables. I know they are important, but they sometimes made it hard to follow the logic of the paper. In summary, many times I felt that the paper was too technical for a broader chemistry audience journal.

On the other hand, I missed some information that would be better presented in a table. For greater clarity on the computational work, it would help to include details like the duration of each simulation in a table. This table could also specify the number of replicates and other details that distinguish each simulation. This would make it easier for others to replicate the study and to assess whether the results are based on a single analysis or a robust study with multiple replicates that also considers the statistical significance of the results. I honestly don't know at this point the answer for this last point.

A couple of specific minor suggestions:

1. Rearrange Figure S24 to improve visibility. Consider placing part A above part B and enlarging A, as it is quite hard to see the details described on page 35.
2. In Table S1, please clarify what the colors represent.

Reviewer #4

(Remarks to the Author)

The manuscript by Otteson et al. the interactions between amino acid residues that are distant in sequence of the N_{tail} protein of the measles virus are investigated using photo-induced electron transfer (PET) and Molecular Dynamics (MD) simulations.

From detected PETs and analysis of extensive MD simulations, the temporary and short-lived interactions between introduced tryptophan and cysteine pairs can be seen. An estimate of the rate of conversion between states S1-S4 from a Markov model would be beneficial but is not absolutely critical here.

This is interesting since the intrinsically disordered protein (IDP) N_{tail} is so flexible and showing a multi-dimensional conformational surface, that this information cannot be obtained otherwise.

It is clear that the authors extend the system beyond the MoRE region flanked by residues of N_{tail} to either side but we would need to see a discussion of previous work by NMR and all-atom MD (their references 19 and 26). Does this region retain its secondary structure in solution in experiment and solution and how do the results of this work agree with refs 19 and 26.

In experiment and MD simulations, the authors show that it is electrostatic interactions between differently positioned amino acid residues, which is responsible for the heteropolymeric behavior of N_{tail}.

It remains to be mentioned critically, that the first coarse-grain (MD) simulations required the adjustment of parameters to be able to get the radius of gyration and the helical propensity of the MoRE region close to experimental values from SAXS and NMR.

When altering the pH in experiment and simulations, a lower pH of 4.0 leads to 'more neutral' Asp and Glu residues and neutral His will become 'more positively charged'. This statement is correct but a quantification is needed, how many protons will be added to these residues at pH 4.0. It requires an estimation of the pKa values of the individual residues and then an annotation of change of N_{tail} residue by pKa.

On the other hand, on p 13, line 9 the authors state that changes in PET are 'entirely due to changes in the protein, rather than pH-induced changes of the electron transfer efficiency'. It is clear that the first will effect the latter. pH-induced changes in protein structure will critically effect the distances between W and C residues.

In Figure 4, the side chain salt bridge and C_{alpha} backbone interactions are shown as contact maps. These are very hard to see and a better coloring scheme or radius is required.

On p 18, line 3, the authors claim to have validated the choice of force field for the protein and water against experimental CD and PET experiments. However, we see only two combinations CHARMM36m/OPC and AMBER99SB-disp/TIP4P in the methods part. The authors need to clarify this.

In Figure 5, the dashed lines are not visible and the figure has to be improved in clarity.

The conclusion section spans from pages 26 to 36 and appears before the summary. First, the summary ought to be communicated to the reader and then a SHORT section of conclusions can be drawn to speculate about the results of this work on isolated N_{tail} with relevance to its physiological process. However, the latter is tentative only and should be presented in a concise way.

Overall, the study is interesting and demonstrates the complexity of the conformational dynamics and flexibility of intrinsically disordered proteins and their possible function in biological processes. Rationalizing PET experiments for two pairs of mutated C/W residues requires VERY extensive sampling by classical MD simulations.

Version 1:

Reviewer comments:

Reviewer #2

(Remarks to the Author)

The revisions significantly improved the manuscript and I have no further concerns.

Reviewer #3

(Remarks to the Author)

The revised form of the manuscript is much easier to read. Although the topic is quite complex, I believe it is now

significantly more accessible to a broad readership. I had no major concerns with the previous version, and I believe that the new revision only strengthens the presentation of the results. I recommend publication.

Reviewer #4

(Remarks to the Author)

The revised manuscript reads much better and all issues with the original were resolved. I can thus only recommend its publication.

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Response to Referees:

We would like to thank the Reviewers for their hard work and constructive comments. In response to the Reviewer's comments, we extensively modified our manuscript, along three main lines:

- we clarified the text of the manuscript to make it more accessible to a broader audience, significantly rewriting entire sections, and reorganizing the material to simplify the reading,
- we added short summaries of our methodology in the main text, wherever needed to allow the reader to understand the essentials of our results without having to rely on the supplementary materials,
- we improved the text to avoid technical jargon.

In the revised manuscript, all changes to the text are highlighted in red.

In addition to these three main issues, we addressed all specific comments, of each reviewer as described in detail below.

Reviewer #1

This paper is a well-executed exploration of how transient non-local interactions shape the behavior of the measles virus NTAIL domain. The authors do an excellent job combining experimental techniques with computational modeling to uncover previously unknown interactions that influence the structure and function of this intrinsically disordered protein. Their findings not only deepen our understanding of NTAIL's structural properties but also shed light on its crucial role in viral transcription and replication.

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I can only recommend acceptance

We thank the reviewer for their detailed and positive evaluation of our manuscript!

Reviewer #2

In this work, the authors present a compelling and impactful set of experiments and simulations to determine the role of flanking regions in the regulation of Ntail. They conclude that transient interactions driven by electrostatics in the flanking regions have a significant effect on Ntail conformations and thus function. The PET experiments show clear heteropolymeric behavior, with charge interactions in two regions of the protein having opposing effect. Both the coarse grained and all atom simulations bring complementary information to understand the protein dynamics. In general, I found the overall work and conclusions to be very robust. I had only fairly minor suggestions or questions.

Comment 2-1:

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We appreciate the reviewer's insightful suggestion. Indeed, comparing across the five PET constructs, as done for the experiments in Figure 1C, would provide important context for validating the simulation trends. However, to determine the overall relaxation-time trends between PET constructs both of CG and all-atom simulations requires additional assumptions we were originally reluctant to make. Encouraged by the reviewer's suggestion, we have now added the requested comparisons, as well as the underlying assumptions.

For the CG simulations, computing absolute PET rates would require knowledge of prefactors, which are now discussed on page 15 of the revised manuscript. Although these prefactors cannot be determined from the simulations, in the revised manuscript, we have assumed that they are identical for all C-W pairs, and have estimated them by fitting a homopolymer model to CG simulation data, rescaling them as discussed in the revised manuscript. The resulting plot is now shown in Fig 2D.

For the all-atom simulations, we computed the absolute relaxation rates for C3W2 and C4W2, which are now added and discussed on page 18 of the revised manuscript. However, in order to compute relaxation rates for the other three variants (C1W1, C2W1, C3W1), we have to assume that the short-range contact dynamics is unaffected by Y->W mutations at the W1 position. Because the size and shape of the two side chains differ, our predicted PET relaxation times are probably less accurate even if the overall dynamics of Ntail is captured accurately by the simulations. To take this effect partially into account, we have now calculated the relaxation times for these variants using a slightly larger (0.6 nm instead of 0.4 nm) contact distance cutoff, as shown in the new supplementary Figure S30. The figure shows that the general trends are indeed captured by the all-atom simulations, although the C3W1 relaxation times are slightly overestimated, potentially due to insufficient sampling of the nonlocal interactions between the regions B and D.

Together, these figures allow for a more direct comparison to the experimental trends in Figure 1C (Fig. 1D in the revised version), and help assess to what extent the simulations capture both the expected scaling and the deviations across constructs. We believe this addresses the reviewer's

concern and strengthens the interpretation of our comparison between coarse-grained simulations and experiments.

Comment 2-2:

The analysis on the four conformations in the all atom simulations is quite nice, but it would also be nice to see a similar analysis averaged across the whole simulation. Contact maps or analysis like figure 2A would both be very useful.

We agree with the reviewer's suggestion, and now provide an averaged contact map in the supplementary information (Fig. S27).

Comment 2-3:

It would be nice to include a small figure of charge density along the sequence, that is the diagonal of Fig 2 top is important enough to take a bit more space.

We gladly agree, and have now added a net charge per residue plot in Fig. 1B.

Comment 2-4:

I don't understand the charge state of D and E residues at pH 4 in the coarse grained simulation. Are they chosen randomly to be charged or not charged and then fixed at that charge, given fractional charges, or stochastically changed throughout the simulation. The above plot could include the simulation charge distribution both at physiological pH and at pH 4.

The description of the charge treatment of ionizable residues in the coarse-grained simulations was not sufficiently clear. In our model, we assigned fixed fractional charges to titratable residues based on the Henderson-Hasselbalch equation $\text{pH} = \text{pK}_a + \log([A^-]/[HA])$ and pK_a values for side chains of Aspartic acid (3.65), Glutamic acid (4.25) and Histidine (6.00). We have now clarified this aspect on page 12 of the revised manuscript as well as in Supporting Methods S5.

Comment 2-5:

It would be nice if figure 2C could include comparisons to experiment for the salt condition as well, although I'm unclear if that experiment has been done. Perhaps it is a different ratio. The presentation is actually a bit confusing, the dots in figure 2B are hard to see, but clearer than the lines in Figure 2C, since the 2C lines are only comparisons to a single condition.

We agree with the reviewer that Fig. 2B and 2C were difficult to interpret, and we have now revised these panels accordingly, also taking the reviewer's previous suggestion (2-1) into account. We note that all available experimental data have been included in Fig. 2B (not all variants were measured in high salt/low pH conditions due to technical difficulties). This is now also indicated in the caption of Fig. 2. We believe the current Fig. 2, with the new panel D and modified B is much easier to interpret, and we thank the reviewer for that.

Comment 2-6:

The calculated bare rates for quenching as a function of pH are the same within experimental error, but still almost 20% different. Is it possible to get more accuracy on these measurements? It might be useful to correct the comparisons between the protein PET results.

Improving the uncertainty of CW relaxation times derived from PET experiments is a continuous effort in our lab. We have made every effort to maximize the accuracy of our bimolecular quenching rate measurements, including careful calibration and repeated experiments to minimize variability. The current level of precision reflects the best achievable with our present methodology. We have added a paragraph clarifying how the error is estimated in the Supporting Methods S2. As the reviewer points out, the observed variation falls within experimental uncertainty and therefore does not impact our conclusions. While further reducing this uncertainty remains an ongoing goal in our lab, it would require substantial technical advances beyond the scope of the current study.

Comment 2-7:

The all-atom simulations were (understandably) only done at one pH, and if I follow the logic correctly, the argument is that since the “physiological” PET rates being so very different between C3W2 and C4W2 is recapitulated in the all atom model, then they are relevant. However, the coarse-grained model were also able to reproduce that difference between rates, but unable to capture the changes due to the pH shift. This is all fine, but from a writing perspective, it makes it unclear whether the all atom have really succeeded in reproducing the experimental data. I wonder, simply from a story-telling perspective, if it would better to focus on the increased sequence information given by the all atom rather than the limitations of the coarse grained model.

The reviewer is correct that all-atom simulations accurately capture the dynamic heterogeneity between the C3W2 and C4W2 relaxation times (2.7 ratio) that was observed in measurements (2.8 ratio) and therefore are relevant for investigating the molecular origin of this phenomenon. Although the coarse-grained simulations predicted a dynamic heterogeneity for the two PET constructs, the extent of the heterogeneity (1.6 ratio) was much closer to that of the homopolymer (1.3 ratio) than the measurements.

We believe that the reviewer was confused about this point because in the original version of the manuscript, the all-atom and CG ratios were not quantified in the main text, and the CG ratio was not presented clearly in Fig. 2D. Although the all-atom relaxation times were presented and discussed in Supporting Methods S17 in the original version, the dynamic heterogeneity and C3W2/C4W2 ratios were not explicitly quantified there either. This was a major shortcoming of the original manuscript, and we would like to thank the reviewer again for pointing it out. In the revised manuscript we have changed Fig. 2D, quantified and discussed the ratios in detail in the main text. We have also restructured pages 12-19 to present our results more clearly, and to highlight the strengths and limitations of both CG and all-atom simulations.

Finally, to reinforce our claim that the all-atom simulations did indeed reproduce the experimental data accurately we now have also estimated the other three relaxation times as well (shown in Fig. S30). In addition, we would like to point out that all-atom simulations also reproduced other types of independent experimental data as well, such as CD spectra for the wild-type and W2 variants, as well as SAXS and NMR chemical shifts for wild type N_{TAIL}, which further increased our confidence in the simulations.

Reviewer #3

In this study, the authors explored how certain internal interactions within the measles virus nucleoprotein (NTAIL) affect its behavior and role, especially in its connection with the polymerase complex, which helps the virus replicate and transcribe. They used experimental and computational methods and discovered that short-lived interactions between separate parts of NTAIL are crucial in shaping its behavior and interactions with other proteins, influencing the virus's ability to replicate and transcribe. The findings suggest these types of interactions might also be important in other similar proteins, although this was not directly investigated or substantiated.

Overall the paper provides a thorough and valuable analysis that contributes significantly to our understanding of these proteins and should definitely be published.

We thank the referee for the very positive assessment.

Comment 3-1:

While I have no major concerns about the scientific value or quality of the research, the presentation could be clearer and less technical. I honestly had a hard time following many parts of the text and managing the large number of supplementary figures and tables. I know they are important, but they sometimes made it hard to follow the logic of the paper. In summary, many times I felt that the paper was too technical for a broader chemistry audience journal.

We appreciate the comments of the reviewer regarding the clarity, technicality, and reliance on supplementary information to communicate our points. In fact, after re-reading our original manuscript in light of these comments, we fully agree. We have therefore extensively revised the results and discussion sections following the reviewer's comments, which we feel has improved the readability substantially (see also our summary reply to the Editor above). Specifically, we now include summaries of the supplementary materials, so that the reader does not have to switch back and forth between the main text and supplements to understand the main lines of our reasoning. We also improved the logical flow of the text, and have worked hard to make the text more easily understandable for non-expert readers with a chemistry background.

Comment 3-2:

On the other hand, I missed some information that would be better presented in a table. For greater clarity on the computational work, it would help to include details like the duration of each simulation in a table. This table could also specify the number of replicates and other details that distinguish each simulation. This would make it easier for others to replicate the study and to assess whether the results are based on a single analysis or a robust study with multiple replicates that also considers the statistical significance of the results. I honestly don't know at this point the answer for this last point.

We agree and have now added two supplementary tables (Tables S8 and S9) that provide simulation lengths and replicate numbers. Regarding the robustness of our simulation results, the large error bars on PET estimates and large variation of the secondary structure content across replicates might suggest that our simulations are not fully converged yet. However, sampling in our simulations is robust enough to simultaneously explain or reproduce several of the experimental observations about N_{TAIL} dynamics, by one set of simulations. Further the two most accurate tested force fields, CHARMM36m-OPC and Amber99SB-disp, independently predict a similar dynamic heterogeneity for the C3W2 and C4W2 variants, as well as a free-energy landscape with similar features (data not shown), which also underscores the robustness of our all-atom simulations.

Comment 3-3:

Rearrange Figure S24 to improve visibility. Consider placing part A above part B and enlarging A, as it is quite hard to see the details described on page 35.

Thank you for the suggestion. We resized and adjusted the Figure accordingly.

Comment 3-4 :

In Table S1, please clarify what the colors represent.

Thank you! We have now added the necessary clarification to the Table header.

Reviewer #4:

The manuscript by Otteson et al. the interactions between amino acid residues that are distant in sequence of the N_tail protein of the measles virus are investigated using photo-induced electron transfer (PET) and Molecular Dynamics (MD) simulations.

Comment 4-1:

From detected PETs and analysis of extensive MD simulations, the temporary and short-lived interactions between introduced tryptophan and cysteine pairs can be seen. An estimate of the rate of conversion between states S1-S4 from a Markov model would be beneficial but is not absolutely critical here.

This is interesting since the intrinsically disordered protein (IDP) Ntail is so flexible and showing a multi-dimensional conformational surface, that this information cannot be obtained otherwise.

We thank the reviewer for the positive assessment, and agree that a simple and intuitive kinetic model would be useful to describe the transition rates between the four conformational states of Ntail. In fact, we did try fitting Markov models to the free energy landscape before submission and obtained averaged transition rates between the states. Unfortunately, upon closer inspection, the shallow free-energy landscape of Ntail, as for many other IDPs, renders analysis with Markov models challenging and potentially misleading.

Specifically, we noticed that – at least when using a number of Markov states small enough to facilitate an intuitive understanding – several criteria of Markovianity were violated. In particular, the duration of actual transitions between the states were generally not much shorter than the preceding dwell times. Further, we observed multiple transition modes with different kinetics between the same pairs of states. Addressing these challenges would have either required a large number of Markov states along these transitions or the use of non-Markovian kinetic models including memory effects. In both cases, it is unlikely that our sampling would have sufficed to define the much larger number of parameters of the models. Additionally, both of these solutions would likely yield less intuitive results than a simple Markov model, would have required the description of yet another complex method, and would make maintaining the actual storyline and focus even more challenging.

For these reasons, we decided to only include the free-energy landscape in Fig. 3 in the manuscript, which suffices to support our main conclusions. Nevertheless, and encouraged by the referee's comment, on pages 18-20 of the revised manuscript, we now provide a more in-depth description of the free energy landscape to better explain how non-local interactions give rise to the observed dynamic heterogeneity.

Comment 4-2:

It is clear that the authors extend the system beyond the MoRE region flanked by residues of N_{tail} to either side but we would need to see a discussion of previous work by NMR and all-atom MD (their references 19 and 26). Does this region retain its secondary structure in solution in experiment and solution and how do the results of this work agree with refs 19 and 26.

We thank the reviewer for this request, and have added such a discussion to the Supporting Methods S23 section of the revised manuscript as described below. We would like to stress, however, that while such consistency checks with existing literature are certainly helpful and fair, the main conclusions of our manuscript rest on our direct comparisons to CD and NMR measurements. Here is our summary of previous work, to our best knowledge:

Robustelli *et al.* (ref 26) used the Amber99SB-disp force field to simulate the MoRE region, and found a 20% helix propensity, and good agreement with C α chemical shift (RMSD: 0.52 ppm, and R: 0.92). Wang *et al.* (ref 19) performed temperature replica exchange MD simulations using the CHARMM22* force field, and estimated a similar 20% helix content in the MoRE region, with a slightly lower Pearson Correlation (R: 0.76). Both simulations assigned considerably higher helical propensity (50-60%) in P_{XD} the binding site (491-501) than the preceding segment (482-490).

Our two best wild-type N_{TAIL} simulations were performed using CHARMM36m-OPC force field (17% helix content, RMSD: 0.47 ppm, and R: 0.83) and the Amber99SB-disp force field (18% helix content, RMSD: 0.36 ppm, R: 0.84) provided by Piana *et al.* Both of these forcefields predict a transient helix at the MoRE that both folds and unfolds. The two force fields also agree in that the binding site (491-501) has a higher content than the preceding segment.

As far as we can tell, no NMR data is available for the PET variants, but the measured CD spectra of W2 variant indicate a change in the secondary structure composition compared to the wild type construct (likely with a small decrease of the helix content). Simulation trajectories C3W2 and C4W2 variants using CHARMM36m-OPC show a smaller helix content (~10%) which is more compatible with the W2 CD spectrum. In contrast, Amber99SB-disp simulations predict an increased helix content in the MoRE region (~29%) as well as in the transient helices outside of it. The helix content of MoRE residues in our simulations are summarized in Supplementary Figure S31.

Comment 4.3:

In experiment and MD simulations, the authors show that it is electrostatic interactions between differently positioned amino acid residues, which is responsible for the heteropolymeric behavior of N_{tail}. It remains to be mentioned critically, that the first coarse-grain (MD) simulations required the adjustment of parameters to be able to get the radius of gyration and the helical propensity of the MoRE region close to experimental values from SAXS and NMR.

We fully agree and now clearly state in the revised version of the manuscript (page 10) that the CG models for N_{TAIL} were fitted to reproduce SAXS and NMR data. Coarse-grained models offer computational efficiency, but they often do not capture all sequence-specific features accurately. This is one of the reasons, why we also included atomistic simulations, despite the considerable computational effort.

Comment 4-4:

When altering the pH in experiment and simulations, a lower pH of 4.0 leads to 'more neutral' Asp and Glu residues and neutral His will become 'more positively charged'. This statement is correct but a quantification is needed, how many protons will be added to these residues at pH 4.0. It requires an estimation of the pKa values of the individual residues and then an annotation of change of N_tail residue by pKa.

Indeed, the description in the original manuscript was somewhat vague. In the revised version of manuscript (p. 12-13), it is now clearly stated that pH effects in CG simulations are modeled by partial charges that reflect the average protonation state of protonatable residues under the relevant conditions. Such a simplified approach is suitable at the level of CG modeling, where the goal is to represent overall electrostatic changes rather than detailed titration states. By contrast, all-atom simulations at low pH would indeed require assignment of discrete protonation states or the use of constant-pH molecular dynamics methods, which would certainly be interesting in future work.

Comment 4-5:

On the other hand, on p 13, line 9 the authors state that changes in PET are 'entirely due to changes in the protein, rather than pH-induced changes of the electron transfer efficiency'. It is clear that the first will effect the latter. pH-induced changes in protein structure will critically effect the distances between W and C residues.

We thank the reviewer for pointing that our wording was unclear. We agree that pH-induced structural changes in the protein directly affect C-W distances and thus affect the measured relaxation times due to PET. However, our statement was intended to clarify that the intrinsic electron transfer efficiency (at contact) is not significantly altered by pH. As shown in Figure S19, the bimolecular quenching rate between W and C freely diffusing in solution, varies only slightly across different pH conditions, indicating that changes in PET rates observed in the constructs are primarily due to structural rearrangements rather than shifts in the intrinsic quenching efficiency. We have now revised the text to make this distinction clearer.

Comment 4-6:

In Figure 4, the side chain salt bridge and C_alpha backbone interactions are shown as contact maps. These are very hard to see and a better coloring scheme or radius is required.

We agree that the individual interactions are difficult to see in Fig. 4. Although we note that the primary function of the figure is to highlight the main interaction clusters that stabilize the four conformational states and the differences between them, we have now improved the visibility of individual salt bridges and sporadic interactions. Specifically, we changed the color scheme of the contact maps and applied an interpolation method to increase salt-bridge pixel size. In addition, we provided the stand-alone contact maps at higher resolution for each state in the supplementary materials (Supplementary Figs. S23-S26, without interpolation).

Comment 4-7:

On p 18, line 3, the authors claim to have validated the choice of force field for the protein and water against experimental CD and PET experiments. However, we see only two combinations CHARMM36m/OPC and AMBER99SB-disp/TIP4P in the methods part. The authors need to clarify this.

We would like to clarify that we did the validation in two steps, which may not have been apparent in the original version of the manuscript, as most of the discussion was relegated to the Supporting Methods S10. In the first step we used 8 different force fields to simulate WT N_{TAIL}, for which extensive experimental data were available. In the second step, we used the resulting two best performing forcefields and performed simulations on the two PET variants (C3W2 and C4W2) to address the dynamic heterogeneity. The PET calculations were performed only for the PET variant simulations, as accurate PET predictions require a tryptophan residue at the appropriate sequence position to capture the short-range contact dynamics of C-W pairs. For this reason, we only computed and presented PET comparisons for the best two forcefields. To avoid this potential confusion, we have now revised our description of the forcefield validation, and highlighted the outlined points on page 17.

Comment 4-8:

In Figure 5, the dashed lines are not visible and the figure has to be improved in clarity.

Thank you! We made the dashed lines more visible in the Figure, streamlined the figure contents, and added clarification to the caption.

Comment 4-9:

The conclusion section spans from pages 26 to 36 and appears before the summary. First, the summary ought to be communicated to the reader and then a SHORT section of conclusions can be drawn to speculate about the results of this work on isolated N_{tail} with relevance to its physiological process. However, the latter is tentative only and should be presented in a concise way.

We agree with the reviewer that our manuscript organization was confusing in this regard. In the revised manuscript we therefore restructured the main text as suggested. The Summary was renamed to Conclusions (which is now shorter than one page), and the previous “Conclusions” section has been more fittingly renamed “Discussion”. In the process, we have also shortened it by two pages.

Although tentative on its own, we believe that our discussion about the physiological aspects of our results is an important part of the manuscript that is supported by a body of evidence available in the literature. Two of four reviewers specifically highlighted this part and appear to have shared this belief. In line with the referee’s remarks, we now present this part more concisely without loss of information.

Comment 4-10: (Summary)

Overall, the study is interesting and demonstrates the complexity of the conformational dynamics and flexibility of intrinsically disordered proteins and their possible function in biological processes. Rationalizing PET experiments for two pairs of mutated C/W residues requires VERY extensive sampling by classical MD simulations.

We thank this Reviewer for the overall positive assessment and for the many suggestions which helped us to substantially improve our manuscript.

Supplementary Information for “Transient Non-local Interactions

Dominate the Dynamics of Measles Virus N_{TAIL}”

A. Supporting Methods

S1: N_{TAIL} Protein Samples

In addition to wild type N_{TAIL}, seven variants were studied in this work, namely Y451W (W1), Y518W (W2), S407C-Y451W (C1W1), S425C-Y451W (C2W1), Y451W-S482C (C3W1), S482C-Y518W (C3W2), and S488C-Y518W (C4W2) (Supporting Table S1). In these, one of the two tyrosines of wild type N_{TAIL}, at positions 451 and 518, was selectively substituted with tryptophan. And one of the serines, at positions 407, 425, 482 or 488 was substituted with cysteine. These mutations were chosen because they introduce minimal perturbations to the sequence, and because several of them had already been introduced and tested in previous studies. Specifically, single-site Y518W, S407C, S482C and S488C N_{TAIL} variants were used in N_{TAIL}-P_{XD} binding studies¹⁻⁴. Proteins were expressed in *E. coli*, extracted, and further purified. In the end, all sequences studied here were free of hexahistidine tag.

Generation of the bacterial expression constructs

In a first step, Y451W and Y518W mutations were separately introduced by overlap extension PCR using wild-type N_{TAIL} coding sequence borne by the Gateway expression vector pETG20A³ as template and, respectively, mutagenic forward (F) and reverse (R) primer pairs Y451WF / Y451WR and Y518WF / Y518WR. After DpnI treatment, each of the two PCR products was inserted in the Gateway shuttle vector pDONR using BP clonase (Invitrogen). The sequence of the insert was checked by DNA sequencing, and inserts were transferred from pDONR to pETG20A using LR clonase (Invitrogen). In a second step, S407C, S425C, S482C and S488C mutations were individually introduced as described above using either the Y451W or the Y518W construct as template and, respectively, mutagenic primer pairs S407CF / S407CR, S425CF / S425CR, S482CF / S482CR and S488CF / S488CR. All the resulting constructs drive the expression, under the T7 promoter, of a thioredoxin (TRX) tagged fusion protein in which the N_{TAIL} protein is preceded by a hexahistidine (His6) tag and a Tobacco Etch virus (TEV) protease cleavage site (TRX-His-Tev-N_{TAIL}). The TRX tag, the hexahistidine tag and the AttB1-encoded amino acid stretch can be cleaved off by TEV protease digestion thus leading to a protein bearing only two non-native N-terminal residues (GS).

Protein expression and purification

T7 cells (New England Biolabs) were transformed with the plasmid bearing the sequence coding for the fusion protein, and transformed cells were selected on agar plates containing 100µg /mL ampicillin and 34µg /mL chloramphenicol. After one night at 37°C, a few colonies were used to seed 50 ml of LB containing the same antibiotics. The preculture was incubated overnight at 37°C with constant shaking at 200 rpm. The next day, four to six liters of terrific broth (TB) containing 100µg /mL ampicillin and 34µg /mL chloramphenicol were seeded with 10mL of preculture per liter of culture and grown at 37°C with constant shaking at 200 rpm until OD at 600 nm reached 0.5 to 0.8. IPTG was added at the final concentration of 0.5 mM and expression was allowed to proceed overnight at 27°C under constant shaking. The culture was then centrifuged for 30' at 4000g. The supernatant was discarded and the cell pellet was resuspended in 10 mL of 8 M urea, 50 mM Tris pH8, 0.3 M NaCl per liter of culture, and frozen. After thawing, the cell lysate was sonicated at room temperature and spun for 30' at 15,000g at 25°C. The supernatant was supplemented with 1 mL per liter of bacterial culture of Ni-sepharose fast flow resin (Cytiva) pre-equilibrated in 8 M urea, 50 mM Tris pH8, 0.3 M NaCl and incubated for 30' on a rotating wheel at room temperature. Beads were recovered by

centrifugation for 1' at 5000g and washed five times with one volume (i.e. the same volume of beads) of 8 M urea, 50 mM Tris pH8, 0.3 M NaCl. Fusion proteins were eluted from the beads with five volumes of 8 M urea, 50 mM Tris pH8, 0.3 M NaCl, 250 mM imidazole. Pooled eluted fractions were dialyzed overnight at room temperature against five volumes of 50 mM Tris pH8, 0.3 M NaCl. The next day, the dialysis buffer was discarded and His-tagged TEV protease was added at 1/26 dilution (weight / weight, with respect to the amount of dialyzed protein) in the dialysis bag. A second dialysis was then performed against five volumes of 10 mM Tris pH8, 0.3 M NaCl at 4°C for seven hours. The content of the dialysis bag was then supplemented with 1 mL per liter of culture of pre-equilibrated Ni-sepharose resin and incubated for 30' on a rotating wheel at 4°C. Ni beads were recovered by centrifugation for 1' at 5000g and the supernatant containing the protein of interest was frozen. This Ni affinity step allows His-tagged proteins (i.e., uncut fusion proteins and TEV protease) to be removed, and the protein of interest to be recovered in the unbound material. After thawing, the protein was concentrated using a centricon (cutoff: 10 kDa) and the concentrated sample was kept at -20°C and shipped in dry ice.

Further protein purification

Initial protein purity before shipping was analyzed by electrophoresis (SDS-PAGE), absorption, and circular dichroism (CD). For storage and transportation between labs, purified proteins were dialyzed into neutral pH 15 mM Tris, 150 mM NaCl, 1 mM TCEP buffer and kept frozen in several small Eppendorf tubes. After shipping and before experiments, variants were repurified by reverse phase HPLC on a semipreparative C18 column (Vydac) and lyophilized. In all cases, protein purity and photophysical properties of W was confirmed after suspension of the lyophilized powder, by analytical HPLC, ESI-QTOF mass spectrometry (Agilent 6530 Quadrupole TOF LC-MS) and absorption spectroscopy (Cary 50 UV-Vis spectrophotometer from Agilent). Unless otherwise noted, all reagents and solvents used in this study were analytical grade obtained from Sigma. Deionized (DI) water was Millipore grade, from Evoqua Water Technologies filtration system. Lab-made buffers were pH adjusted with acid titration (Accumet AB 15, Fisher Scientific). To prevent dimerization due to C-C disulfide bridge formation, all samples contained 1 mM TCEP (freshly added to samples before measurements, in a way to ensure final pH values reported in manuscript).

S2: Photo-induced Electron Transfer (PET) experiments

Measuring tryptophan-cysteine quenching times due to PET

To quantify the intra-protein dynamics of full-length N_{TAIL} in solution, we probed intra-molecular contact formation, using a technique based on photo-induced electron transfer (PET) between a tryptophan (W) and cysteine (C)⁶⁻⁹. For these experiments, we used the N_{TAIL} variants of Supporting Table S1, produced as described above, containing a unique W and C placed at different positions within the sequence. A nanosecond UV excitation is used to excite W to its triplet state, and the population of this state is monitored as a function of time through transient absorption⁸. In the absence of quenchers, the W triplet state is long-lived and eventually decays to the ground state via its normal relaxation routes. In the presence of C, the W triplet state relaxes via an additional route, due to the transfer of an excited state electron from W to C. This occurs when the two amino acids come within Van der Waals distance of each other, via stochastic collisions, as the disordered protein samples its conformational states in solution. The observed relaxation rate (the inverse of the triplet state lifetime) is the sum of the natural relaxation rate in the given solution conditions, k_o , and the C-W quenching rate due to contact formation, k_{CW} :

$$k_{obs} = k_o + k_{CW}$$

At the protein concentrations used in this study, intra-molecular collisions between W and C placed on the same protein, are much faster (100ns-us time scales) than inter-molecular ones (between W and C placed on different proteins). The C-W quenching rate k_{CW} therefore reflects

the rate at which those two positions in the IDP contact each other⁸). We note that for simple and short IDP model peptides near physiological conditions $k_o \ll k_{CW}$ (as in our previous works) and the observed relaxation rate can be approximated by k_{CW} . For larger and more complex IDPs, such as N_{TAIL}, this is not necessarily the case. On one hand k_{CW} may be significantly smaller (slower) in larger and more complex IDPs, due to larger C-W sequence separations, and to intrinsically slower intra-chain dynamics caused by interactions; on the other, the relaxation rates k_o may be larger (faster), depending on pH, temperature, additional quenchers, and the position of W in the sequence. Therefore, k_o should be determined on variant containing W and no C, under the given solution conditions. Throughout this work, C-W quenching times (plotted in main text Fig 1D) were obtained as:

$$\tau_{CW}^{-1} = \tau_{obs}^{-1} - \tau_o^{-1}$$

PET sample preparation

Samples for photo-induced electron transfer (PET) measurements were prepared by dissolving the lyophilized protein directly into filtered buffers (for reference conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP pH 7.6; for low pH measurements: 20 mM NaAc, 150 mM NaCl, 1 mM TCEP pH 4.0; for high salt concentrations¹: 15 mM Tris, 500 mM NaCl, 1mM TCEP pH 7.6). Protein concentration was adjusted to ~100 μ M as determined by absorbance at 280 nm (extinction coefficient 6990 $\text{cm}^{-1} \text{M}^{-1}$). Samples (~350 μ L) were placed in 5x10x30(h) mm Spectrocell quartz gas tight cuvettes with screwcap (equipped with Teflon coated silicon membrane), and bubbled with USP grade nitrous oxide for at least one hour, to reduce the concentration of dissolved oxygen (a quencher of excited tryptophan triplet state), and to introduce solvated electron scavenger (to prevent reactions with electrons, produced by water decomposition under UV laser pulses).

PET experimental setup

We measured the lifetime of the excited W triplet state after UV pulse excitation, using a homebuilt nanosecond transient absorption apparatus. Our apparatus uses two coherent light sources, a 10 Hertz pulsed 289 nm excitation beam, and a 457.9 nm continuous probe beam, arranged in a pump-probe optical setup, similar to that in a previous work⁸. The 3 mW excitation pulse train is provided by the fourth harmonic of a Surelight Nd:YAG laser (266nm), shifted to 289 nm using a deuterium gas Raman Converter (Light Age). The excitation, Stokes, and anti-Stokes components from the Raman Converter are separated by a Pellin-Broca prism. The 289nm component is selected via a spatial pinhole filter, reduced to ~ 3.5 mm diameter, and eventually blocked by an aperture after passing through the 1 cm long and 5 mm wide sample cell equipped with a magnetic stirrer. The 200 mW, 457.9 nm Argon Ion laser beam (Coherent Innova 90C) is split into two beams. One beam, used as the probe light source, is directed to pass into the sample cell collinearity (with small angle offset) with the excitation light beam, and the other is used as the reference light source. Optical probe and reference signals are each filtered using interference filters and focused onto two separate diode detectors (New Focus, Model 2032). The outputs from each detector are directly recorded by two digital storage oscilloscopes (DPO 3032, Tektronix) to simultaneously record kinetics in both short (800 ns; with 150 MHz bandwidth, 1.25 GS/s sampling rate) and long (1 ms; with 20 MHz bandwidth, 1.00 MS/s) timescales, as to maintain high time resolution and high dynamical range. High speed shutters are employed to control the probe and excitation laser beams prior to entering the sample to protect it against unnecessary radiation exposure between measurements and to obtain background noise data.

¹ The pH was adjusted by adding amounts of acetic acid (this could be significantly different from the one estimated by weight).

With each sample, multiple measurements, each averaged over 256 laser pulses, were taken. Digital data from oscilloscopes were initially processed to reduce background noise, produce transient absorption values, and to combine the short and long timescale data using in house MATLAB (MatchWorks) scripts. Measurements, noise reduction, and initial data processing were automated using LabView. An example of the resulting raw data, monitoring the population of the W triplet state as it decays in time, is shown in S1 (all PET data presented in this manuscript can be found in Supporting Figs. S3 through S16).

Transient absorption data analysis

Relaxation times, τ_{obs} , were obtained by fitting transient absorption data (Fig. S1) using locally written software (ASUFIT) [<https://www.public.asu.edu/~laserweb/asufit/asufit.html>] developed in a MATLAB environment. Multiple measurements (from multiple sample preparations as well as from repeated measurements of the same sample), taken under given solution conditions, and for a given N_{TAIL} variant, were globally fit to a sum of exponentials including a rise time (Eq 1) convoluted with the Gaussian function (FWHM ~4 ns) associated with the excitation pulse time profile:

$$\Delta A(j, t) = \sum_{i=1}^n A_i(j) \exp(-t/\tau_i) \quad (1)$$

Where $\Delta A(j, t)$ is the transient absorption signal for an individual measurement j as a function of time t and n is the number of kinetic components used in the fitting ($n=5$, including a rise time, delivered the best fit), $A_i(j)$ is the amplitude of the i -th kinetic component for the given measurement j , and τ_i is the global fitting parameter representing the lifetime of the i -th kinetic component, optimized over all repeated measurements j (these included multiple measurements of the same sample, as well as measurements from multiple sample preparations). An example of raw data obtained from PET measurements, highlighting the contribution to the fitting function from individual kinetic components can be found in Supporting Fig. S1. The kinetic component with the shortest lifetime (τ_1 , orange) has the largest amplitude A_1 (i.e., is the main component of the signal) and corresponds to the intact protein population where the excited W triplet state is quenched by electron transfer to C. The following components, τ_2 (green), τ_3 and τ_4 (not shown) have much smaller amplitude and correspond to contributions from photochemical decomposition products (as explained in the following). Kinetic component τ_5 (~20-50 ns) has negative amplitude A_5 and represents the initial rise in W triplet state population (formation of the lowest triplet excited state). All fitting parameters for each variant, under each solvent condition, are reported in Supporting Table S2. The corresponding raw data and fits can be seen in Supporting Fig. S3 through S16.

To understand the nature of components τ_2 , τ_3 , and τ_4 , we note that under intensive UV light, water decomposes and produces free solvated electrons and various chemically reactive radicals^{10,11}. Electrons can reduce C, thus preventing electron transfer from W to C, and radicals can damage W and other parts of the protein. The effects of these photochemical reactions on the transient absorption signal are evident in Supporting Fig. S2. The total amplitude (the sum of all amplitudes associated with each kinetic component) decreases with each measurement (not shown). This is expected and indicates the chemical destruction of some W after long intense UV light exposure (this is more evident when small sample volumes are used, as in the present study). The amplitudes of each individual component (shown in Supporting Fig. S2) reveal which one is due to photodamage, and which is due to the actual quenching process between intact W and C. The figure shows that the largest amplitude, corresponding to the protein population where W is quenched by C (τ_1 blue), decreases steadily as a function of UV laser irradiation, due to gradual depletion of intact W and C pairs. The next significant component (τ_2 , orange) starts at much lower amplitude and increases steadily, indicating that this decay results from a photodamage. The remaining two components (τ_3 , and τ_4 , green and

red), have negligible amplitudes and increase slightly as a function of UV irradiation, indicating that these too result from photodamage.

To avoid including measurements that were dominated by photodamage, in our global fits we only included data for which the amplitude A_i of the main decay (τ_i) contributed 35% or more to the total signal amplitude $\Delta A(j,0)$ (Eq (1)). This cutoff was chosen to allow enough measurements for sufficient statistics, while avoiding unwanted effects due to photodamage. All transient absorption kinetics raw data with their fits are plotted in Supporting Figs. S3 through S16. The corresponding fitting parameters are reported in Supporting Table S2. Finally, τ_{CW} , the actual contribution to the observed relaxation time due to quenching by C (as opposed to the contribution from natural relaxation), was obtained as explained above from:

$$\tau_{CW}^{-1} = \tau_{obs}^{-1} - \tau_o^{-1},$$

where $\tau_{obs}^{-1} = \tau_1^{-1}$ measured for the variant containing both W and C, and $\tau_o^{-1} = \tau_1^{-1}$ measured for the corresponding reference variant, containing only W, under given solution conditions. Values of τ_{CW} are reported in Table S2 and in the main text Fig. 1D. Only values for τ_{CW} (corresponding to relaxation due to C-W electron transfer) are reported and discussed in the main text. Errors on the observed times (obtained by globally fitting multiple transient absorption signals, obtained from repeated measurements) were estimated from block averaging on mutants with the most measurements. Namely, repeated measurements were divided into groups of different bin size and globally fit within each bin. The standard error of the mean of the main decay time was calculated within each group, and plotted as a function of bin size. The plateau value yielded an error of about $\pm 5\%$ on the observed times obtained from global fits.

Bimolecular quenching experiments

To check whether the differences in C-W relaxation times τ_{CW} , observed between pH 4.0 and pH 7.6 (shown in Fig. 1D main text), were due to actual conformational changes of the protein as opposed to changes in the electron transfer efficiency between W and C, we measured the bimolecular quenching rates of W and C freely diffusing in solution, at the two pH values. For this purpose, we carried out a series of PET measurements on 100 μ M N-Acetyl Tryptophan Amide (NATA) solutions, containing varying cysteine concentrations, at both pH 4.0 and pH 7.6. The buffers were identical to those used for N_{TAIL} in Fig. 1D (20 mM NaAc, 150 mM NaCl, pH 4.0 and 15 mM Tris, 150 mM NaCl pH 7.6). The smaller cysteine concentration range used at pH 7.6 (0 mM – 300 mM) compared to pH 4.0 (2 mM - 10 mM), was due to the lower solubility of cysteine at pH 7.6. Cysteine concentrations were determined using Ellman's reagent protocol. All sample preparation was carried out under oxygen free atmosphere in a glovebox, to avoid disulfide formation. All other details for PET experiments and data analysis were identical to those described in the above sections. The corresponding Stern-Vollmer plots of the observed relaxation rates (from the largest component of the multiexponential fit) measured at 20°C as a function of cysteine concentration are shown in Supporting Fig. S19. The slopes of the linear fits yield a bimolecular quenching rate of $1.97 (\pm 0.20) \times 10^8 (\text{M}\cdot\text{s})^{-1}$ at pH 7.6 and $1.67 (\pm 0.17) \times 10^8 (\text{M}\cdot\text{s})^{-1}$ at pH 4.0, which are the same within experimental error (error calculated by propagating the $\pm 5\%$ error on k_{obs} to $\pm 10\%$ error on the slopes).

S3: Circular dichroism measurements

The CD spectra of Fig. 1D inset were measured using a Jasco J-815 spectropolarimeter, in a 1mm quartz cuvette (Starna Cells), a wavelength range 190-250 nm, with a 0.4 nm step resolution, 50 nm/min speed, 2 s response time, and 1 nm bandwidth. Each sample spectrum was obtained by averaging eight spectra. Solvent conditions were 10mM NaPO₄, 150 mM NaF for pH 7.6 measurements, and 20 mM NaAc 150 mM NaF for pH 4.0 measurements. Protein

stock solution concentration was calculated from UV-VIS absorption spectrum. To ensure high quality data at low wavelengths, where spectral shape is most significant for IDPs, we carried out measurements on samples at multiple protein concentrations, obtained by stock solution dilution (9.5 uM, 7.0 uM, 5.0 uM at pH 7.4, and 8.2 uM, 6.0 uM at pH 4.0). We used these to check measurement consistency, and signal intensity optimization at different wavelength (requiring linear concentration dependence at all wavelengths included). The data obtained then were buffer subtracted and baseline corrected, spectrally truncated according to maximum high tension (HT) voltage limitation, renormalized based on concentration, and averaged together to extend spectral range and reduce the overall signal-to-noise ratio.

S4: Scaling of PET rates using a polymer model.

In order to understand the scaling behavior of C-W quenching rates, τ_{CW}^{-1} , measured in PET experiments, as a function of sequence separation ($|i-j|$), we utilized a polymer model, SAW- ν model¹², in which the distance distribution function $p(r, \nu)$ is written as

$$p(r, R, \nu) = A \frac{4\pi}{R} \left(\frac{r}{R}\right)^{2+g} \exp[-\alpha \left(\frac{r}{R}\right)^\delta].$$

Here ν is a varying scaling parameter, $g = (\gamma - 1)/\nu$, $\gamma \approx 1.1615$, $\delta = 1/(1 - \nu)$,¹³⁻¹⁵ the constant A and α are determined by normalizing the integration of $\int p(r)dr$ and $\int r^2 p(r)dr$ to 1, and R is the RMS distance. We then calculated for a polymer chain, using the SAW- ν model, the expected C-W quenching rates for a range of sequence separation values as

$$1/\tau_{CW} = \int k_q(r) p(r, R, \nu) dr,$$

in which

$$k_q(r) = k_0 H(r - r_0),$$

a step function with $r_0 = 0.4 \text{ nm}$ ¹⁶. This is the estimated reaction limited rate contributing to the relaxation times exclusively due to chain compaction/expansion, and the diffusion limited rate due to dynamics of the chain is not considered. We note that k_0 is not important here, since it will cancel out when investigating the scaling behavior or ratio of rates. For a given Flory scaling exponent ν , the RMS distance is estimated using the scaling law

$$R_{i,j} = b|i - j|^\nu$$

with $b=0.55\text{nm}$ ¹⁷. By fitting the theoretically calculated values of τ_{CW} above, as a function of $|i-j|$ using

$$\tau_{CW} \propto |i - j|^{\nu_\tau}$$

we obtain the scaling exponent ν_τ for the τ_{WC} , corresponding to a given Flory scaling exponent ν , for R . As shown in Supporting Fig. S17, we find a linear relation between the distance scaling exponent ν and quenching time scaling exponent ν_τ . For the scaling exponent of ~ 0.52 estimated using the radius of gyration from SAXS experiment¹⁸ and the SAW- ν model¹², we obtained a corresponding scaling exponent of ~ 1.69 for C-W quenching time. Using this value, we fit the experimental values τ_{CW} at pH 7.6 in Fig 1D (main text), and obtain a best fit value of $0.011\mu\text{s}$ for the prefactor.

S5: Coarse-grained models for N_{TAIL}

We applied the coarse-grained HPS model to N_{TAIL} . In the original HPS model¹⁹, each amino acid is represented by a bead with charge (+1, 0, -1) and hydrophathy. There are three types of interactions: bonded interactions, electrostatic interactions and short-range pairwise

interactions. The bonded interactions are characterized by a harmonic potential with a spring constant of 10 kJ/\AA^2 and a bond length of 3.8 \AA . The electrostatic interactions are modeled using a Coulombic term with Debye-Hückel²⁰ electrostatic screening to account for variation of salt concentrations. Specifically, the strength of this screening effect decays exponentially with distance, and the range is governed by the Debye length, which decreases with increasing salt concentration. The short-range pairwise potential accounts for both protein-protein and protein-solvent interactions with an adjustable parameter epsilon for the interaction strength, which was optimized using the experimental radius of gyration of IDPs and can be tuned specifically for N_{TAIL} based on its radius of gyration.

Here for N_{TAIL} , we further added additional terms for angle and dihedral preference so that the secondary structure of the MoRE region can be captured. Specifically, for the angle potential, we used a statistical potential from a previous literature²¹ for all types of amino acids. For the dihedral potential, we used a statistical dihedral potential from a previous literature²², in which the parameters are dependent on the types of the two centering amino acids. We found that even with the statistical angle and dihedral potentials, the helical propensity of the MoRE region is still underestimated (Supporting Fig. S20). Therefore, we further introduced a harmonic dihedral potential

$$E_{\text{harmonic-dih}} = k_{\text{dih}} [1 + \cos(\theta - 28)],$$

just to the amino acids in the MoRE region (residues 489 to 499), Here k is scanned in a variety of values (Supporting Fig. S20) to also test the role of secondary structure of MoRE region to the dynamics of N_{TAIL} . We determined $k_{\text{dih}}=0.2 \text{ kcal/mol}$ so that the secondary structure of the MoRE region is close to that from NMR measurement²³. With these angle and dihedral terms, we further scanned the interaction strength ϵ of the pairwise interaction term and found that $\epsilon=0.16 \text{ kcal/mol}$ best captures the radius of gyration of N_{TAIL} .

The C-W quenching rate is calculated using the same step function as shown in Supporting Methods S4,

$$k_q(r) = k_0 H(r - r_0),$$

in which $r_0 = 0.4 \text{ nm}$ but now averaging over all the conformations in the simulation. We do not need an explicit k_0 here, since it will cancel out when investigating the scaling behavior or ratio of rates.

In order to simulate N_{TAIL} at a low pH, we calculated the average charge of D and E based on their pKa values and changed the charge of these two amino acids in the CG model accordingly. For example, the pKa values of D and E are 3.65 and 4.25, respectively. At pH 4.0, the averaging charge of D and E are therefore -0.69 and -0.36, respectively.

The HOOMD-Blue software v2.9.3²⁴ together with the azplugins (<https://github.com/mpowardlab/azplugins>) were used for running the molecular dynamics simulations. All simulations were run using a Langevin thermostat with a friction coefficient of 0.01 ps^{-1} , a time step of 10fs and a temperature of 298 K for 5 μs . The first 0.5 μs were dumped for equilibration. Further simulation conditions and lengths are provided in Table S9.

To estimate the contribution to the C-W relaxation time, exclusively due to changes in the C-W distance from the CG simulation, we first obtained the distances between the two measured amino acids (i.e. W and C) from the simulation, then we calculated the distance dependent quenching rate $k_q(r)$ for each conformation in the trajectory as $k_q(r) = k_0 H(r - r_0)$ (as described above), and at last estimated the τ_{WC} as the inverse average of $k_q(r)$ from all

conformations (i.e. $\langle k_q(r) \rangle^{-1}$). We note that for an all-atom model, the same distance dependent quenching rate $k_q(r) = k_0 H(r - r_0)$ as previously used in the polymer model can be used with a characteristic distance r_0 of 4 Å¹⁶. For a CG model, both r_0 and k_0 are unknown. The cutoff distance r_0 is expected to be larger than an all-atom model considering the missing atomistic interactions. We used an r_0 of 10 Å for all the analysis. Regarding k_0 , we always compared the ratio of the PET rates instead of an individual PET rate so determining k_0 for the CG model is not necessary.

S6: All-atom Molecular Dynamics Simulations

Molecular dynamics (MD) simulations on three Measles Nucleocapsid protein C-terminal domain (N_{TAIL}) variants (WT, C3W2, C4W2) were performed using the GROMACS 2019 package²⁵ under periodic boundary conditions, constant temperature and pressure.

S7: General simulation setup

The temperature of 298 K was maintained separately for solute (protein) and solvent (water and ions) atoms using a velocity rescaling algorithm²⁶ with a coupling constant of 0.1 ps for both atom groups. For constant pressure simulations a Parinello-Rahman barostat²⁷ was used to maintain a pressure of 101 kPa using a 0.1 ps coupling constant and the isothermal compressibility of water 4.5 10⁻⁵ bar⁻¹. All simulations were propagated using a leapfrog integrator²⁸ in 4 fs time steps. Fast vibrational degrees of freedom were removed by using LINCS algorithm²⁹ using a sixth order iterative restraint on the bond angles. Apolar hydrogen positions were constrained using virtual atom sites to eliminate hydrogen bond vibrations. Electrostatic and van der Waals interactions were explicitly calculated within 1.0 nm cutoff distance, long-range electrostatic interactions beyond the cutoff were calculated by particle-mesh Ewald summation³⁰ with a grid spacing of 0.12 nm. Long-range van der Waals dispersion corrections³¹ to the total energy of the system were applied in all simulations.

S8: Force field parameters

Initial simulations of WT N_{TAIL} were carried out using multiple simulation parameter sets of the and Amber force field families. The tested parameter sets included two modified AMBER99 force fields, one with a rescaled TIP4P water model^{32,33}, and a parametrized version by Robustelli *et al*³⁴, AMBER03 with a scaled TIP4P water model³², the CHARMM22* force field³⁵ either using a modified TIP3P water model³⁶ or a TIP4P water model³⁷, and the CHARMM36m³⁸ force field with either a TIP3P, TIP4P, or OPC³⁹ water parameters. Production MD simulations of all three N_{TAIL} variants were performed using the CHARMM36m protein parameter set with OPC water parameters (C36M-OPC), and the AMBER99SB-disp (A99SB-d) parameter set including modified TIP4P water parameters optimized for intrinsically disordered protein simulations by Robustelli *et al*³⁴.

S9: System preparation

Three wild type N_{TAIL} (aa. 399-525, with an N-terminal His6-tag) conformations including a helical alpha-More region were selected from a structural ensemble deposited in the protein ensemble databank⁴⁰ (PED-ID: 00020), and three further extended conformations were used to start all production runs. All conformations were solvated in a dodecahedral box of approximately 12 nm unit cell size. The box was filled with water molecules, Na⁺ and Cl⁻ ions appropriate for a neutral 150 mM salt solution. All amino acids were protonated according to their dominant protonation state at pH = 7.2.

Solvent molecules around the protein were energy minimized, then initial velocities to all atoms in the system were assigned according to a Maxwell-Boltzmann distribution appropriate for 298 K temperature. The systems were then allowed to equilibrate under constant temperature and volume (NVT) using a Berendsen⁴¹ thermostat for 100 ps, where the solute atomic positions were restrained by a harmonic potential and a force constant of 2500 kJ/mol/nm. In a second

equilibration step, 100 ps MD simulation performed on all systems under constant temperature and pressure (*NPT*) using a velocity rescaling thermostat and a Parinello-Rahman barostat as described in the general simulation setup but maintaining the protein harmonic restraints. Finally, a 250 ps *NPT* MD simulation was performed on all systems with position restraints, and the atomic positions and velocities of the systems were used to start production MD simulations.

S10: Force field comparison

The behavior of intrinsically disordered proteins in all-atom MD simulations depends strongly on the choice of the parameter set used (force field)⁴². Additionally, MD force field performance for IDPs often varies in a system-dependent manner⁴³. We therefore chose to validate several force fields and water models against existing independent experimental data on the wild type N_{TAIL} sequence. The two best performing parameter sets were then selected to perform longer production runs of the two N_{TAIL} variants used in our PET experiments. The validation data includes information on protein compactness based on small angle X-ray scattering (SAXS), secondary structure information based on circular dichroism (CD) and nuclear magnetic resonance (NMR) chemical shifts. Supporting Table S3 summarizes the computed deviations between measured observables and those predicted from MD simulations using various force fields of the AMBER and CHARMM families. For reference, the table also shows the deviations of a molecular ensemble of N_{TAIL} which was obtained by fitting to NMR chemical shifts, and was deposited in the protein ensemble databank (PED).

A more detailed discussion of force field accuracy is provided below, but in summary: two all-atom force fields reproduced both N_{TAIL} compactness and secondary structure propensities accurately. The most accurate simulations were obtained with the C36M force field³⁸ combined with optimal point charge (OPC) water³⁹ parameters. The second-best accuracy was achieved by the A99SB-d force field. For validation of this AMBER force field variant, we used previously published wild type N_{TAIL} simulation trajectories³⁴, and force field parameters adapted to the GROMACS simulation package, both kindly provided by Piana et al.

We therefore selected the C36M-OPC and A99SB-d force field variants to generate a new set of simulations to sample the dynamics of two N_{TAIL} variants. These variants, S488C-Y518W (C4W2) and S482C-Y518W (C3W2), displayed considerable dynamic heterogeneity in PET experiments (*i.e.* they deviated from homopolymer scaling behavior in Fig. 1D main text), and we sought to understand the structural causes of such heterogeneity through MD simulations. A summary of observables estimated from the variant simulations is shown in Supporting Table S3.

In both force fields the average radius of gyration for variants is smaller (2.5 and 2.2 nm) compared to the WT (3.0 and 2.7 nm). This is most likely due to the presence of a hexa-histidine (His6) purification tag at the N-terminus of the WT N_{TAIL} sequence, which was absent in the variant sequences used in all experiments and MD simulations. Unfortunately, no SAXS measurements of the C4W2 or C3W2 sequences were available at the time of writing to confirm or disprove this hypothesis.

However, we performed CD measurements on a similar variant, Y518W (W2), without a His6 tag to estimate the expected change in secondary structure. Bayesian secondary structure estimates using the SESCO analysis package⁴⁴ indicated a significant reduction in beta-strand content (from 13 ± 6 % to 2 ± 6 %), even after compensating for side chain contributions of the tag. As shown in Supporting Fig. S21, the C36M-OPC simulations of both WT and variant were consistent with the measured CD spectra, whereas the A99SB-d simulations reported a significantly higher helix content (12% vs. 5 ± 5 %) than the CD estimates.

Further, as explained in detail below, we also computed the predicted relaxation curves for the excited tryptophan triplet state population (measured in PET experiments) from simulations with both force fields, according to the method established by Zerze et al¹⁶. The predicted relaxation curves, shown in Supporting Fig. S22, suggest a good qualitative agreement with those measured in PET experiments. The rather large uncertainty of predicted curves is due to sparse sampling of the electron donor-acceptor (C-W) contact conformation on the timescale of the performed MD simulations.

In conclusion, our unbiased MD simulations using the C36M-OPC parameter set reproduce the experimental observables, within uncertainty, for both WT N_{TAIL} and the selected N_{TAIL} variants used in PET experiments. In fact, the deviations between experimental observables and those estimated from the WT N_{TAIL} simulations are on par with the N_{TAIL} ensemble model, deposited in the protein ensemble databank (PED.) The A99SB-d simulations also predict the wild type observables accurately but seem to overestimate the helicity of the variants. Therefore, we decided to investigate the structural reasons for the dynamic heterogeneity observed in PET experiments, based on the C36M-OPC simulations.

S11: Comparison between all-atom molecular dynamics simulations and experimental data

To assess the accuracy of all MD simulations, CD spectra, SAXS curves, and NMR chemical shift profiles were predicted from MD simulation trajectories conformation by conformation, then the ensemble averaged observables were compared to the measured experimental data as described below.

S12: CD spectrum predictions and comparison

Circular dichroism spectra were predicted from 190 to 250 nm using the SESCO package version 0.95⁴⁵. The predicted CD spectra were computed using the DS-dTSC3 basis set, which includes three secondary structure (SS) signals (helix, beta-strand and other), and base-line corrections for side chain contributions. The SS composition W_i of the protein in each conformation of simulated MD trajectories was determined by the DISICL algorithm and averaged over all trajectories of a given force field. The predicted CD spectrum is computed as

$$S_{\lambda}^{calc} = \sum_i W_i \cdot B_{i\lambda},$$

where $B_{i\lambda}$ is the CD signal of SS element i at wavelength λ . Side chain corrections for the protein are calculated in a similar manner, based on the extracted mean CD signal of amino acid side chains ($B_{k\lambda}$), weighted by fraction of amino acid k in the sequence of the protein (W_k).

The root means squared deviation (RMSD) of the predicted and measured CD intensities in 1000 mean residue ellipticity (kMRE) units were calculated and used as a metric to validate the MD trajectories. The global intensity of the measured CD spectrum was automatically rescaled by the factor α to minimize this RMSD and eliminate potential errors in the estimated experimental protein concentration (as described in a previous work⁴⁵)

$$RMSD = \sqrt{\frac{\sum_{\lambda}^L (S_{\lambda}^{calc} - \alpha \cdot S_{\lambda}^{exp})^2}{L}}.$$

The accuracy of MD simulations in terms of N_{TAIL} SS composition was determined by comparing the RMSD between predicted and measured CD spectra to the average deviations obtained for a set of eight disordered proteins with known CD spectra and validated structural ensembles⁴⁶. The mean RMSD of these model IDP-s was 2.2 kMRE with a standard deviation of 1.1 kMRE. Based on these values, the CD predictions were considered exceptionally good if they were below 1.1 kMRE (mean-sigma) and of acceptable accuracy below 3.3 kMRE (mean+sigma).

We note that four of the IDP models were taken from the protein ensemble databank (PED)⁴⁷, including a deposited model of WT N_{TAIL}. The mean RMSD of the IDP set is close to the 2.07 kMRE observed for globular proteins using the same parameter set, with a similar standard deviation of 1.0 kMRE.

The CD-based validation of N_{TAIL} simulation ensembles relied on two CD spectra; a previously measured spectrum of WT N_{TAIL} with a hexa-histidine (His6) tag attached, kindly provided by Longhi *et al*, and a more recent CD measurement of the N_{TAIL} variant Y518W (without a His6 tag) measured in this work, as described in Supporting Methods S3 (pH=7.6 phosphate buffer.).

In Figure S21 we compare our MD simulation ensembles of the C36M-OPC and A99SB-d force fields to the measured CD spectra. Figure S21-A shows that for the WT N_{TAIL} both force fields predict the CD spectrum with similarly good accuracy (1.4 and 1.5 kMRE, respectively). In contrast, the predicted CD signal of C36M-OPC variant simulations in Supporting Fig. S21B show a much better agreement with the measured spectrum (0.9 kMRE) than those computed from A99SB-d simulations (3.3 kMRE). This difference is attributed to the differences between measured CD spectra themselves, and the significantly higher α -helix content in Amber-based simulations. To investigate further, we estimated the SS composition of N_{TAIL} variants based on their measured CD spectra (as discussed in Supporting Methods S13), and compared these estimates with SS composition obtained from all-atom MD simulations.

S13: Secondary structure predictions

The secondary structure (SS) of N_{TAIL} was estimated based on the measured CD spectra described in using a bayesian SS estimation algorithm⁴⁴ of the SESCO package version 0.95 and the CD basis set DS-dTSC3. To estimate most likely SS composition, the possible SS composition space is sampled by 50 Monte Carlo - Hastings chains started from random initial SS compositions, yielding a total 100 000 sampled SS compositions, from which the first 20 000 were discarded to exclude very unlikely initial SS. The posterior probability of sampled SS compositions $P(SS_j | CD_\lambda)$ given the measured CD spectrum is determined according to

$$P(SS_j | CD_\lambda) = P(CD_\lambda | SS_j) \cdot P(SS_j) ,$$

where $P(SS_j)$ is prior probability of SS compositions j , taken from a uniform distribution, and $P(CD_\lambda | SS_j)$ is the likelihood of the measured CD spectrum given that SS composition. This likelihood is the joint probability of the deviation $RMSD_j$ and scaling factor α_j required to compare the measured CD spectrum and the predicted spectrum of the SS composition j , as described in Supporting Methods S12. The likelihood was estimated previously⁴⁴ from approximately 300 SS predictions of globular proteins with known CD spectra and structures.

Using the SESCO basis set DS-dTSC3, the expected SS composition of was determined from the CD spectra measured for the WT N_{TAIL} with a His6 tag, as well as for the N_{TAIL} variant Y518W (W2) without a tag (shown in Supporting Fig. S21). This basis set provides estimates for the protein's helix, β -strand, and random coil contents, while applying baseline corrections for amino acid side chain CD signals.

SS estimation of the measured CD spectra shows that both N_{TAIL} variants remain disordered, but whereas the N_{TAIL} Y518W (W2) variant shows only a small amount periodic SS (1 ± 3 % helix and 2 ± 6 % β -strand) content, the WT variant likely has a higher propensity for β -strands (14 ± 6 %). The difference in estimated β -strand content suggests that the N-terminal His6 tag may have significant impact on the overall N_{TAIL} structure, and the experimental data available for the WT N_{TAIL} should not be used to validate our variant simulations. The estimated SS composition also defines a maximum helical propensity of the MORE region, which is approximately 58% for the WT N_{TAIL}, and 32% for the variant.

The estimated SS compositions were also compared to the SS compositions observed in MD simulations for both the WT and variant N_{TAIL} variants (in Supporting Tables S3 and S4). The comparison reveals that the unexpectedly large deviation between the measured CD spectrum and the predicted spectrum of the N_{TAIL} variants based on Amber99SB-disp simulations is due to an over-estimation of α -helix content (11% vs. 3 \pm 4%).

S14: SAXS predictions and comparison

Small angle X-ray scattering between 0.01 and 0.05 1/Å was calculated using the program CRY SOL (from ATSAS version 2.7.2.5)⁴⁸. SAXS curves from each conformation of simulation ensembles were averaged and compared to the measured SAXS curve using a modified χ^2 value, calculated as

$$\chi^2 = \frac{1}{N} \sum_q \left(\frac{I_q^{exp} - \alpha \cdot I_q^{calc}}{\sigma_q^{exp}} \right)^2 .$$

Here I_q^{exp} and I_q^{calc} are the measured and calculated SAXS intensities, and σ_q^{exp} is the experimental uncertainty of the intensity at scattering vector q , respectively. The scaling factor α rescales the predicted SAXS curve to match the experimental intensity, which is expressed in arbitrary units and depends on the experimental setup. To this aim, the χ^2 deviation is minimized as a function of α , and this minimized value quantifies the deviation between the predicted and measured SAXS curves. Because χ describes the average deviation in relation to the experimental uncertainty, simulation ensembles were considered of good accuracy if χ is below 1.0 and still acceptable when χ is between 1.0 and 2.0. The SAXS-based evaluation of simulation ensembles relies on previously measured SAXS curve of the WT N_{TAIL} provided by Longhi *et al*¹⁸.

S15: NMR chemical shift predictions and comparison

C α chemical shifts for each residue in the protein was calculated for each conformation using the SPARTA+ software (version 2.6)⁴⁹. The calculated primary chemical shifts were averaged for each C α atom, and compared to measured chemical shifts using the equation

$$RMSD = \sqrt{\sum_j^N \frac{(S_j^{calc} - S_j^{exp})^2}{N}}.$$

Here, S_j^{calc} and S_j^{exp} are the predicted and measured values of chemical shift j , respectively. To analyze the data, the predicted primary chemical shifts were also converted to secondary chemical shifts. To this aim, the difference of primary and secondary chemical shifts of the predicted data were subtracted from the measured primary chemical shifts. This procedure applies the same chemical shift corrections for random coil values for both measured and computed shifts.

For the C α chemical shift-based evaluation of N_{TAIL} simulation ensembles, chemical shifts provided by Blackledge *et al* for the PED entry. This entry was later updated into PED entry 00020. The chemical shifts are identical to those published by Gely *et al*.⁵⁰ The agreement with these C α chemical shifts was considered good if the RMSD between the measured and predicted chemical shifts is smaller than the typical error of the chemical shift prediction for IDP ensembles 0.48 ppm with a standard deviation of 0.27.⁴⁶ Based on these results we considered exceptional accuracy for C α chemical shift predictions and below 0.21 ppm of acceptable accuracy below 0.75 ppm. Similarly, we set the accuracy for C β chemical shifts based on the typical error of 0.52 ppm and SD of 0.1 ppm, with exceptionally good predictions below an RMSD of 0.42 ppm and acceptable accuracy below 0.62 ppm.

S16: Triplet-state survival predictions and comparison to PET experiments

Relaxation curves for the excited tryptophan triplet-state population, due to electron transfer between W and C, were calculated from MD trajectories according to Zerze et al¹⁶. The C-W relaxation times due to electron transfer τ_{WC} , can be estimated from the survival probability curve $S(t)$ of the tryptophan triplet state.

The tryptophan triplet-state survival curves are calculated from the minimum donor-acceptor distances r_{CW} , which in turn are computed from atomic coordinates at every timestep of the MD simulation as $r_{CW} = \min(|\underline{R}_{CY} - \underline{R}_{Wi}|)$. Where \underline{R}_{Wi} are the positions of tryptophan ring atoms (electron donor) and \underline{R}_{CY} is the S_γ atom position of the cysteine (acceptor). Note that in order to increase sampling efficacy, the donor-acceptor distances were also estimated using the tryptophan ring atom positions and the O_γ atom position of serines in alternative acceptor positions. This allowed us to determine both C3W2 and C4W2 distances for all variant simulation trajectories, regardless of which serine-to-cysteine mutation was actually applied to N_{TAIL} in the trajectory.

The survival probability curve of the triplet-state tryptophan, is calculated as

$$S(t) = \langle e^{-\int_{t_0}^{t_0+t} q_c \cdot H[r_{CW}(t') - r_c] dt'} \rangle_{t_0},$$

where the donor-acceptor distance r_{CW} , $q_c = 8 \cdot 10^{-8} s^{-1}$, and $r_c = 0.4$ nm are the parameters that determine the probability of quenching upon contact, and H denotes a Heaviside step function that describes the number of contact conformations encountered within a time lag t after the initial conformation at time t_0 . A conformation is considered to be in donor-acceptor contact if the r_{CW} distance is smaller than the contact cutoff distance r_c . To compute the predicted survival probability curves of N_{TAIL}, this calculation was averaged for every t_0 timestep where the integral until $t_0 + t$ can be fully evaluated within any one of the $N = 6$ simulation trajectories performed using the same force field parameters. A bootstrapping algorithm was used to assess the uncertainties of $S(t)$ by drawing 1000 sets of N trajectories (with replacements) and computing the predicted survival curves. The expected value and uncertainty of $S(t)$ at each t was estimated by computing the mean and standard deviation of the predicted values from each set drawn. This approach allowed us to obtain $S(t)$ curves up to 2.5 μs shown in Supporting Fig. S22 for both the C3W2 and C4W2 N_{TAIL} variants. These curves were compared directly to the tryptophan triplet state relaxation curves measured in PET experiments (Supporting Methods S2) and were used to estimate the timescales of the dynamic processes in MD simulations that would influence measured transient absorptions due to C-W quenching.

For a direct comparison with the relaxation curves computed from the MD simulations, the primary data (transient absorption ΔA vs. t) measured in PET experiments, were transformed into survival probability curves (S vs. t). To this aim, the initial data points before $t_{ex} = 360$ ns of the measurements were discarded. During this period, tryptophan molecules in the sample are excited, and ΔA undergoes a rapid increase to a plateau. The remaining ΔA values are transformed into survival probabilities by the equation $S(t'') = \Delta A(t) / \Delta A_{ex}$, where ΔA_{ex} is the average ΔA of the plateau, and $t'' = t - t_{ex}$ is the adjusted time after reaching maximum excitation. This transformation is necessary, because the initial excitation depends on the experimental setup, and MD simulations can only provide information on the relaxation decay after maximum excitation is reached. Finally, we note that (as explained in Supporting Methods S2, and further discussed below), the W triplet-state relaxation curves measured in PET experiments are the result of two types of relaxation processes: one due to C-W quenching from electron transfer to C (computed in MD), and one due to all other processes occurring naturally in solution in the absence C (not computed in MD).

S17: Analysis of the predicted relaxation curves

All MD-based survival curves shown in Fig. S22 panels A and B indicate several dynamic processes, occurring on multiple timescales, associated with C-W quenching upon contact. These processes can be explained using the free-energy landscape shown in Fig. 3 (main text). The first of these processes for both variants (in both force fields) are fast processes on the 10 ns timescale that affect only a small fraction of the excited N_{TAIL} population. These processes are most likely associated with side chain rearrangements within the contact conformational states (S4 for C3W2 and S3 for C4W2, respectively), which bring the electron donor and acceptor in and out of contact without major structural changes of N_{TAIL} . Unfortunately, these processes are too fast to be captured by PET experiments, because they would occur during the triplet-state excitation process by the laser pulse.

The second fastest process is likely associated with crossing the first free-energy barrier between the free-energy minimum in the S1 conformational state and the respective contact state (S3 or S4). Time scales for these processes were determined by fitting the initial part of the survival probability curve (shown as process 2 on Supporting Fig. S22) with a single exponential function $S(t) = e^{-t/\tau_{CW}}$. We applied a Bayesian fitting method that takes into account the uncertainty of $S(t)$ to provide a mean and uncertainty estimate of the average quenching time τ_{CW} .

In addition, predicted triplet-state survival curves of both force fields indicate one or two slower dynamic processes for both N_{TAIL} variants, likely associated with the transitions from S2 to the contact state and the diffusion of an extended N_{TAIL} C-terminus to a more compact conformation (S1 or S2). Estimating precise time scales of these processes from the MD simulations is difficult due to the sparse sampling of these slow events, and due to the expected time scales being comparable or longer than the typical trajectory length (10-20 μ s) of our MD simulations.

The obtained estimated C-W quenching times from the first barrier crossing are shown in Supporting Table S4, where they are compared with the C-W quenching times (τ_{CW}) of the measured PET relaxation processes. N_{TAIL} MD simulations using the CHARMM36m/OPC force field predict τ_{CW} values of 5.3 (sd: 3.3) and 2.0 (sd: 0.5) μ s for the C3W2 and C4W2 variants, respectively. These agree remarkably well with the τ_{CW} estimates of 7.6 and 2.8 μ s from the measured PET relaxations. In comparison, crossing the first free-energy barrier appears to be a slower process using the Amber99SB-disp force field simulations, for which τ_{CW} values of 13.3 (sd: 3.2) and 4.7 (sd: 1.0) μ s were computed. We note that even though the CHARMM36m/OPC simulations agree better with the measured PET relaxation times the two force fields estimate similar 2.7 (CHARMM36m/OPC) and 2.8 (Amber99SB-disp) ratios between C3W2 and C4W2 relaxation times, in very good agreement with the measured 2.8 ratio.

In Supporting Figs. S22C and S22D, we compare the predicted W triplet-state survival curves (dashed lines) with the measured PET relaxation (dots). To this aim, we corrected the predicted decay curves to account for the measured natural relaxation $\tau_0 = 39.4 \mu$ s of the N_{TAIL} variant Y518W (W2), which does not have a cysteine quencher. Assuming that the natural and quenched decays are independent of each other, the corrected survival probability is calculated as $S(t)_{corr} = S(t)_{CW} \cdot S(t)_0$, where $S(t)_0 = e^{-t/\tau_0}$. For the simulations using the CHARMM36m/OPC force field, the corrected survival curve for the C3W2 N_{TAIL} variant shows an excellent agreement with the measured data (RMSD: 0.007), whereas for the fast C4W2 variant triplet-state relaxation appears to be slightly over-estimated, although this deviation (RMSD: 0.03) does not exceed the computational uncertainty (shaded areas). Natural relaxation corrections also improve the agreement between PET experiments and simulations from the Amber99SB-disp force field, which appear to underestimate the measured triplet-state

relaxation for both C3W2 (RMSD: 0.04) and C4W2 (RMSD: 0.05), but these deviations are only slightly larger than the considerable computational uncertainty.

As an additional probe of how well the simulations agreed with experimental data, we also computed the relaxation times for the C1W1, C2W1, and C3W1 variants from Charm36m/OPC force field trajectories. These simulations had the wild-type tyrosine (Y) residue in position 453. This of course limits the accuracy of PET relaxation time calculations, because the Y side chain is smaller and shaped differently than a W side chain, which in turn alters the short-range dye-quencher distance distribution. To partially compensate for the differences, the cutoff-distance for a reactive contact r_c was increased 0.6 nm for these calculations. With this modification, triplet-state relaxation curves and times were estimated as described above, and the relaxation times of all five variants were plotted as the function of sequence separation in Fig S30, similarly as was done for Fig 1D. The plot indicates that within the significant computational uncertainty, three out five relaxation times of the N_{TAIL} PET variants are reproduced. The relaxation time for the C4W2 variant is slightly underestimated ($2.0 \pm 0.5 \mu\text{s}$ vs. $2.7 \mu\text{s}$), whereas the relaxation time for C3W1 is overestimated ($5.7 \pm 1.8 \mu\text{s}$ vs. $3.2 \mu\text{s}$) by the simulations. The latter deviation may suggest that non-local interactions and contact frequency between N_{TAIL} regions B and D (seen state S2) are under-sampled in Charmm36m/OPC simulations.

S18: Data Processing

Basic plots and images in this work were generated using Grace v5.1.25 and Python environment v3.7.3, with the Numeric Python (NumPy) v1.16.4, Matplotlib v3.1.0, Scientific Python (SciPy) v 1.3.0, and Seaborn v0.9.0 extensions. Images were processed into finalized figures using GIMP v2.10.22 and Microsoft Office 2019.

S19: Free energy landscape calculations

The free energy surface shown in Fig. 3 of the main manuscript was computed from the minimum PET donor-acceptor distances described in Supporting Methods S16. Donor-acceptor minimum distance trajectories were converted into histograms between 0 and 100 Å, with bin width of 1 Å. Histogram counts were normalized and converted into probability densities using the Gaussian kernel density estimator from scipy v1.3.0. The free energy of bin i (in Boltzmann units) was determined as $F_i = -\log(P_i)$. The free energies of each bin were adjusted by an offset $F_0 = \min(F_i)$ to set the global free energy minimum to zero, and all free energy values above the upper limit $F_{max} = 6 \text{ kT}$ was set to F_{max} to provide an easy visualization of the landscape.

S20: Contact map calculations

The contact maps shown in Fig. 4, were calculated from ensembles of conformations extracted from all-atom MD simulations that are within free-energy minima of Fig. 3 of the main manuscript (see highlighted areas). Backbone contact maps of N_{TAIL} variants were calculated based on C α - C α pair distance distributions of these conformational ensembles. For each conformation, residues i and j were considered to be in a backbone contact if the distance between their C α positions was smaller than the cutoff distance $D_{C\alpha} = 0.6 \text{ nm}$. Salt bridge contact maps were calculated in a similar manner, with $D_{salt} = 0.35 \text{ nm}$ cutoff distance between amide hydrogen atoms of positively charged side chains donor residues (arginines and lysines) and negatively charged acceptor residues (O γ of aspartate, O δ of glutamate, or terminal carbonyl oxygen). Additionally, salt bridges were required to have N-H-O angle of 110° or larger.

S21: Coevolutionary analysis

Coevolutionary analysis was performed using EVcouplings (<https://evcouplings.org/>) with default parameters⁵¹. The statistical inference method is Pseudo-likelihood maximization. The Bitscore sequence inclusion threshold is 0.1. The sequence coverage is from residue 443 to 501

(59 amino acids). A total of 12072 sequences have been aligned and selected for analysis. Therefore, there are 204 sequences per amino acids with a run quality score of 8 (10 best, high quality). The coupling score instead of the folding model has been used to generate the contact map (Fig. S28) since for an IDP, these strongly coupled sites might not always be in contact.

S22: Mutual information calculations

To quantify correlations (including non-linear correlations) between the occurrence of PET donor-acceptor contact formation and other interactions, we calculated normalized pointwise mutual information (NPMI) scores based on the MD simulation trajectories. NPMI scores of Table S5 between PET donor-acceptor contacts (C3W2 or C4W2) and either backbone contacts or salt-bridge contacts (as described in Supporting Methods S20) were calculated from every conformation of all simulation trajectories of a given force field (*i.e.* not just within conformational states). PET donor-acceptor pairs were considered to be in contact if their minimum distance (see Supporting Methods S16) was below 0.4 nm. The NPMI scores were calculated according to:

$$NPMI(x, y) = -\frac{1}{\log(P(x, y))} \cdot \log\left(\frac{P(x, y)}{P(x) \cdot P(y)}\right)$$

Where $P(x)$ and $P(y)$ are, respectively, the probabilities of a PET donor-acceptor contact and of intramolecular contact (either backbone or salt-bridge) observed throughout the simulations. $P(x, y)$ is the probability of the two contacts observed simultaneously.

To establish if there is a direct correlation between the presence of identified contacts and the C-W electron transfer, we selected several contacts that are prominent in at least one of the conformational states, and computed the normalized pointwise mutual information (NPMI) between them and the two C-W donor-acceptor contacts as shown in Supporting Table S5. NPMI describes the co-occurrence of two events (see Supporting Methods S21). NPMI values range on a logarithmic scale between -1 (never co-occurring) and 1 (always co-occurring), with independent occurrences resulting in an NPMI value of zero.

As Table S5 shows, all NPMI values for S1 non-local interactions are negative, indicating that the observed interactions in this conformational state rarely co-occur (*i.e.*, are anti-correlated) with donor-acceptor contacts. This is expected because, as mentioned above in the contact map analysis, these interactions stabilize a non-contact conformational state. The negative NPMI values indicate that these interactions would indeed hinder electron transfer for both C4W2 and C3W2. There are few prominent interactions observed in the S2 state, with interaction between regions A, B, C often having positive correlations to C-W contact formation, and interactions between regions B and D showing negative correlations with C-W contact formations of both variants. Interactions observed in the S3 and S4 states have considerable positive NPMI-s with C4W2 and C3W2 donor-acceptor contact formation, respectively, confirming that these interactions co-occur and likely promote electron transfer.

Interestingly, the salt bridge R413-E480, which connects regions A and D, also has a positive NPMI with both C-W contacts, possibly stabilizing conformations that enable electron transfer for both C4W2 and C3W2. Further, the table highlights certain charged residues, like R413 from region A, K441 and R444 from region B, R490 from region D, as well as E511 and D520 from region E that apparently form different prominent stabilizing interactions in multiple conformational states. The mutual information between salt bridges occurring in different states shows strong negative correlations (NPMI of -0.25 – -0.50), indicating a set of competing interactions that may drive transitions between conformational states. The highlighted charged residues in region B apparently can form either non-local salt bridges with residues distant in

sequence close to the MoRE region, or local salt bridges with other charged residues within region B. We observe that many of these salt bridges are accompanied by clusters of other interactions (see encircled clusters in Fig. 4). It is possible that the formation of these clusters is nucleated by the salt bridges. In any case, switching between these sets of interactions is necessary for the system to transition between the most populated conformational states S1 and S2 in Fig. 3, and competition between these interactions is what determines the populations of the conformational states themselves.

We also used mutual information analysis to determine if certain non-local interactions are correlated to helix formation in the MoRE region, which may be of functional importance. To this aim, we computed NPMI scores between non-local contact formation and three substates of helix folding observed in our MD simulations. The low helix substate is defined by less than 20% of the N_{TAIL} MoRE residues found in a helical conformation which is the dominant helix sub-state both in the S1 and S4 states. The medium helix substate is intermediate folding state, where 20-50% of the MoRE residues are helical. This substate appears frequently in the S3 state, where the N-terminal part of the MoRE is folded more often than the C-terminal part. Finally, the high helix substate refers to conformations where 50-100% of MoRE folded, which is observed in (but not limited to) S2 conformational state.

Computing NPMI values between helix substates and interactions of region D, with particular attention paid to the interactions of residues that participate in binding the phosphoprotein X-domain (P_{XD}) shown in Supporting Table S7. The NPMI scores correlating selected prominent non-interactions with helix substates (low, medium, or high) are shown in Supporting Table S6. Most of the interactions can be placed in one of three broader categories.

The first group (red) interactions are negatively correlated with both medium and high helix contents, which includes several salt bridges from the S1 and S4 conformational states. These interactions likely stabilize non-helical conformations, and thus are expected to have an indirect and negative effect on phosphoprotein binding affinities. The second group of interactions (green) are positively correlated with both medium and high helix contents. Examples of such interactions include residues that directly form the binding site for phosphoprotein X-domain (P_{XD}) such as A447-M501. Such interactions observed in MD simulations may suggest a competition for the folded MoRE binding site between P_{XD} and intramolecular interactions between regions B and D. Interestingly this group also contains local salt-bridges within region B, such as K441-K449, which suggests that such interactions may allosterically promote helix formation. The third group of interactions (blue) are defined by positive correlations towards the medium helix substate, but low or negative correlations to the high and low helix substates. These interactions, such as R405-D484 observed in the S3 conformational state, or A409-T469 in S2, may stabilize folding intermediates of the MoRE region, thus reducing free-energy barriers and increasing binding/unbinding rates of P_{XD} .

S23: MoRE helicity in all-atom MD simulations

The most prominent secondary structure element in N_{TAIL} is the MoRE region (residues 484-502) that according to the NMR chemical shifts determined by Gely *et al.*⁵⁰, folds into a transient α -helical structure for up to 50% of time. This transient helix is stabilized and folds into four-helix bundle when binding to Measles phosphoprotein X domain (P_{XD}) which is an important part of the viral replication process of measles.

There have been previous simulation studies that focused on the helicity in the MoRE region and the coupled binding and folding of the MoRE- P_{XD} complex. Robustelli *et al.*⁵² used the Amber99SB-disp force field to perform a single 100 μ s simulation of the MoRE region, and used two metrics to compare to NMR experiments; the RMSD between measured and computed $C\alpha$ chemical shifts (RMSD: 0.52 ppm) and Pearson correlation of the same data (R: 0.92). Wang *et al.*⁵³ performed 96 130 ns temperature replica exchange simulations using the

CHARMM22* force field for a total sampling time 12.4 μ s on MoRE. They only provided Pearson correlation (0.76) for the chemical shift profiles, which appeared to be in good agreement with the measured profiles. The two studies had differences in details, but both agreed on the approximately 20% helix content in the MoRE region, considerably more helical content (~50%) at the P_{XD} binding site (491-501) than the region preceding it (482-490). They both suggested a strong induced fit character of coupled binding and folding mechanism.

Our two best wild-type N_{TAIL} (401-525) simulations were performed using CHARMM36m-OPC force field (RMSD: 0.47 ppm, R: 0.83) and the Amber99SB-disp (RMSD: 0.36 ppm, R: 0.84) simulation provided by Piana *et al.* The former set of simulations were performed in six replicates using different starting structures for 5-10 μ s length (for a total of 41 μ s), whereas the latter was a single 30 μ s trajectory, which was divided into three parts and analyzed separately for determining sampling uncertainty (Table S8). Both of these simulations of the full N_{TAIL} predict transient helical structures of the MoRE region, with similar 16-18% helix content. The two force fields also agree in that the P_{XD} binding site (491-501) has a higher content than the preceding segment as shown in Fig. S31, but the difference is considerably smaller for the Charmm36M-OPC simulations than for Amber99SB-disp. This difference appears to be a force-field family dependent difference, because similar distributions are shown in the helicity profiles of Wang and Robustelli as well.

As far as we can tell, no NMR data is available for the PET variants, but the measured CD spectra of W2 variant indicate a change in the secondary structure composition compared to the wild type construct (likely with a small decrease of the helix content). This difference may be caused by a combination of the different ionic strength, the absence of the N-terminal hexahistidine tag, and the Y518W point mutation.

Simulation trajectories C3W2 and C4W2 variants using CHARMM36m-OPC show a smaller helix content (~10%) in the MoRE, with a much lower helicity at the 482-490 MoRE segment, which is compatible with the W2 CD spectrum. In contrast Amber99SB-disp simulations predict an increased helix content (~29%) in the same region. We also note that helix folding dynamics in the CHARMM36m forcefield appears to be considerably slower than in Amber, and the average secondary structure content did not fully converge within the 20 μ s simulation trajectories (showing a slow upwards trend in the second half of the simulations).

B. Supporting Figures

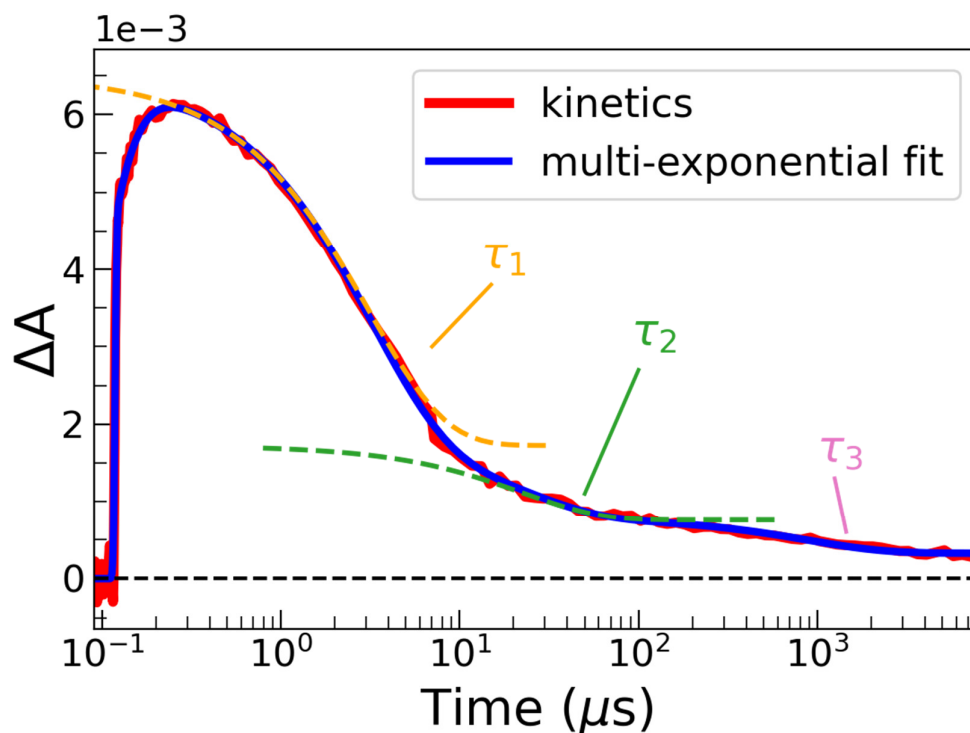


Fig. S1. Example of PET experiment raw data (red) with a multi-exponential fit (blue). ΔA (the transient absorption measured at 458 nm) is proportional to the W triplet state population, which decays in time. Dashed curves represent individual exponential decays contributing to the fit. The main component of the signal τ_1 (orange) corresponds to the observed triplet state relaxation time from the intact protein, in the presence of the C quencher. The smaller components at longer times (τ_2 , τ_3 ,...) correspond to photoproducts, as shown in Supporting Fig. S2.

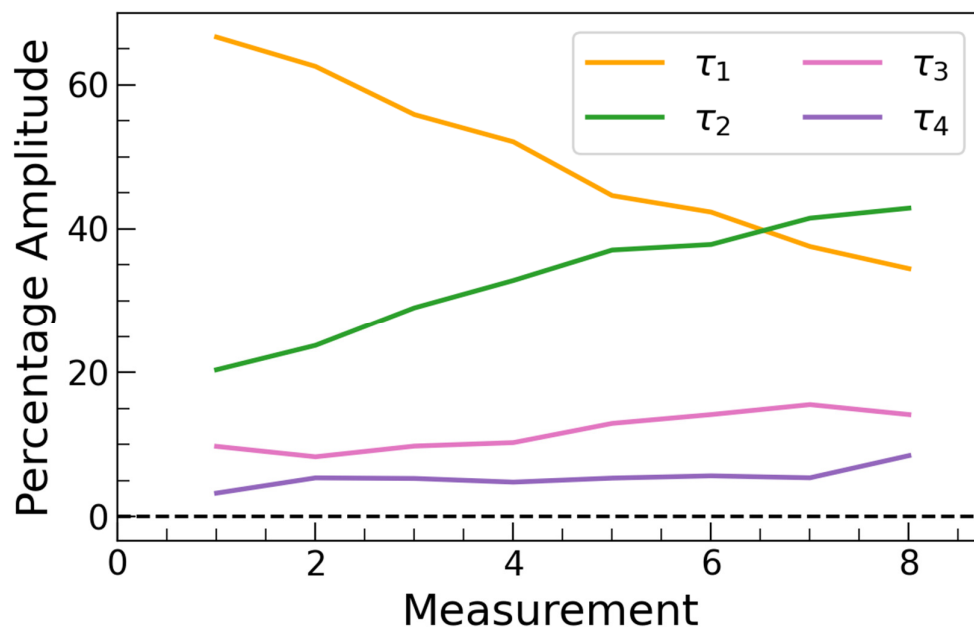


Fig. S2. The percentage amplitude of the multi-exponential fit components, as a function of UV pulse irradiation (each measurement corresponds to 256 UV pulses). The largest amplitude corresponds to the main component τ_1 (6.37 μ s, orange), and decreases steadily as a function of UV laser irradiation, indicative of photo-damage. The next significant component ($\tau_2 = 48.1$ μ s, green) starts at much lower amplitude and increases steadily, indicating that this is a photoproduct. τ_3 (1020 μ s, pink) and τ_4 (1×10^6 μ s, purple) components follow suit and are much smaller. Sample conditions: 482C-518W in 15 mM Tris, 150 mM NaCl, pH 7.6.

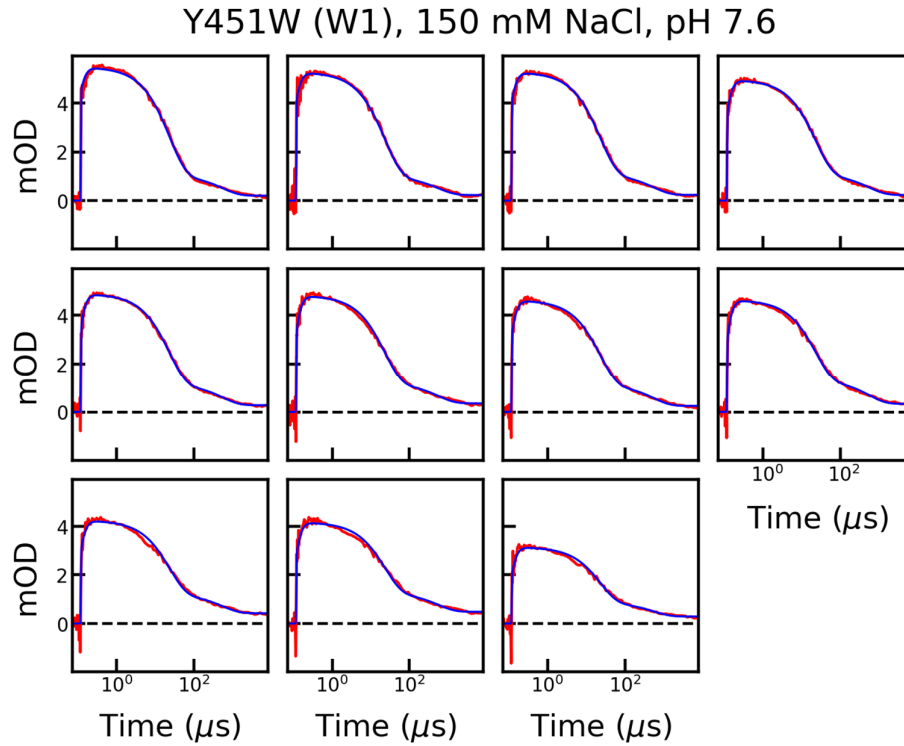


Fig. S3. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of Y451W (W1). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

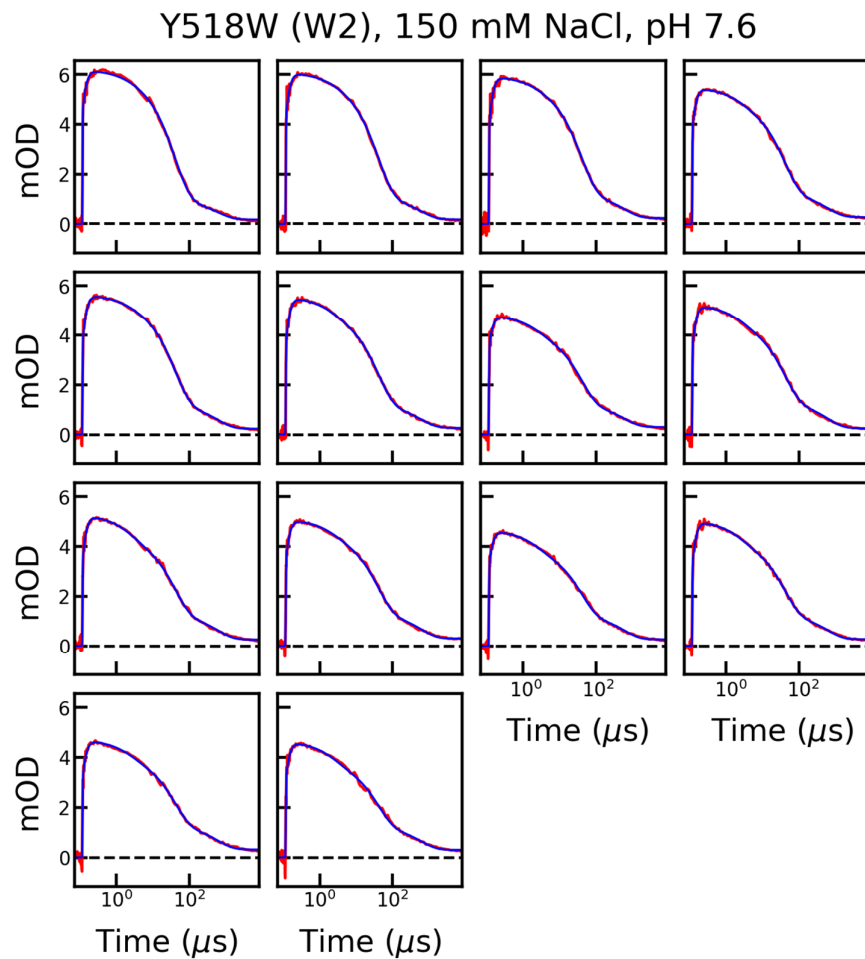


Fig. S4. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of Y518W (W2). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

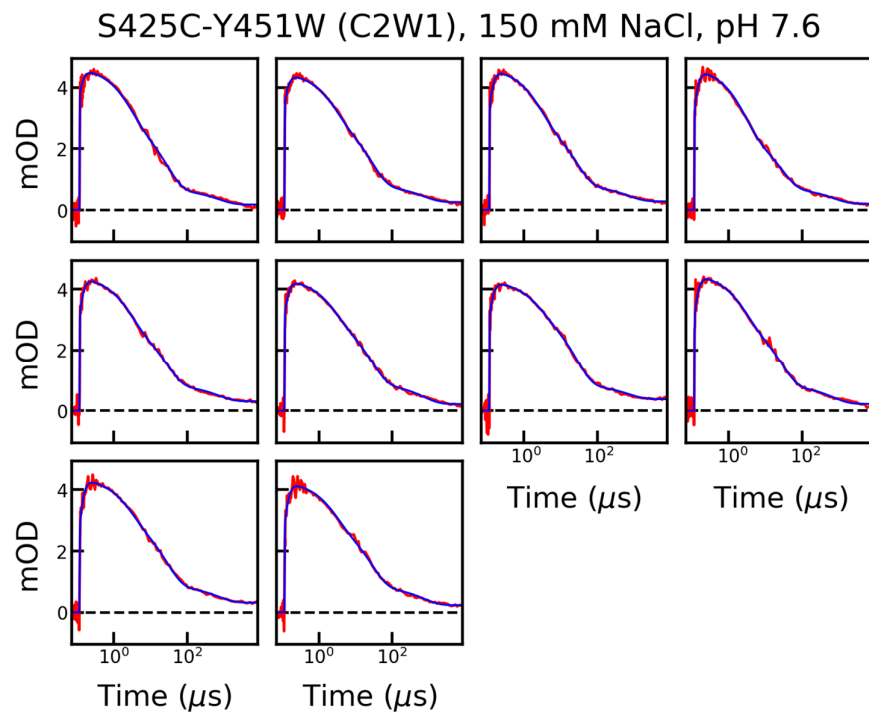


Fig. S5. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S425C-Y451W (C2W1). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

S488C-Y518W (C4W2), 150 mM NaCl, pH 7.6

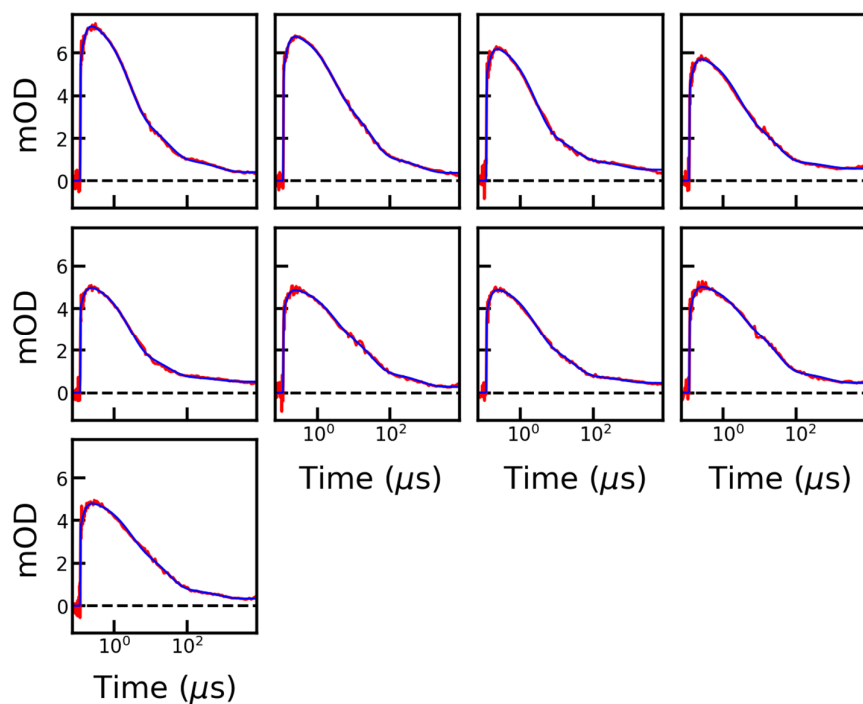


Fig. S6. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S488C-Y518W (C4W2). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

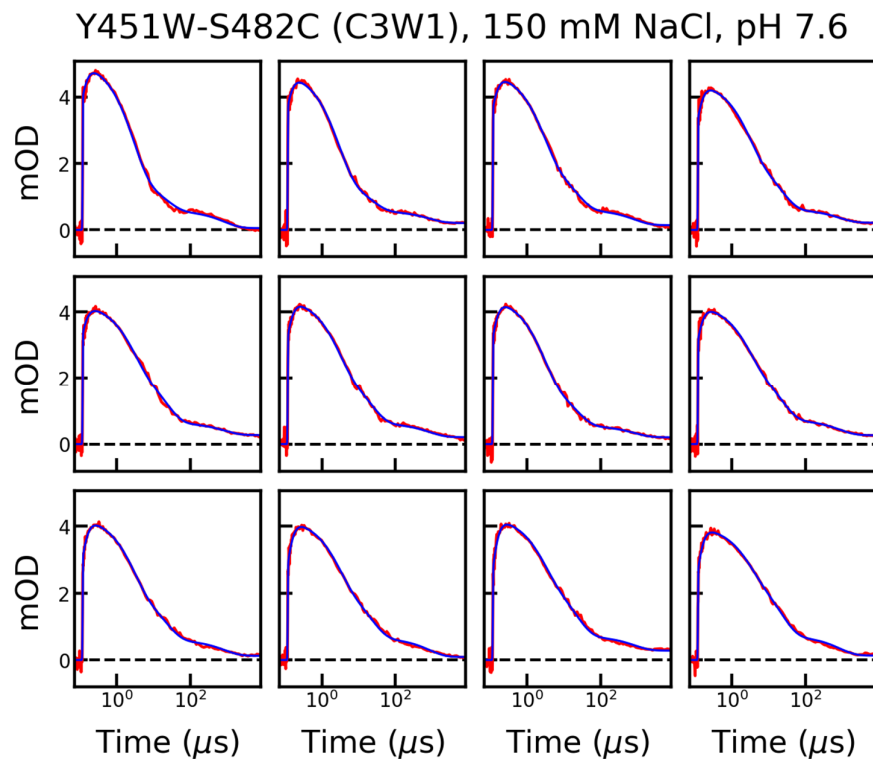


Fig. S7. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of Y451W-S482C (C3W1). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

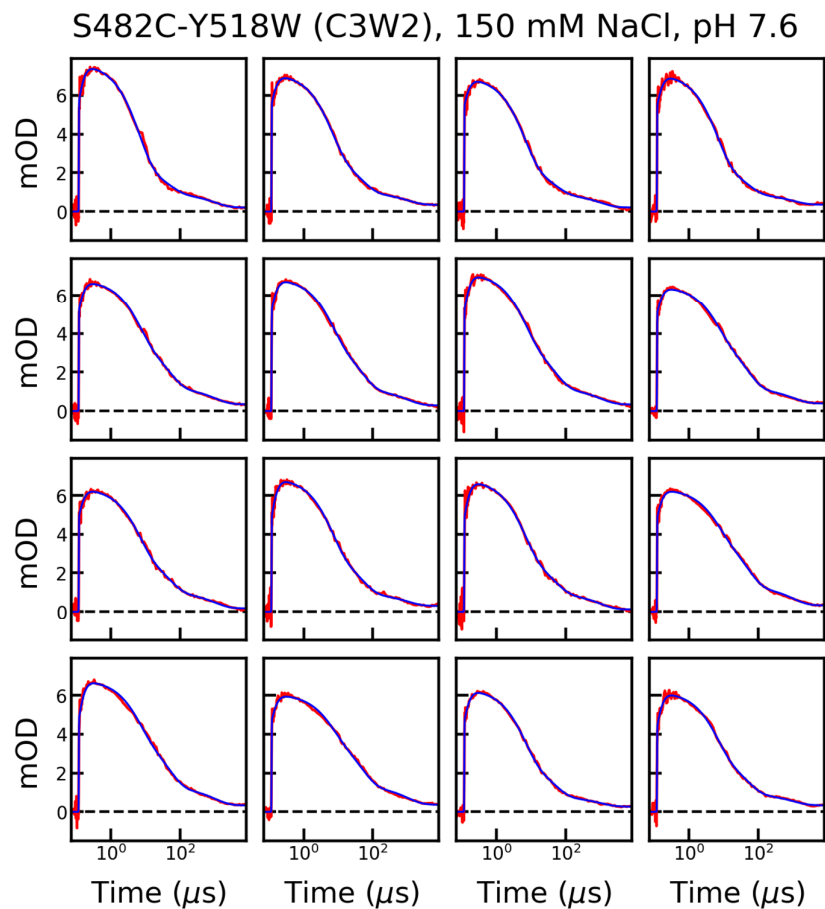


Fig. S8. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S482C-Y518W (C3W2). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

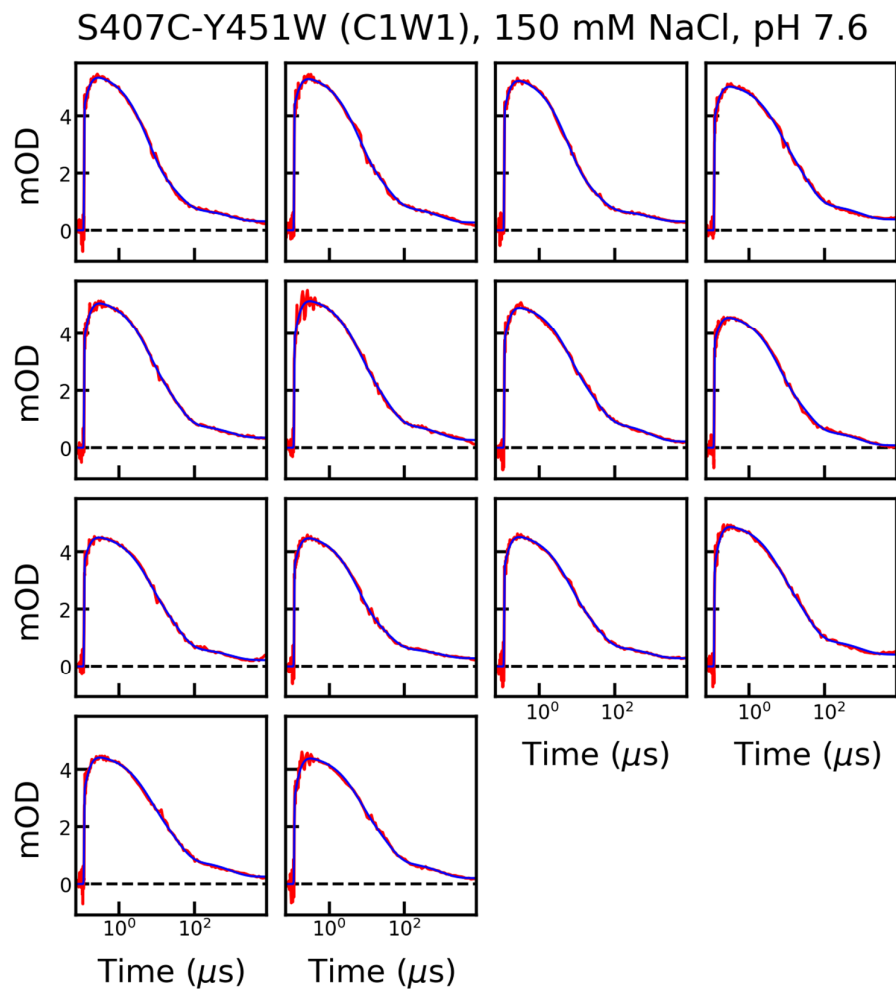


Fig. S9. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S407C-Y451W (C1W1). Experimental conditions: 15 mM Tris, 150 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

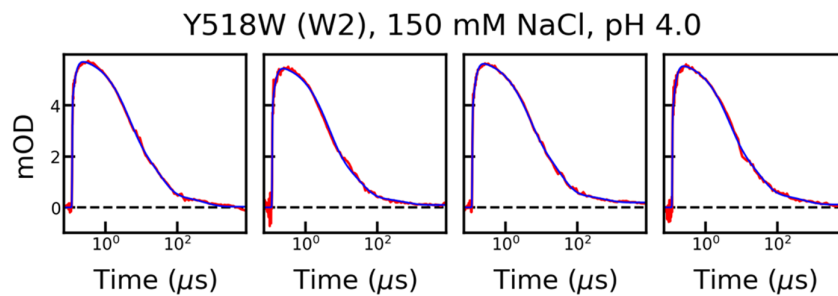


Fig. S10. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement Y518W (W2). Experimental conditions: 20 mM NaAc, 150 mM NaCl, 1 mM TCEP, pH 4.0. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

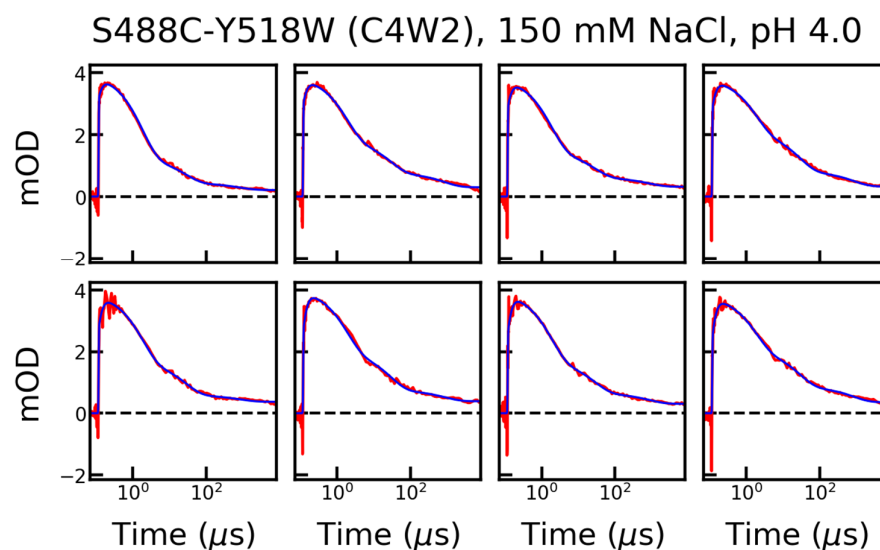


Fig. S11. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S488C-Y518W (C4W2). Experimental conditions: 20 mM NaAc, 150 mM NaCl, 1 mM TCEP, pH 4.0. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

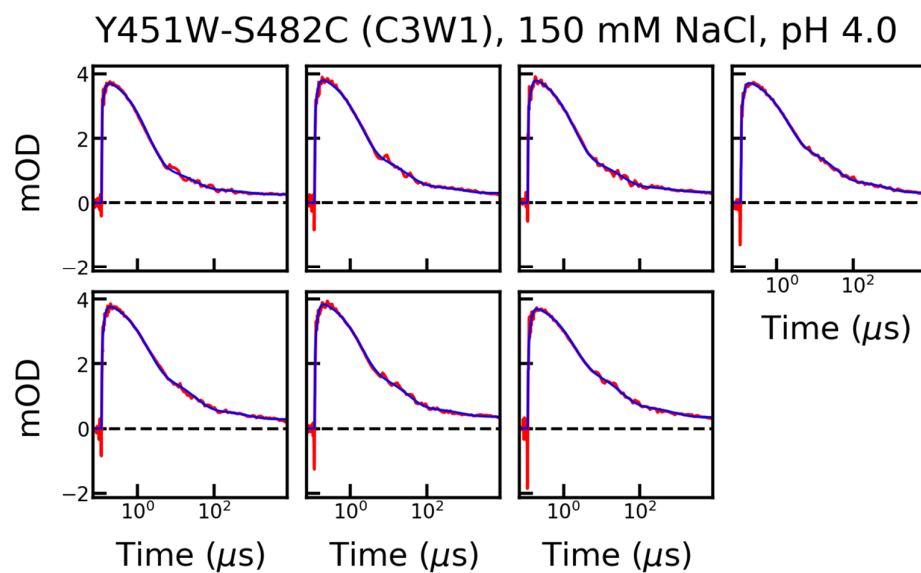


Fig. S12. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of Y451W-S482C (C3W1). Experimental conditions: 20 mM NaAc, 150 mM NaCl, 1 mM TCEP, pH 4.0. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

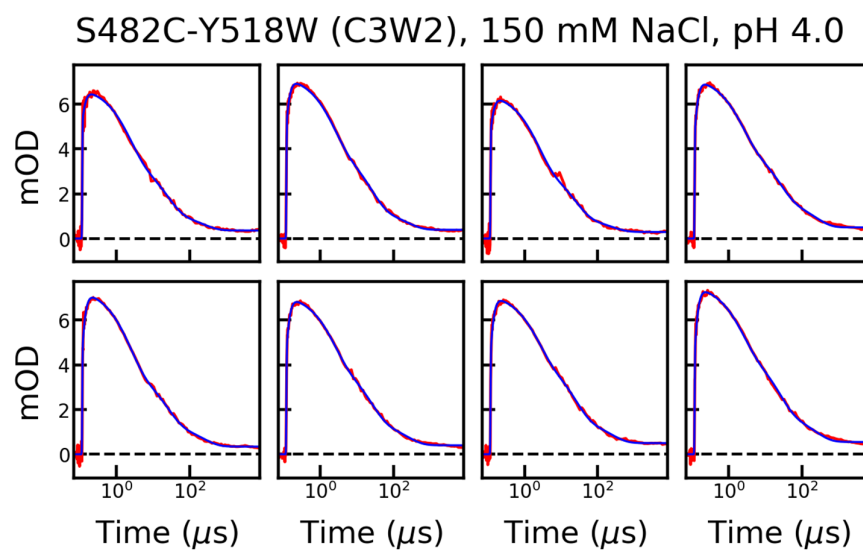


Fig. S13. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S482C-Y518W (C3W2). Experimental conditions: 20 mM NaAc, 150 mM NaCl, 1 mM TCEP, pH 4.0. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

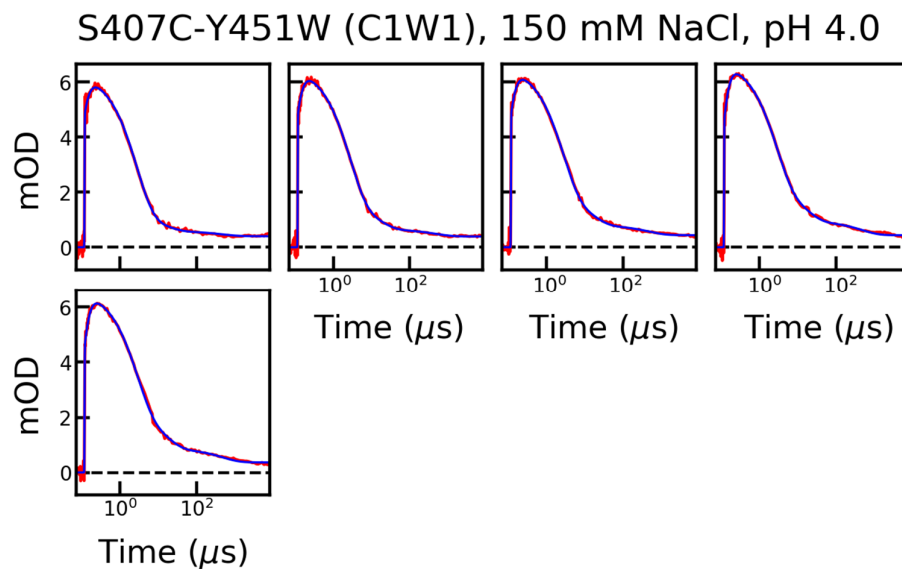


Fig. S14. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S407C-Y451W (C1W1). Experimental conditions: 20 mM NaAc, 150 mM NaCl, 1 mM TCEP, pH 4.0. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

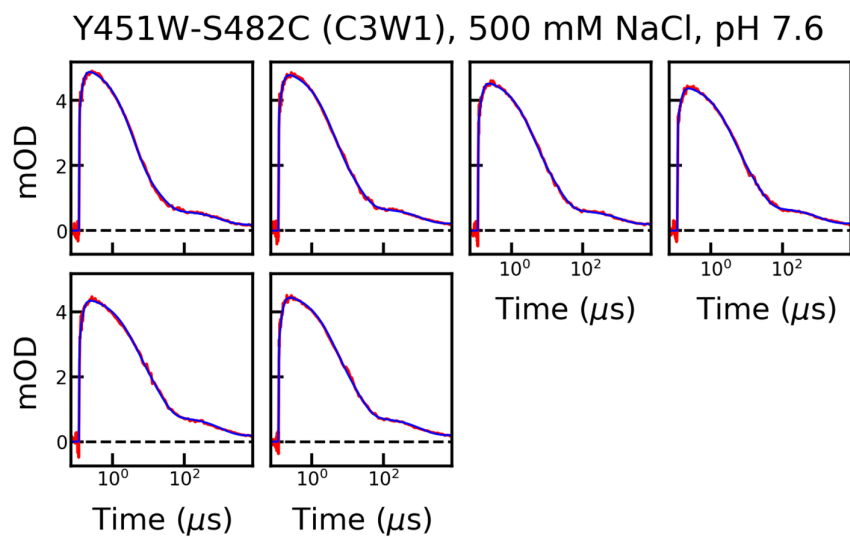


Fig. S15. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of Y451W-S482C (C3W1). Experimental conditions: 15 mM Tris, 500 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

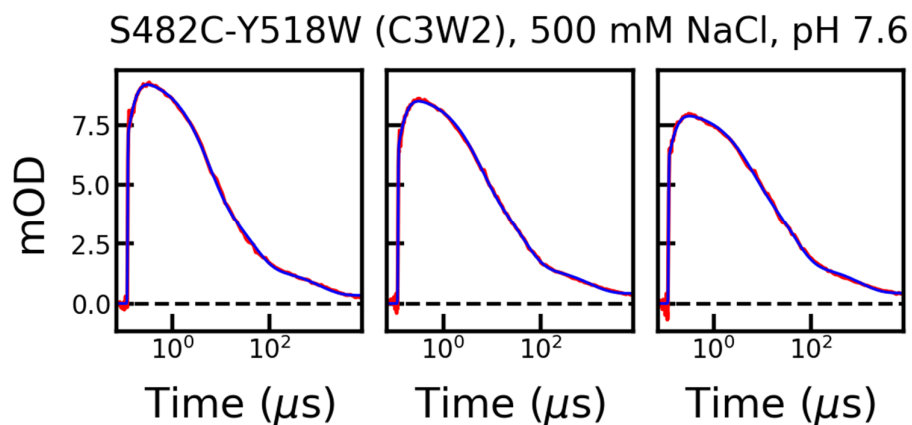


Fig. S16. PET experiment raw data (red) with multi-exponential fit (blue) for each measurement of S482C-Y518W (C3W2). Experimental conditions: 15 mM Tris, 500 mM NaCl, 1 mM TCEP, pH 7.6. To avoid excessive photodamage, only measurements for which the main component contributed 35% or more to the amplitude are used.

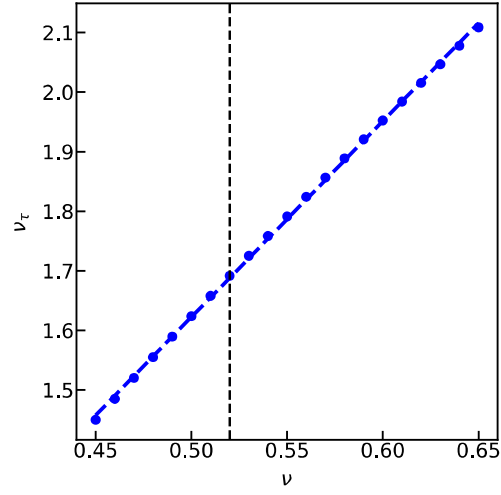


Fig. S17. Linear dependence between the C-W quenching time scaling exponent ν_τ and the Flory (distance) scaling exponent ν . ν_τ was calculated for different values of ν , using the SAW- ν model (blue points). The black dashed line indicates the value of ν for N_{TAIL} at pH 7.6. The linear best fit function (blue dashed line) $\nu_\tau = 3.292 \times \nu - 0.024$ can be used to describe the relation between the two scaling exponents.

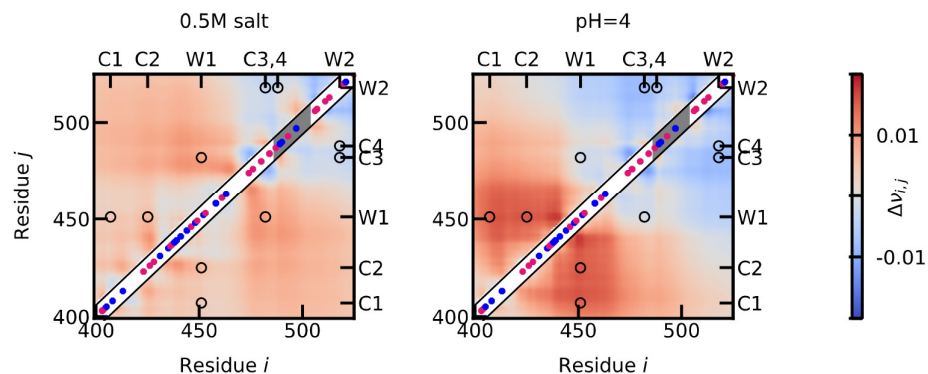


Fig. S18. A) The difference of the pairwise scaling exponents ($v_{i,j}$) between the CG simulations at 0.5M salt (Left) and pH 4.0 (Right) and the simulations at the reference condition (0.15 M salt and pH 7.6). The pairwise scaling exponents are calculated from the pairwise distances ($R_{i,j}$) as $R_{i,j} = b|i - j|^\nu$, with $b=0.55\text{nm}^{12,17}$. The diagonal shows the charged amino acid positions with positively charged amino acids in blue and negatively charged amino acids in magenta. The MoRE is highlighted in grey.

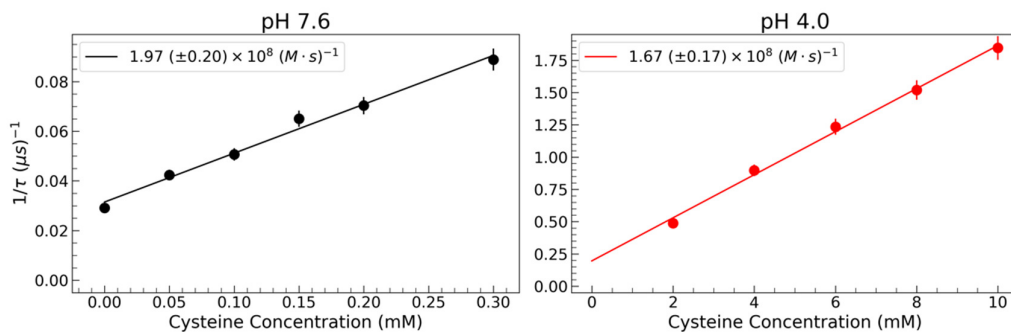


Fig. S19. Bimolecular quenching rates at pH 7.6 (black) compared to pH 4.0 (red). Note the different scales on both the x- and y-axes. The calculated rate at pH 7.6 is $1.97 (\pm 0.20) \times 10^8 \text{ (M} \cdot \text{s)}^{-1}$ compared to $1.67 (\pm 0.17) \times 10^8 \text{ (M} \cdot \text{s)}^{-1}$, which are the same within experimental error.

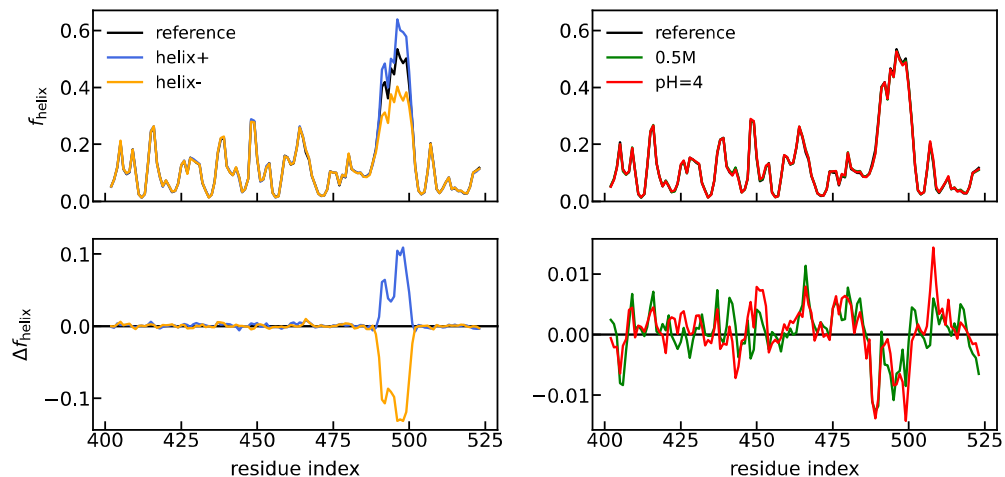


Fig. S20. Top left: Helical propensities of the CG model when varying a free parameter ϵ_{dih} controlling the stability of the helical minimum so as to introduce more helix (helix+ in blue with $\epsilon_{dih} = 0.4 \text{ kcal/mol}$) or less helix (helix- in orange with $\epsilon_{dih} = 0.1 \text{ kcal/mol}$). Top right: Helical propensities of the CG model at 0.5M salt (green) and pH=4 (red). The reference condition at 0.15 M salt, neutral pH and $\epsilon_{dih} = 0.2 \text{ kcal/mol}$ is shown in black. The corresponding difference of helical propensities to the reference condition is shown in the bottom.

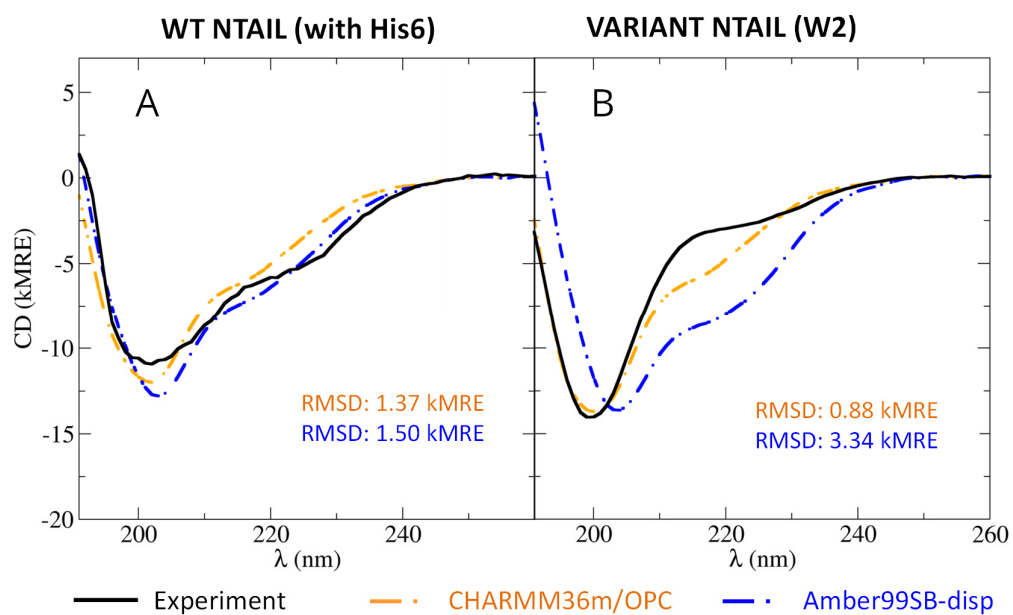


Fig. S21 Predicted CD spectra calculated from the wild type (A) and variant (B) MD simulations. CD spectra predicted from CHARMM36m-OPC simulations are shown as orange dashed lines, whereas CD predictions from AMBER99SB-disp simulations blue dashed lines. The root-mean-squared deviation (RMSD) from the measured CD spectra (black solid lines) are shown in a corresponding color.

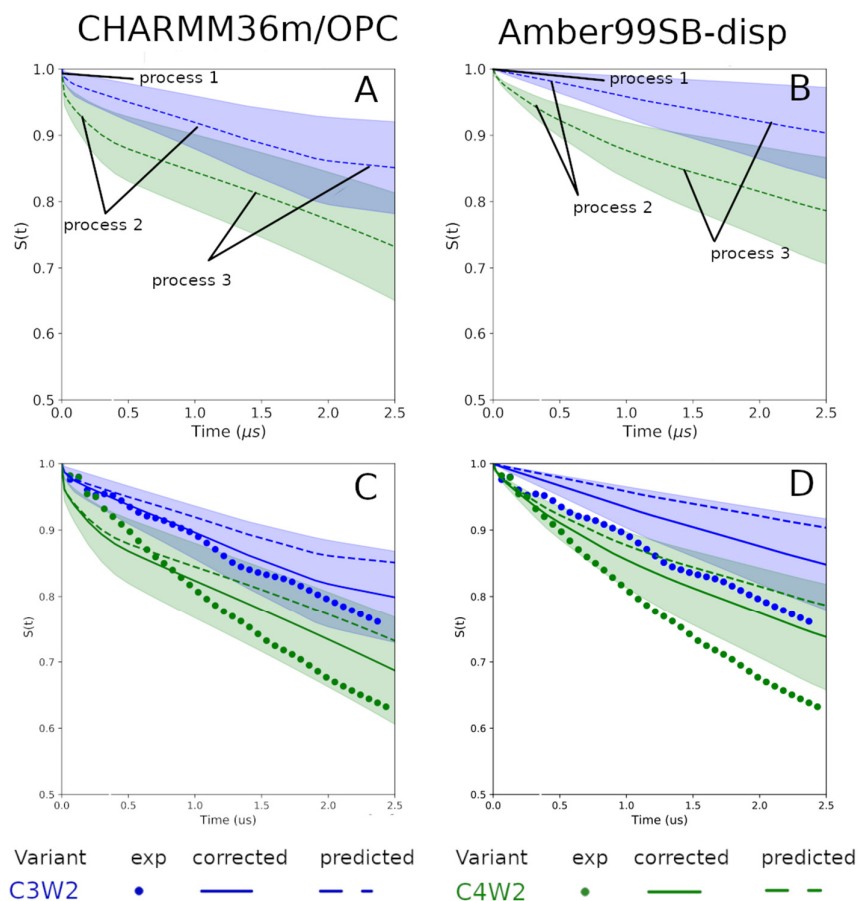


Fig. S22. PET experiment curves calculated from the CHARM36m-OPC (left) and AMBER99SB-disp (right) force fields are shown as dashed lines, with their computational uncertainty depicted as shaded areas. Estimated PET data of C3W2 (S482C-Y518W) and C4W2 (S488C-Y518W) labelling positions are shown in blue and green, respectively. The uncorrected PET experiment curves in (A) and (B) indicate several dynamic processes present in all MD simulations with different rate-limiting steps that change the decay rate of the survival probabilities ($S(t)$). In panels (C) and (D) the computed decays are corrected for the estimated unquenched decay (solid lines), and compared to the measured decay curves (dots). The unquenched decay of 39.4 μ s was estimated from relaxation decay of Y518W single variant as shown in Supporting Table S2.

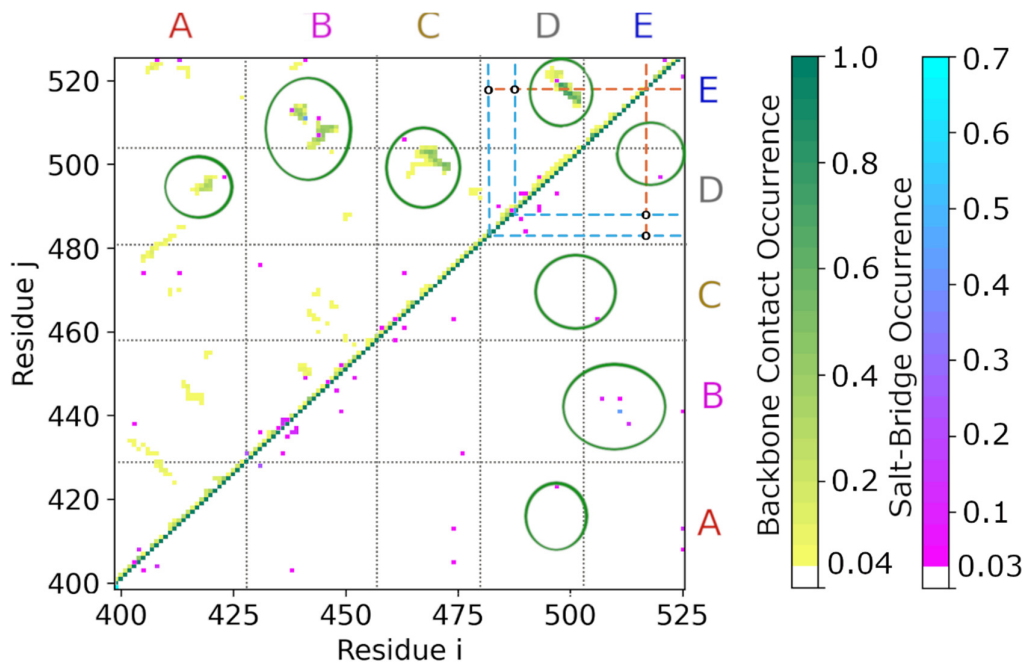


Fig. S23. Contact map for the S1 state shows C_{α} contact probabilities (above the diagonal, yellow to green) and salt bridge probabilities (above and below the diagonal, purple to blue). Cysteine acceptor positions C3 and C4 (residues 482 and 488) as well as tryptophan donor position W2 (residue 518) are indicated by cyan and orange dashed lines, respectively, and hypothetical C-W contacts are indicated by black circles. The boundaries of identified interaction regions A-E and are separated by black dotted lines on the contacts maps. The most prominent interactions in each state are encircled.

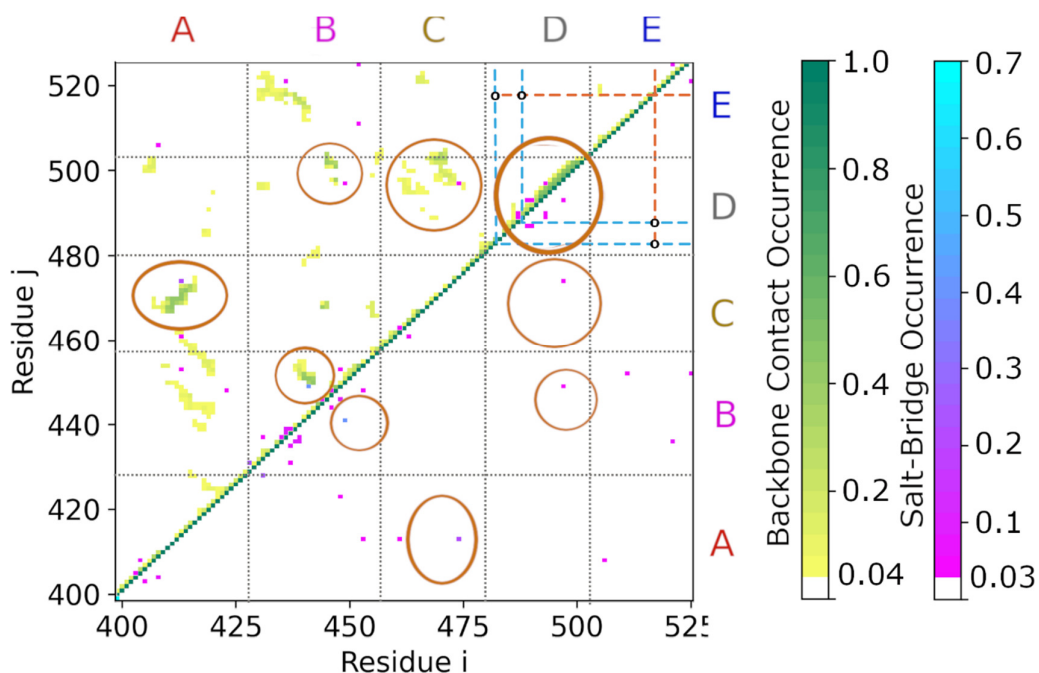


Fig. S24. Contact map for the S2 state shows C_{α} contact probabilities (above the diagonal, yellow to green) and salt bridge probabilities (above and below the diagonal, purple to blue). Cysteine acceptor positions C3 and C4 (residues 482 and 488) as well as tryptophan donor position W2 (residue 518) are indicated by cyan and orange dashed lines, respectively, and hypothetical C-W contacts are indicated by black circles. The boundaries of identified interaction regions A-E and are separated by black dotted lines on the contacts maps. The most prominent interactions in each state are encircled.

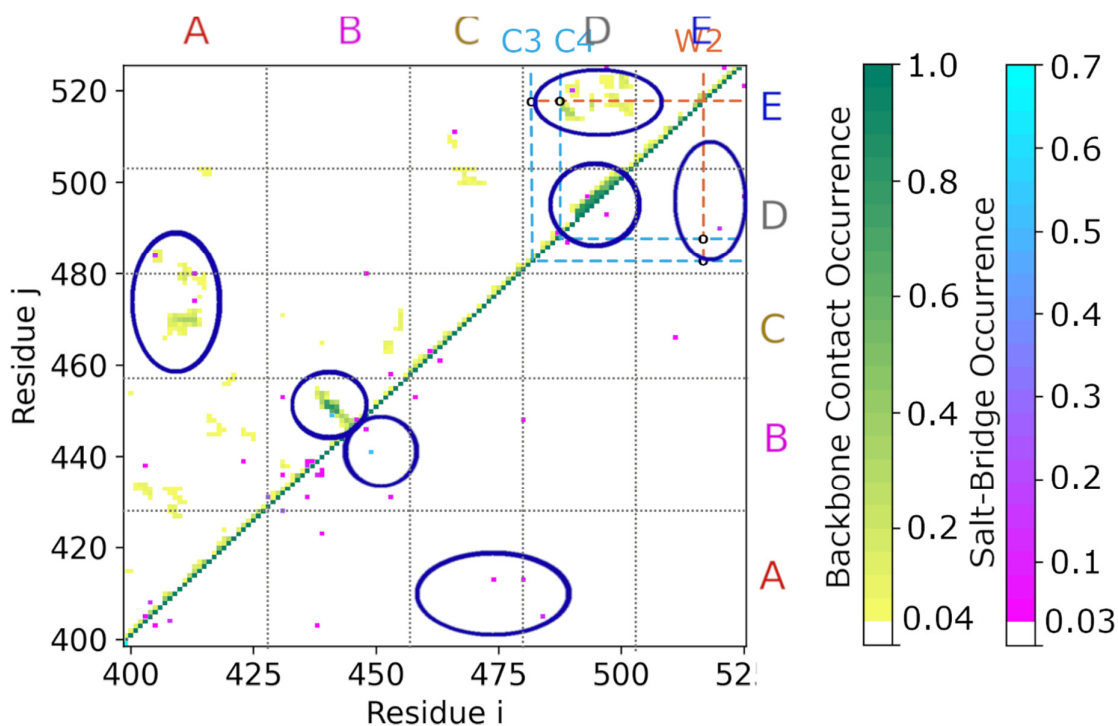


Fig. S25. Contact map for the S3 state shows C_{α} contact probabilities (above the diagonal, yellow to green) and salt bridge probabilities (above and below the diagonal, purple to blue). Cysteine acceptor positions C3 and C4 (residues 482 and 488) as well as tryptophan donor position W2 (residue 518) are indicated by cyan and orange dashed lines, respectively, and hypothetical C-W contacts are indicated by black circles. The boundaries of identified interaction regions A-E and are separated by black dotted lines on the contacts maps. The most prominent interactions in each state are encircled.

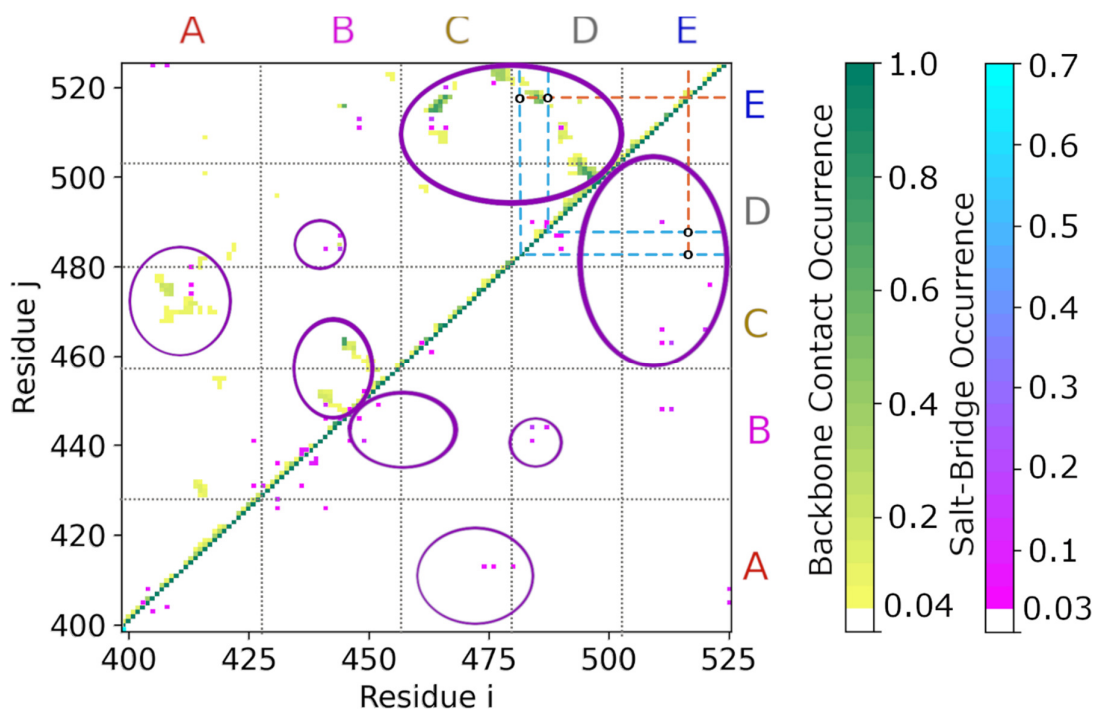


Fig. S26. Contact map for the S4 state shows C_{α} contact probabilities (above the diagonal, yellow to green) and salt bridge probabilities (above and below the diagonal, purple to blue). Cysteine acceptor positions C3 and C4 (residues 482 and 488) as well as tryptophan donor position W2 (residue 518) are indicated by cyan and orange dashed lines, respectively, and hypothetical C-W contacts are indicated by black circles. The boundaries of identified interaction regions A-E and are separated by black dotted lines on the contacts maps. The most prominent interactions in each state are encircled.

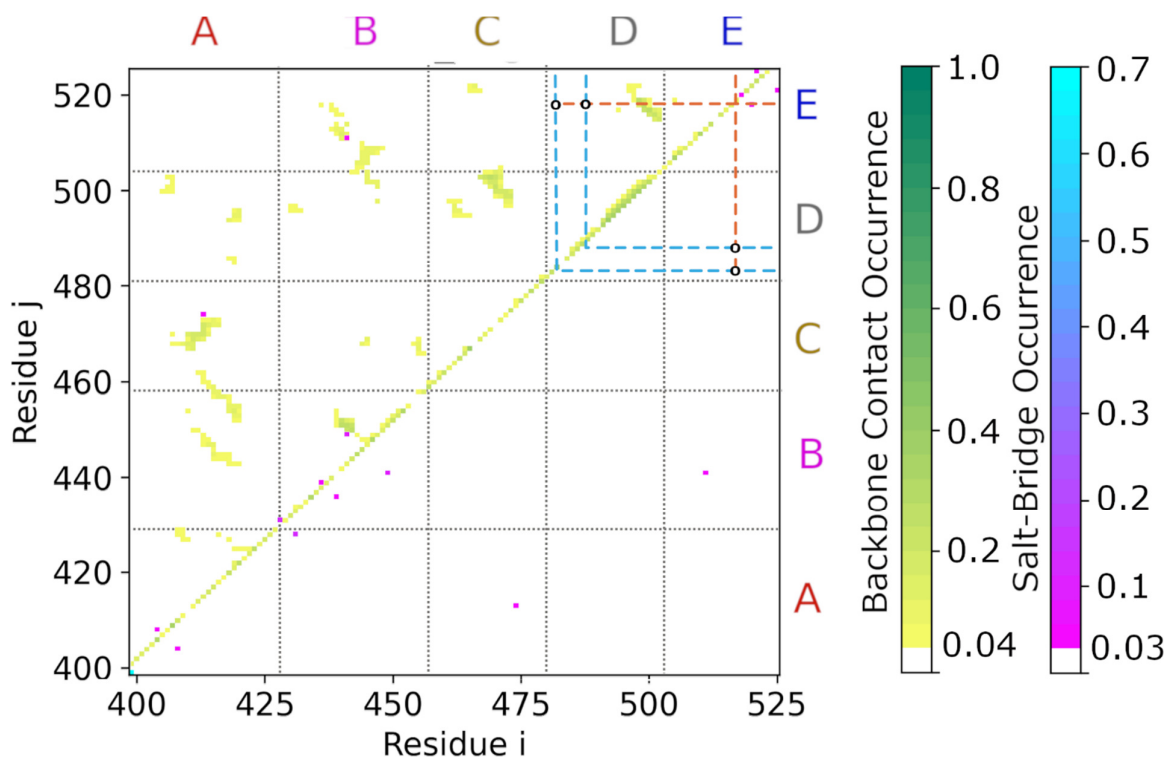


Fig. S27. Averaged contact map for all Charmm36m-OPC simulations. Contact map shows C_{α} contact probabilities (above the diagonal, yellow to green) and salt bridge probabilities (above and below the diagonal, purple to blue). Cysteine acceptor positions C3 and C4 (residues 482 and 488) as well as tryptophan donor position W2 (residue 518) are indicated by cyan and orange dashed lines, respectively, and hypothetical C-W contacts are indicated by black circles. The boundaries of identified interaction regions A-E and are separated by black dotted lines on the contacts maps.

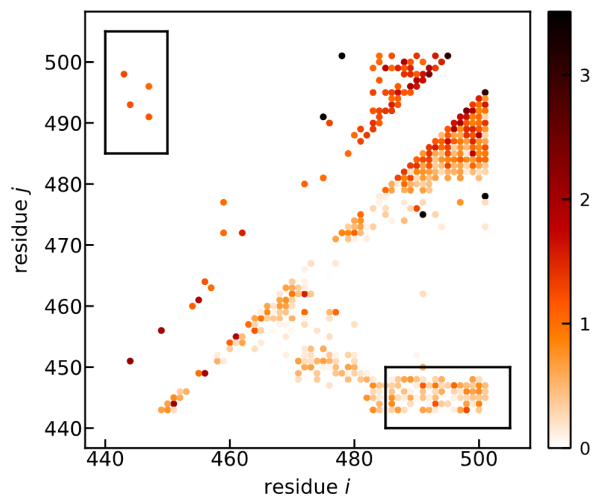


Fig. S28. Coupling score from using EVcouplings analysis. The region of the plot above the diagonal shows only the amino acid pairs with scores larger than 1. The region below the diagonal shows all pairs with scores larger than 0. The black box highlights the amino acid pairs between region B (residues 440-450) and region D (485-505)

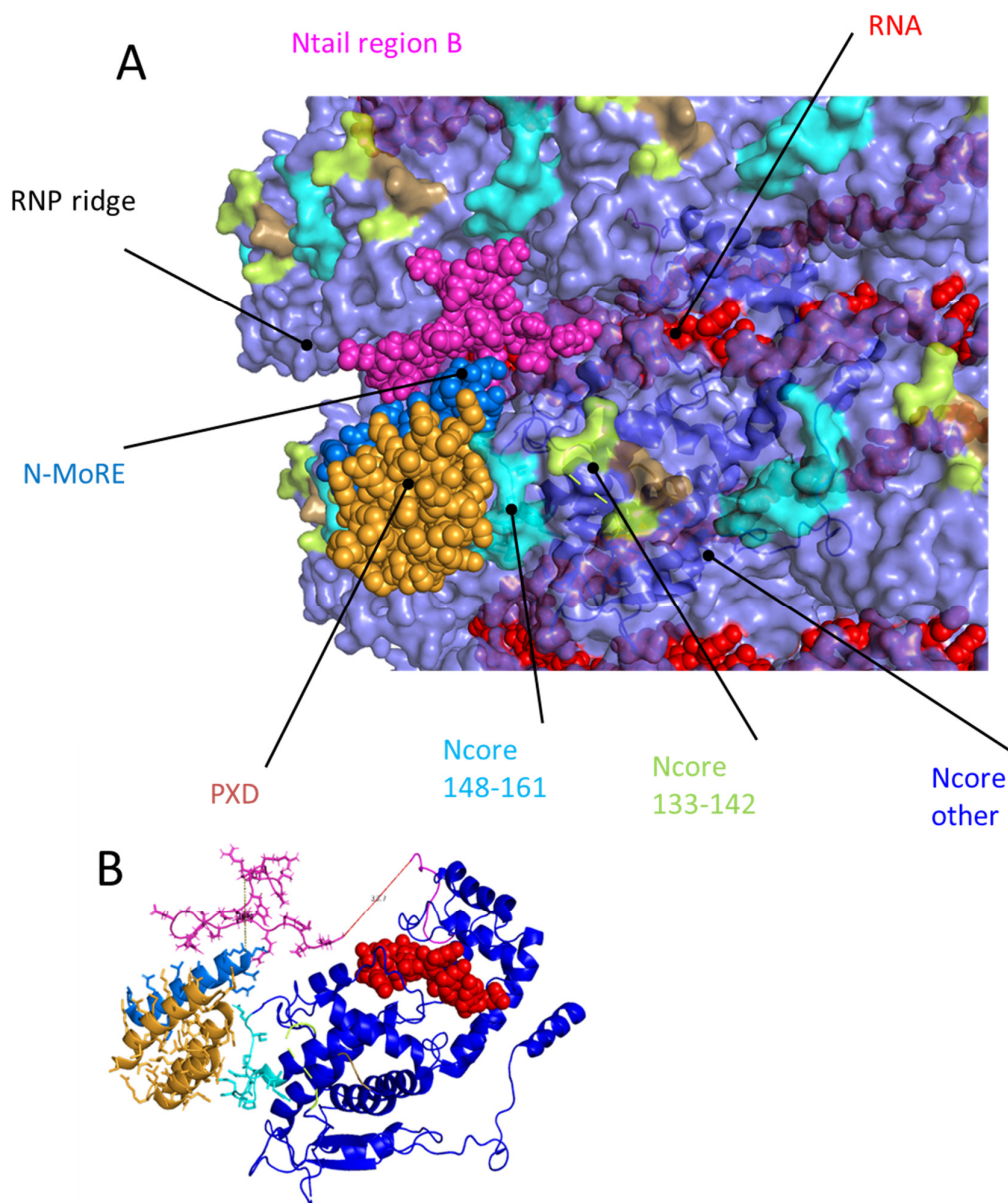


Fig. S29. Illustrative model of N_{TAIL} in the nucleocapsid RNA-binding ridge. A) surface and sphere representation of the RNA binding ridge. B) Cartoon representation of the RNA binding ridge with only one N protomer displayed. The model shows how the N_{TAIL} MoRE-PXD complex can bind N_{CORE} at the RNA binding ridge, with sufficient space in the ridge to facilitate N_{TAIL} intramolecular interactions between MoRE and region B. Note that Region B in this model is 3.5 nm away from the C_{ARM} region of N_{CORE} , which may be bridged by the 28 amino acids of region A. This model was generated manually, based on the Cryo-EM structure of the nucleocapsid (pdb code: 4uft). The available structure of the P_{XD} -MoRE complex (pdb code: 1to6) was docked to the P_{XD} - N_{CORE} binding site. Finally, a conformation of Ntail region B which contacted MoRE in the MD simulations was added into the nucleocapsid ridge.

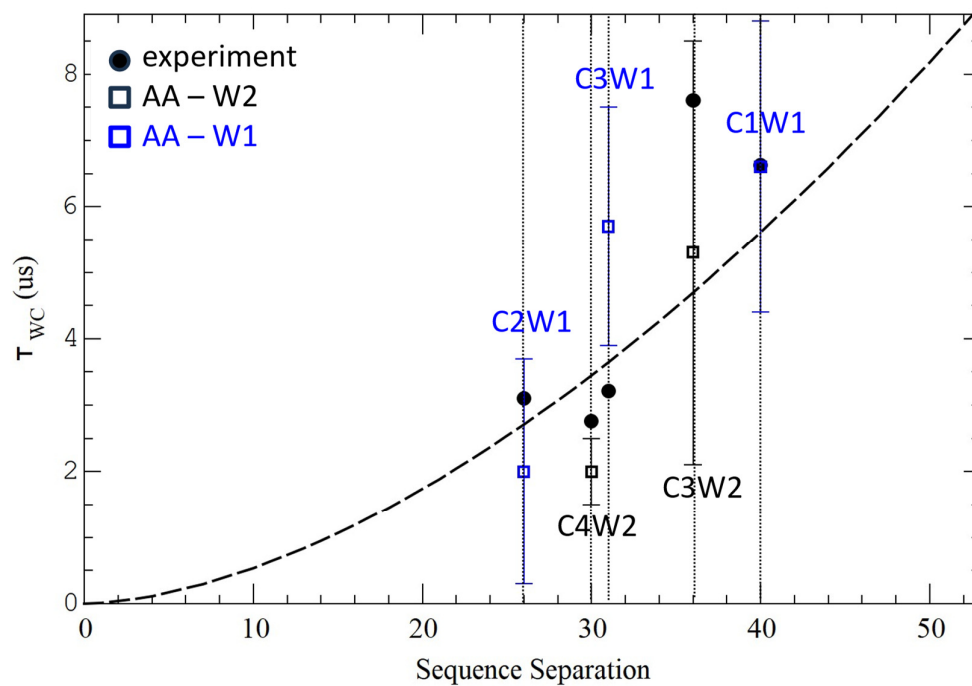
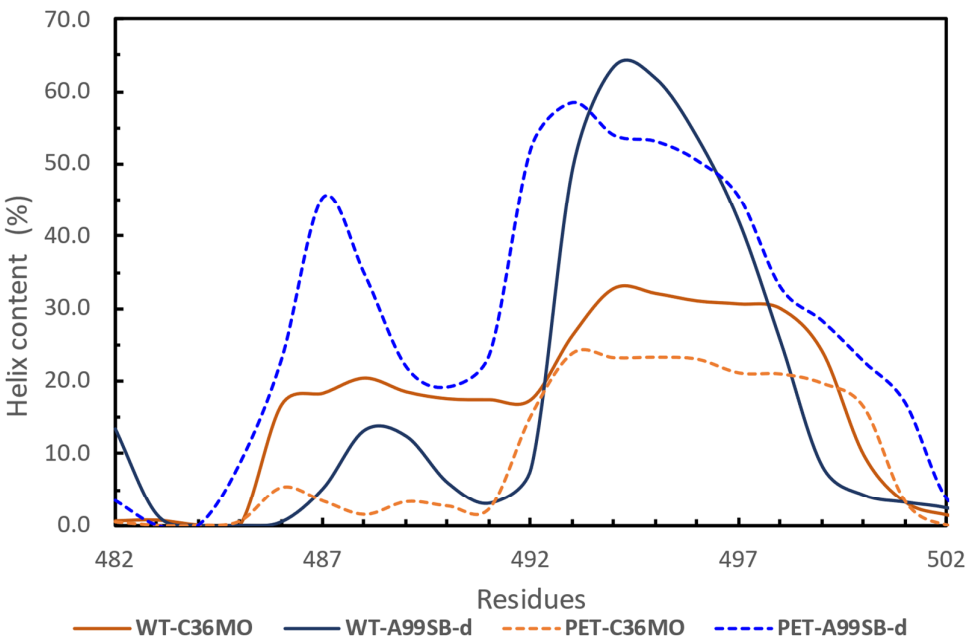


Fig. S30. Comparison between estimated C-W relaxation times for N_{TAIL} PET variants obtained from W triplet-state relaxation experiments at the reference condition (circles, pH 7.6 and 150 mM NaCl concentration), and all-atom molecular dynamics (AA, squares) as function of donor-acceptor sequence separation. The estimated error of the estimated relaxation times is indicated by bars. The homopolymer model fitted to the measured relaxation times (Fig. 1D) is shown as a black dashed line. The AA relaxation times for the C1W1, C2W1, and C3W1 variants (blue squares) were estimated based S-Y contact distances with a cutoff distance 0.6 nm as described in Supplementary Methods S17.

MoRE Helix content



Forcefield	WT-MoRE	WT	-total	PET	-MoRE	PET	-total
C36M-OPC	17%	3%	10%	2%			
A99SB-disp	18%		8%	29%	12%		

Fig S31: More Helix content in AA simulations. Top: The average helix content for residues in the MoRE region for WT (solid lines) and PET variant (dashed lines) simulations using the CHARMM36m-OPC (C36MO, orange) and Amber99SB-disp (A99SB-d, blue) force fields are shown. Bottom: the average helix content for MoRE region as a whole, and the full N_{TAIL} (total) are shown.

C. Supporting Tables

Table S1. N_{TAIL} variants and sequences used in the PET experiments. Residues in magenta denote acidic residues, blue letters indicate basic residues, cysteines are shown in cyan and tryptohans are highlighted in orange.

Variant	Sequence
Wild type	GS 401TT EDKIS RAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S Y RE TG PS RA S DARA AHL PTG TPLD IDTASE 481SSQ DPQDSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV YND 521 RNLLD
Y518W (W2)	GS 401TT EDKIS RAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S Y RE TG PS RA S DARA AHL PTG TPLD IDTASE 481SSQ DPQDSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV W ND 521 RNLLD
Y451W (W1)	GS 401TT EDKIS RAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S W RE TG PS RA S DARA AHL PTG TPLD IDTASE 481SSQ DPQDSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV YND 521 RNLLD
S425C-Y451W (C2W1)	GS 401TT EDKIS RAV GPRQAQVSFL HGDQ CE NE LP RLGGK EDRR V 441 KQSRGEARE S W RE TG PS RA S DARA AHL PTG TPLD IDTASE 481SSQ DPQDSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV YND 521 RNLLD
S488C-Y518W (C4W2)	GS 401TT EDKIS RAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S Y RE TG PS RA S DARA AHL PTG TPLD IDTASE 481SSQ DPQDCRR S AD ALL RLQA MAGISE EE QGS DTDTPIV W ND 521 RNLLD
Y451W-S482C (C3W1)	GS 401TT EDKIS RAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S W RE TG PS RA S DARA AHL PTG TPLD IDTASE 481S CQ DPQ DSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV YND 521 RNLLD
S482C-Y518W (C3W2)	GS 401TT EDKIS RAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S Y RE TG PS RA S DARA AHL PTG TPLD IDTASE 481S CQ DPQ DSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV W ND 521 RNLLD
S407C-Y451W (C1W1)	GS 401TT EDKI CRAV GPRQAQVSFL HGDQSE NE LP RLGGK EDRR V 441 KQSRGEARE S W RE TG PS RA S DARA AHL PTG TPLD IDTASE 481SSQ DPQDSRR S AD ALL RLQA MAGISE EE QGS DTDTPIV YND 521 RNLLD

Table S2. Measurements, conditions, and fitting parameters obtained after global analysis of the transient absorption data for each N_{TAIL} variant studied. Only measurements for which the amplitude A_1 of the main decay (τ_1) contributed 35% or more to the total signal amplitude were included in the global fit (see Supporting Methods S2 for more details). $|i-j|$ refers to the sequence separation between the tryptophan and cysteine for each variant. Solvent conditions only include the variable conditions (see section Supporting Methods S2 for full solvent details). The blue highlighted columns denote the main decay τ_1 , used to obtain τ_{CW} values reported in main text, and shown in Figure 1D (from $\tau_{CW}^{-1} = \tau_1^{-1} - \tau_o^{-1}$, where τ_o is the main decay measured on variants containing only W, highlighted in blue). There is a $\pm 5\%$ error on the τ_{CW} values. The amplitudes (A_1, A_2, \dots, A_5) are the averages of the percentage contribution of the first measurement of each sample for each variant, before significant photodamage began. The τ_5 component with its negative percentage amplitude A_5 is the rise time.

Variant	$ i-j $	Solvent Conditions	Number of Measurements	τ_1 (μ s)	A_1 (%)	τ_2 (μ s)	A_2 (%)	τ_3 (μ s)	A_3 (%)	τ_4 (s)	A_4 (%)	τ_5 (ns)	A_5 (%)	τ_{CW} (μ s)
Y451W (W1)	--	pH 7.6 150mM NaCl	11	23.0	81	--	--	554	15	1	4	34	-22	--
Y518W (W2)	--	pH 7.6 150mM NaCl	14	39.4	78	2.43	6	605	13	1	3	30	-26	--
		pH 4.0 150mM NaCl	4	4.06	61	31.6	32	554	6	1	1	28	-30	--
S425C-Y451W (C2W1)	26	pH 7.6 150mM NaCl	10	2.73	57	27.7	29	710	7	1	7	31	-25	3.10
S488C-Y518W (C4W2)	30	pH 7.6 150mM NaCl	9	2.58	62	29.5	23	694	7	1	8	35	-24	2.76
		pH 4.0 150mM NaCl	8	1.84	68	28.8	21	849	6	1	6	29	-16	3.38
Y451W-S482C (C3W1)	31	pH 7.6 150mM NaCl	12	2.82	43	21.8	40	842	13	1	4	42	-24	3.21
		pH 4.0 150mM NaCl	7	1.88	71	31.9	18	812	4	1	7	19	-27	3.52
		pH 7.6 500mM NaCl	6	3.33	57	19.0	31	1120	9	1	4	28	-30	3.89
S482C-Y518W (C3W2)	36	pH 7.6 150mM NaCl	16	6.37	56	48.1	30	1020	9	1	5	44	-25	7.60
		pH 4.0 150mM NaCl	8	2.26	49	23.6	36	233	10	1	5	25	-26	5.11
		pH 7.6 500mM NaCl	3	5.08	52	43.0	33	1060	12	1	4	47	-26	5.83
S407C-Y451W (C1W1)	40	pH 7.6 150mM NaCl	14	5.15	45	33.0	39	930	10	1	6	44	-26	6.63
		pH 4.0 150mM NaCl	5	2.71	81	15.7	9	481	3	1	7	37	-21	8.19

Table S3. Force field comparison of wild type N_{TAIL} to experimental data. The table shows the deviation between measured observables (SAXS intensities, CD spectra, and NMR chemical shifts) and the same observables predicted from simulation ensembles of particular forcefields (FF, see Supporting Methods S8). The experimental radius of gyration was determined based on the measured SAXS curve using CRY SOL⁴⁸. Experimental SS composition was estimated from the CD spectrum using SESCA bayes, and the experimental CD deviation is calculated by comparing the predicted spectrum of the estimated SS composition to the measured CD spectrum. A structural ensemble of MEV-N_{TAIL} from the protein ensemble databank (PED) fitted to match measured chemical shifts is also shown for comparison. The exceptionally good (SD-HQ) and maximum acceptable (SD-max) deviations for the considered observables for a high-quality ensemble are shown on the bottom.

Summary System/ FF	SAXS data		CD spectra	C α shifts	C β shifts	SS content (%)		
	χ (A.U.)	Rg (nm)	RMSD (kMRE)	RMSD (ppm)	RMSD (ppm)	Alpha	Beta	Other
Exp.	0.00	2.75	0.87	0.00	0.00	3.3	13.6	83.1
PED*	0.61	3.28	1.46	0.29	0.40	7.3	2.5	90.4
C36M-OPC	0.63	2.99	1.37	0.54	0.36	3.3	6.5	90.2
A99B-disp	0.79	2.73	1.50	0.44	0.44	7.9	2.9	89.2
A03-ws	0.49	2.91	2.88	0.66	0.49	16.0	6.2	77.8
A99sb-ws	1.67	2.63	2.98	0.43	0.35	16.7	4.9	78.4
C36M	2.29	2.07	0.95	0.56	0.49	4.6	7.0	88.4
C36M-Tip4	2.64	1.92	1.08	0.61	0.55	3.9	7.8	88.3
C22*-Tip3S	2.57	1.95	0.65	0.51	0.46	6.1	6.6	87.2
C22*-OPC	1.93	2.19	0.75	0.58	0.51	5.4	7.1	87.5
SD-HQ	1.0	0.3	1.1	0.21	0.42	4.0	6.0	9.0
SD-max	2.0	0.6	3.1	0.75	0.62	8.0	12.0	18.0

Table S4. Force field comparison of NTAIL variants to experimental data. The table shows the deviation between measured observables (C-W quenching times due to PET relaxation and CD spectra) and the same observables predicted from simulation ensembles of particular force fields (FF, see Supporting Methods S8). The experimental C-W quenching times τ_{CW} are calculated from direct fits to the measured decay curves for each variant, and corrected for the natural decay τ_0 without the C quencher (as in Supporting Table S2). Experimental SS composition was estimated from the CD spectrum using SESCA bayes, and the experimental CD deviation is calculated by comparing the predicted spectrum of the estimated SS composition to the measured CD spectrum. The uncertainty of relaxation times shows the standard deviation of the posterior relaxation time distribution calculated from a Bayesian fit to the bootstrapped relaxation curves of all-atom MD simulation.

Summary System/FF	PET predictions		CD spectra	Compactness	SS content (%)			Ensemble size
	C3W2 (μ s)	C4W2 (μ s)	RMSD (kMRE)	Rg (nm)	Alpha	Beta	Other	
experiment	7.6	2.8	0.708	2.72	0.9	1.6	97.5	-
C36M-OPC	5.3 (3.3)	2.0 (0.5)	0.884	2.49	1.9	7.1	90.9	87785
A99SB-disp	13.3 (3.2)	4.7 (1.0)	3.431	2.26	11.7	3.2	85.2	103945

Table S5. Normalized pointwise mutual Information analysis of N_{TAIL} variant simulation trajectories from the CHARMM36m/OPC simulations. Prominent non-local interactions of Fig. 4 of main manuscript were analyzed to determine their correlation with PET donor-acceptor contact formation (both with C3W2 or C4W2 contacts). The residue numbers of the contact pair (residue 1 and residue 2), residue and region codes are also indicated in the table. Finally, the NPMI scores for conserved contacts found during our Coevolution analysis are also shown on the bottom.

S1 interactions				NPMI	
Residue 1	Residue 2	residue	Regions	C3W2	C4W2
413	525	R-D	A/E	-0.31	-0.44
441	511	K-E	B/E	-0.44	-0.25
444	507	R-E	B/E	-0.42	-0.24
497	520	R-D	D/E	-0.38	-0.19
420	495	L-L	A/D	-0.42	-0.56
472	501	M-L	C/D	-0.29	-0.31
500	518	A-W	D/E	-0.39	-0.30
S2 interactions:					
409	469	A-T	A/C	-0.02	0.16
413	476	R-D	A/C	0.30	0.03
441	449	K-E	B/B	0.03	0.28
447	501	A-M	B/D	-0.36	-0.42
497	453	R-E	B/D	-0.14	-0.01
S3 interactions:					
405	484	R-D	A/D	0.08	0.49
413	480	R-E	A/D	0.27	0.40
441	449	K-E	B/B	0.03	0.28
490	520	R-D	D/E	-0.04	0.70
S4 interactions:					
413	476	R-D	A/C	0.30	0.03
444	487	R-D	B/D	0.29	-0.37
463	513	R-E	C/E	0.42	-0.42
490	511	R-E	D/E	0.31	-0.21
465	516	A-I	C/E	0.45	-0.47
Coevolution:					
444	493	R-D	B/D	-0.06	-0.20
443	498	S-L	B/D	-0.04	-0.10
447	491	A-L	B/D	-0.03	-0.17
447	496	A-L	B/D	0.00	-0.14

Table S6. Normalized pointwise mutual Information analysis of N_{TAIL} variant simulation trajectories from the CHARMM36m/OPC simulations. Prominent non-local interactions of Fig. 4 of main manuscript were analyzed to determine their correlation with the helical conformations of N_{TAIL} MoRE. The helical conformations are defined by the fraction of MoRE residues in helical conformations; low (H-low) below 0.2, medium (H-med) between 0.2-0.5, and high (H-high) above 0.5. The residue numbers of the contact pair (residue 1 and residue 2), residue and region codes are also indicated in the table. Finally, the NPMI scores for prominent contacts involving the phosphoprotein XD binding site are also shown on the bottom. Contacts that have strong negative correlations with helix content in the MoRE are shown in red, ones that are positively correlated with medium helix content only are shown in blue, ones that are positively correlated with helix content are shown in green.

Residue 1	Residue 2	Res. type	Regions	NPMI		
#S1 specific:				H-low	H-med	H-high
413	525	R-D	A/E	0.07	-0.61	-0.42
441	511	K-E	B/E	0.10	-0.34	-0.05
444	507	R-E	B/E	0.11	-0.50	-0.29
497	520	R-D	D/E	0.04	-0.08	-0.50
420	495	L-L	A/D	-0.01	0.04	-0.12
472	501	M-L	C/D	0.06	-0.27	-0.41
500	518	A-W	D/E	0.07	-0.19	-0.51
#S2 specific:						
409	469	A-T	A/C	-0.08	0.17	-0.41
413	476	R-D	A/C	0.01	-0.50	-0.08
441	449	K-E	B/B	-0.01	0.77	0.16
447	501	A-M	B/D	-0.70	0.56	0.44
497	453	R-E	B/D	0.03	-0.30	-0.12
#S3 specific:						
405	484	R-D	A/D	-0.32	0.33	-0.44
413	480	R-E	A/D	0.30	0.03	0.08
441	449	K-E	B/B	-0.01	0.77	0.16
490	520	R-D	D/E	0.03	0.28	0.49
#S4 specific:						
413	476	R-D	A/C	0.06	-0.41	-0.36
444	487	R-D	B/D	0.06	-0.53	-0.35
463	513	R-E	C/E	-0.02	0.05	-0.40
490	511	R-E	D/E	0.04	-0.16	-0.32
465	516	A-I	C/E	0.08	-0.48	-0.45
#XD-bind:						
447	498	A-L	B/D	-0.62	0.57	0.47
474	497	D-R	B/D	0.06	-0.18	-0.56
491	514	S-T	D/E	-0.21	0.34	-0.49
498	523	L-L	D/E	0.03	-0.01	-0.56
501	516	M-I	D/E	0.13	-0.25	-0.48
501	522	M-N	D/E	-0.21	0.35	-0.52

Table S7. MoRE binding residue (on the left) interactions in the conformational states (S1-S4) identified in Fig. 3 of the main manuscript. The residue numbers of significant interaction partners, the highest contact formation probability of those partners, and region code for those residues are indicated in each state. We note that MoRE residues are located in region D of N_{TAIL} .

MoRE Residue	S1 partners Resid, Pmax, region	S2 partners Resid, Pmax, region	S3 partners Resid, Pmax, region	S4 partners Resid, Pmax, region
S491	----	----	514, 0.32, E 519-522, 0.18, E	509-512, 0.15, E
L495	417-420, 0.34, A	463-465, 0.08, C	-----	501-503, 0.71, D
L498	518-519, 0.73, E	447, 0.35, B 472-473, 0.14, C 461-463, 0.11, C	517-519, 0.17, E	-----
M501	515-517, 0.62, E 469-471, 0.41, C	445-447, 0.37, B 406-407, 0.25, A 470, 0.14, C	521-524, 0.34, E 515-517, 0.17, E	495-497, 0.74, D
W518	498-500, 0.75, D	433-438, 0.25, B	488, 0.58, D 498-499, 0.17, D 494, 0.25, D	465-467, 0.74, C 484-486, 0.70, D

Table S8. Further details on all-atom molecular dynamics (AA) simulations and force field comparison. For particular forcefields (FF, see Supporting Methods S8), the number replicate trajectories (Ntraj) started from different initial conformations, the trajectory length range (length) in microsecond, and the number of frames used in the final ensemble analyzed.

System/ FF	Ntraj	length (us)	No. of frames
N _{TAIL} wild type (with His6 tag)			
C36M-OPC	6	5-10	41230
A99B-disp	3	10	30000
A03-ws	3	2-3	7531
A99sb-ws	5	2-5	17460
C36M	5	1-3	10440
C36M-Tip4	3	1-3	7524
C22*-Tip3S	3	4-7	17233
C22*-OPC	3	4-7	16607
N _{TAIL} Variants (C3W2, C4W2)			
C36M-OPC	2X3	6-20	87790
A99B-disp	2X3	13-20	103961

Table S9. Coarse-grained (CG) molecular dynamics simulation details, including simulation conditions, simulation length, radius of gyration (R_g), and scaling exponent (ν) computed from the trajectories.

Conditions	length (equilibration)	R_g (nm)	scaling exponent (ν)
0.15M salt, pH=7.6, $k_H=0.2$ kcal/mol	5 μ s (100ns)	2.971 (3)	0.5382 (2)
0.5M salt	5 μ s (100ns)	3.035 (2)	0.5436 (2)
pH=4	5 μ s (100ns)	3.015 (4)	0.5414 (3)
helix+, $k_H=0.4$ kcal/mol	5 μ s (100ns)	3.014 (1)	0.5415 (3)
helix-, $k_H=0.1$ kcal/mol	5 μ s (100ns)	2.945 (3)	0.5360 (2)

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