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Self-aggregation and coaggregation of the p53 core fragment with its aggregation gatekeeper variant†

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Recent studies suggested that p53 aggregation can lead to loss-of-function (LoF), dominant-negative (DN) and gain-of-function (GoF) effects, with adverse cancer consequences. The p53 aggregation-nucleating ²⁵¹ILTIITL²⁵⁷ fragment is a key segment in wild-type p53 aggregation; however, an I254R mutation can prevent it. It was suggested that self-assembly of wild-type p53 and its cross-interaction with mutants differ from the classical amyloid nucleation-growth mechanism. Here, using replica exchange molecular dynamics (REMD) simulations, we studied the cross-interactions of this p53 core fragment and its aggregation rescue I254R mutant. We found that the core fragment displays strong aggregation propensity, whereas the gatekeeper I254R mutant tends to be disordered, consistent with experiments. Our cross-interaction results reveal that the wild-type p53 fragment promotes β -sheet formation of the I254R mutant by shifting the disordered mutant peptides into aggregating states. As a result, the system has similar oligomeric structures, inter-peptide interactions and free energy landscape as the wild type fragment does, revealing a prion-like process. We also found that in the cross-interaction system, the wild-type species has higher tendency to interact with the mutant than with itself. This phenomenon illustrates synergistic effects between the p53 ²⁵¹ILTIITL²⁵⁷ fragment and the mutant resembling prion cross-species propagation, cautioning against exploiting it in drug discovery.

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1. Introduction

Many neurodegenerative and metabolic diseases are associated with amyloidogenic proteins which have an intrinsic propensity to self-assemble into toxic oligomers and linear fibrils.^{1–6} Such proteins include the β -amyloid peptide (A β) in Alzheimer's disease,^{1,7} α -synuclein in Parkinson's disease⁸ and the islet amyloid polypeptide (IAPP) in type 2 diabetes mellitus.^{2,9} Recently, amyloid deposits of tumor suppressor protein p53 observed in cancer cells revealed surprising related pathogenic mechanisms between cancer and these neurodegenerative diseases.^{10,11}

p53 plays a central role in cellular life and death by controlling many pathways related to apoptosis, cell arrest, and DNA repair in response to stress.^{12,13} Besides its critical role as a tumor suppressor, p53 regulates hundreds of genes and is a guardian maintaining genome stability.^{12,14} The p53 monomer has three

domains: the N-terminal activation domain (residues 1–92), the C-terminal domain (residues 313–393), and the DNA-binding core domain (residues 93–312). In about half of all human tumors, p53 is mutated mainly in the DNA-binding core domain, which leads to either loss of its tumor suppressor function or gain of tumor promoting functions.^{15–20} All three p53 domains are able to aggregate, leading to a loss of function for p53.^{15,21,22} Recently, it was realized that prion-like aggregation of the mutant p53 may be related to the dysfunction of p53 in cancer.^{10,11,23,24} Mutant p53 aggregates induce not only the aggregation of wild type p53¹¹ but also the aggregation of its paralogs p63 and p73.^{25,26} The heterotetramers (mutant and wild type) show dominant-negative (DN) effects.^{27,28} Co-aggregation of p53 with its paralogs p63 and p73 might lead to interaction with new binding sites in the DNA and explain gain-of-function effects,²³ which may increase cancer aggressiveness and progression.²⁴ The destabilized, oncogenic p53 mutants are likely to increase the exposure of the hydrophobic core region,²⁹ thus making it prone to aggregation *via* self-assembly of the aggregation-nucleating stretch into an intermolecular β -sheet-like structure.¹¹ Recent experiments show that the strong hydrophobic region ²⁵¹ILTIITL²⁵⁷ plays an important role in the aggregation of p53. The segment not only drives full-length p53 aggregation *in vivo*³⁰ but also facilitates mutant p53 coaggregation with and inactivation of p63 and p73.^{24,31}

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A bioinformatics study showed that protein sequences often encode gatekeeper residues (charged residues and proline) against aggregation.³² Mutations of these gatekeeper residues, especially Arg, cause a significant rise in aggregation and polymorphism.³³ Consistently, it has been shown that introducing a gatekeeper residue (I254R) into the p53 aggregation core region²⁵¹ILTIITL²⁵⁷ can rescue p53 from aggregation.²³ A study of a related short peptide²⁵⁰PILTIITL²⁵⁷ indicated that multiple arginine substitutions (PIRTIITR and PIRTRRTL) also abolish peptide aggregation. Introducing various rescue residues into wild type or mutant p53 might be an intriguing potential strategy in p53 gene therapy, even though it may also be toxic. It has been shown that the loss of the aggregation propensity caused by the I254R mutation alleviated the interaction of aggregating mutants with wild-type p53;²³ however, it is unclear how the I254R mutation changes the aggregation behavior of p53 at the atomic level.

Exploring the cross interactions between two co-existing proteins with similar or different aggregation propensities is of paramount importance in understanding prion-like propagation in general^{34,34} and the triggering p53 dysfunction in particular. It has been suggested that the mechanism of wild type p53 aggregation and its cross-interaction with mutant p53 is different from the classical nucleation-growth mechanism of amyloid fibril formation.³¹ On the one hand, it appears that small seeds from already polymerized molecules are the initiators of aggregation,^{35,36} followed by a relatively rapid spread.³⁷ On the other hand, the mechanism corresponds more to trapping by cross-reaction and co-aggregation³¹ than to classical seeding and growth.^{38,39} In this study, we use all-atom replica exchange molecular dynamics (REMD) simulations to investigate the self-assembly of the p53 aggregation-nucleating segment²⁵¹ILTIITL²⁵⁷ and its cross interactions with its aggregation rescue I254R mutant. We found that the nucleation core fragment displays strong aggregation propensity, whereas its gatekeeper I254R mutant tends to form a disordered structure, consistent with experiments.²³ However, the co-aggregation of the aggregation prone fragment with the gatekeeper I254R mutant still displays a significant prion-like behavior. Wild type²⁵¹ILTIITL²⁵⁷ has a higher propensity to form the β -sheet and interact with the I254R mutant, resulting in trapping the disordered oligomers onto the ordered β -sheet-rich oligomeric structures. This observation questions the efficacy of such a mutant fragment as a drug in p53 cancers.

2. Materials and methods

Peptide systems

We studied the self-aggregation of a p53 wild type (WT) aggregation-nucleating fragment²⁵¹ILTIITL²⁵⁷ and its I254R gatekeeper mutant (MT), and the cross-interaction (CI) between the WT and MT fragments. The simulated WT, MT and CI systems, respectively, consist of six²⁵¹ILTIITL²⁵⁷ chains, six I254R mutant chains, and 3WT + 3MT chains. Selection of a hexamer of amyloid size in the simulation is common. Among these, studies of hexamers of amyloid-beta peptide (16–35) and its mutants revealed the influence of charge states on amyloid formation.⁴⁰

The²⁵¹ILTIITL²⁵⁷ peptide and its I254R mutant were both capped by the ACE (CH₃CO) group at the N-terminus and the NH₂ group at the C-terminus. The mutation of Ile254 to Arg (amyloid gatekeeper amino acid)³² introduces a positive charge.

Simulation methods

Three 250 ns REMD⁴¹ simulations were performed using the GROMACS-4.5.3 software package.⁴² We chose the AMBER99SB-ILDN force field⁴³ and carried out REMD simulations in the *NPT* ensemble at a pressure of 1 bar. The parameters used for our REMD simulations are widely used in numerous REMD studies.^{44–47} There are 48 replicas, each of 250 ns duration, at temperatures exponentially spaced between 307.86 and 421.82 K.⁴⁸ Thus the second replica has the physiological temperature of 310 K, which helps ensure more exchange chances with its neighboring replicas, thus accelerating sampling at the physiological temperature of 310 K.

Six²⁵¹ILTIITL²⁵⁷ chains, six I254R mutant chains, or 3WT + 3MT chains, with random conformations for each chain, were initially placed randomly in a $6.1 \times 6.1 \times 6.1$ nm³ box filled with TIP3P water molecules. There are 7322, 7296, and 7301 water molecules in the WT, MT and WT + MT systems, respectively. The peptide concentration in the three systems is 43.9 mM. The attempt swap time between two neighboring replicas is 2 ps. The acceptance ratio is $\sim 24\%$, as shown in the ESI† (Fig. S1). A large number of REMD simulation studies demonstrate that an acceptance ratio of 20–30% is good.^{34,44,49–53} Periodic boundary conditions were applied in all three directions. Constraints were applied to all-bond lengths using the SETTLE algorithm and the LINCS method for the peptides, allowing an integration time step of 2 fs. The protein and non-protein (water and counterions) groups were separately coupled to an external heat bath with a relaxation time of 0.1 ps using a velocity rescaling coupling method.⁵⁴ The pressure was kept at 1 bar using the Parrinello–Rahman method⁵⁵ with a coupling time constant of 1.0 ps. A cutoff of 1.4 nm was used for van der Waals interactions. The particle mesh Ewald (PME) method⁵⁶ with a real space cutoff of 1.0 nm was used for electrostatic interactions. The coordinates were saved every 2 ps.

Analysis methods

Trajectory analysis was performed using our in-house-developed codes and the facilities implemented in the GROMACS-4.5.3 software package.⁴² We discarded the first 150 ns data for each REMD run to remove the bias of the initial states. Therefore, the structural properties of each system were based on the simulation data generated in the last 100 ns. The secondary structure of the peptide was identified using the DSSP program. The tertiary structure analysis was performed by combining a C $_{\alpha}$ root-mean-square deviation (RMSD) cluster analysis method with the percentages of various sizes of the β -sheet. In this study, we performed a chain-independent RMSD calculation because all the chains are topologically identical, as done previously by Li *et al.*⁵⁷ In this study, for a single pdb file generated from the REMD trajectory, we calculated its RMSD using 6! ($6 \times 5 \times 4 \times 3 \times 2 \times 1$) different coordinate files (the *x*, *y*, *z* values of all the atoms are

unchanged, only the numbering order of the atoms is changed). That is to say, we got 6! RMSDs. The smallest RMSD, *i.e.* the chain-independent RMSD, was taken for structure clustering.

The size of a β -sheet is the number of β -strands in an n -stranded β -sheet, *e.g.*, the β -sheet size of a three-stranded β -sheet is three. Two peptide chains are considered to form a β -sheet if (i) at least two consecutive residues in each chain visit the β -sheet state and (ii) the two chains form at least two backbone hydrogen bonds (H-bonds). One H-bond is taken as formed if the N...O distance is less than 0.35 nm and the N-H...O angle is greater than 150° . We used a topological parameter, connectivity length (CL),⁵⁸ to describe the orderliness of the peptide aggregates. CL is defined as the sum over the square root of the β -sheet size and the number of disordered chains in each conformation. For example, the CL of a hexamer consisting of a 4-stranded β -sheet and two random chains is $\sqrt{4} + \sqrt{1} + \sqrt{1} = 4$. Thus, the larger the connectivity length, the more disordered the hexamer is. The VMD program⁵⁹ was used for graphical structure analysis.

The solvation extent of the peptide backbone was estimated by the number of water molecules that are within 0.35 nm from the peptide backbone. The interpeptide interactions were analyzed by the residue-residue (including main-chain-main-chain (MC-MC) and side-chain-side-chain (SC-SC)) contact probabilities. Here, a contact is defined when the aliphatic carbon atoms of two nonsequential side chains (or main chains) come within 0.54 nm or any other atoms of two nonsequential side chains (or main chains) lie within 0.46 nm. The free energy surface of each system was constructed using $-RT \ln H(x, y)$, where $H(x, y)$ is the histogram of two selected reaction coordinates, H-bond number and R_g . Here, H-bond number and R_g denote the total number of hydrogen bonds (including intra- and intermolecular H-bonds) and the radius of gyration of the hexamer, respectively. The value of the minimum free energy is defined as zero. We also estimated the total energy and binding free energy using the molecular mechanics/linear Poisson-Boltzmann surface area (MM/PBSA) method implemented in the AMBER package.⁶⁰ In MM/PBSA, the total energy is calculated as: $E_{\text{total}} = E_{\text{angle}} + E_{\text{dih}} + E_{\text{elec}} + E_{\text{vdW}} + E_{\text{polar}} + E_{\text{nonpolar}}$. Here, E_{angle} , E_{dih} , E_{vdW} and E_{elec} are, respectively, the angle, the dihedral angle, the van der Waals (vdW) and the electrostatic interaction energies in vacuum. The $G_{\text{polar}} + G_{\text{nonpolar}}$ is the solvation free energy that is required to transfer a solute from vacuum into the solvent, where, G_{polar} and G_{nonpolar} are the electrostatic and non-electrostatic contributions to the solvation free energy, respectively. G_{polar} is calculated using the PB model and G_{nonpolar} is estimated using the solvent accessible surface area (SASA). The binding free energy ($\Delta G_{\text{binding}}$) between a ligand and a receptor is also calculated as: $\Delta G_{\text{binding}} = \Delta E_{\text{vdW}} + \Delta E_{\text{elec}} + \Delta G_{\text{polar}} + \Delta G_{\text{nonpolar}}$.

3. Results and discussion

To examine the convergence of the REMD simulations, we first checked the time evolution of the replica initiated from 310 K. Fig. S2A (ESI[†]) shows that this replica visited sufficiently the

whole temperature space within a 250 ns REMD simulation, indicating that the replica was not trapped in one single temperature. Other replicas display similar sampling behavior (data not shown). The convergences of the three REMD runs were further verified by comparing the β -sheet probability of each residue and the probability density function (PDF) of the end-to-end distance of each chain within two different time intervals using the 150–200 ns and 200–250 ns data. As shown in Fig. S2B (ESI[†]), the residue-based β -sheet probabilities are almost the same between the two time periods. The distributions of the end-to-end distance within the two independent time intervals overlap very well for the three systems (Fig. S2C, ESI[†]). These data suggest that our REMD simulations for the three systems are reasonably converged within 250 ns. Unless specified, all the REMD simulation results presented below are based on the last 100 ns ($t = 150$ –250 ns) simulation data generated at 310 K.

I254R gatekeeper mutant decreases the β -sheet propensity of the P53²⁵¹ILTIITL²⁵⁷ peptide

We first examined the secondary structure properties of the wild type ²⁵¹ILTIITL²⁵⁷ peptide. As can be seen in Fig. 1A, the hydrogen bonding counts steadily increase with the simulation time. Our highly efficient REMD sampling provides much better results than a previous 65 ns simulation of a system consisting of 24 ²⁵⁰PILTIITL²⁵⁷ peptide chains, where only about 20 hydrogen bonds were obtained.³⁰ Fig. 1B shows the percentage of secondary structure, including the coil, β -sheet, β -bridge, bend, turn, α -helix and 3-helix. The equilibrated ²⁵¹ILTIITL²⁵⁷ hexamer contains 36.0% β -sheet. When the two terminal residues I251 and L257 were not considered, the β -sheet content of the peptide reaches as high as 40–60%. As expected, introducing I254R gatekeeper mutation changes the aggregation propensity of the ²⁵¹ILTIITL²⁵⁷ peptide. Compared with the aggregation prone wild type sequence, the substitution of the hydrophobic isoleucine residue with a positively charged arginine considerably decreases the average β -sheet content from 36.0% to 25.7% and increases the average coil content from 48.0% to 54.7%. The residue-based β -sheet probability (Fig. 1C) shows that the β -sheet probability for all residues drops from 36–61% for the wild type to 24–47% for the mutant. While it seems that the 47% β -sheet probability is still not too low, the size of the β -sheet decreases and the dimer is dominant (Fig. 2A), indicating that the mutation suppresses the aggregation of the peptide.

The majority of the β -sheets are three β -stranded β -sheets, but large sizes of β -sheets consisting of 4–6 β -strands were also observed (Fig. 2A). Fig. 2B shows the distribution of the orientations of two neighboring β -strands in all β -sheet sizes. For the WT system, the two peaks located at 25° and 150° correspond to, respectively, parallel and antiparallel alignments of β -strands, with the antiparallel alignment being dominant. Similar results are seen for the MT and CI systems. These results indicate that the peptide chains in the three systems have a preference to adopt antiparallel β -stranded sheets.

The ordering of the WT, MT and CI hexamers is examined by monitoring the time evolution of the connectivity length (CL) (Fig. 2C). The CLs of the hexamers of the WT, MT and CI

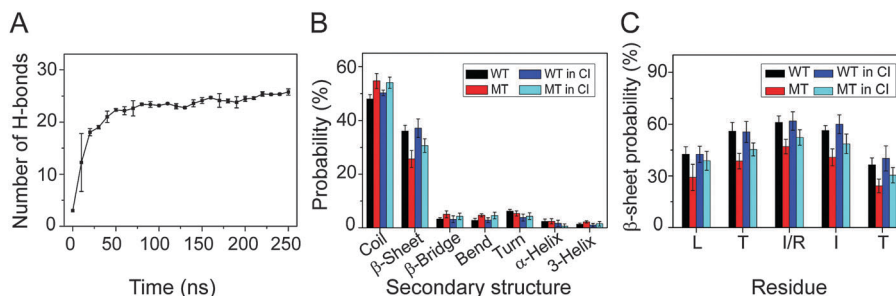


Fig. 1 Secondary structure analyses. (A) The number of hydrogen bonds as a function of simulation time in WT hexameric systems. Hydrogen bonding counts increase to 24 after 150 ns, showing highly efficient REMD sampling. (B and C) Analysis of the secondary structure of the WT, MT, and CI hexameric systems. The positively charged arginine considerably decreases the average β -sheet content and increases the average coil content. The average β -sheet probabilities of the WT increase slightly and those of MT increase significantly in the CI system.

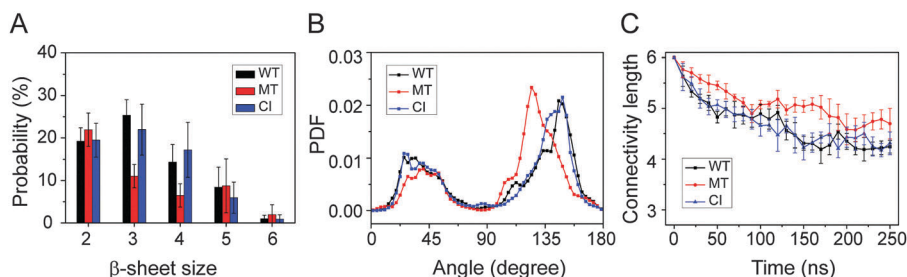


Fig. 2 Characteristic analyses of interpeptide β -sheets. (A) Analysis of the β -sheet size distribution. The majority of β -sheets are three β -stranded β -sheets in WT and CI systems, while the size of β -sheet decreases and the dimer is dominant in MT systems. (B) The distribution of the angle between two β -sheet-forming chains in WT, MT, and CI systems. Two peaks located at 25° and 150° indicate that the antiparallel alignment is dominant for three systems. PDF is the probability density function. (C) The time evolution of the connectivity length for the WT, MT, and CI hexamers. The smaller CL values of the WT and CI systems than those of the MT systems reveal the much more ordered nature of the WT and CI hexamers.

systems decrease rapidly from the initial value of 6.0 to, respectively, 4.2, 5.0, and 4.2 within the first 150 ns of the simulations and fluctuate around these three values during $t = 150\text{--}250$ ns. The CLs of CI hexamers during the last 100 ns is very close to those of the WT hexamer, indicating that the CI system displays similar aggregation properties as the WT system. The larger CL value of the MT system as compared to those of the WT and CI systems reveals much more disordered nature of MT hexamers.

Experiments show that high temperature can make wild type p53 lose its function.^{61–63} Thus, we further investigated the β -sheet probability as a function of temperature for WT, MT, and CI hexamers (Fig. 3). For both MT and CI systems, the probability of the β sheet decreases monotonically with increasing temperature, whereas the β -sheet probability of the WT increases slightly from 36.1% (308 K) to 37.1% (325 K) and then decreases, also revealing that the WT sequence prefers to aggregate at higher temperatures.

We then calculated the average number of water molecules within 0.35 nm from the backbone atoms of each residue to monitor the extent of solvation of the peptide backbone (Fig. 4). In the wild type system, the number of water molecules drops successively from 2.1 to 1.1. The backbone atoms of residues T256 (~ 2.0) and L257 (~ 2.7) are more solvent-exposed than other residues. The three residues (TII) in the middle of the amino acid sequence of the WT peptide are well protected from the solvent. Due to the smaller oligomer size and lower β -sheet

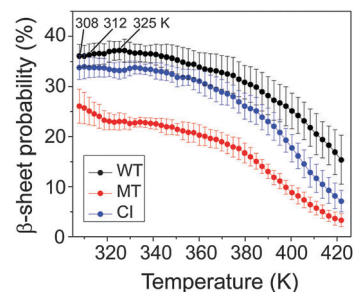


Fig. 3 β -Sheet probability as a function of temperature for WT, MT, and CI hexamers. β -Sheet probability is 36.1% at 308 K, 36.2% at 312 K and 37.1% at 325 K, revealing that the WT sequence prefers to aggregate at higher temperature.

content, the residue has higher solvation (Fig. 4, red bar). Compared with the wild type system, the backbone of all residues of the mutant hexamer is more solvent-exposed, except I251, indicating that the I254R mutation affects not only itself, but also the entire peptide.

We also calculated the interpeptide MC–MC and SC–SC contact probabilities between all pairs of residues for WT, MT, and CI hexamers. As seen from Fig. 5, while the wild type hexamers have balanced both MC–MC interactions and SC–SC interactions, the I254R gatekeeper mutant hexamers lost most of their MC–MC interactions. Interestingly, the nearby Ile255

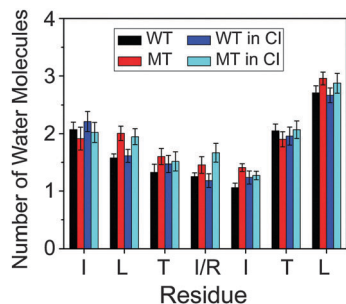


Fig. 4 The average number of water molecules within 0.35 nm of the mainchain atom of each residue. The figure shows that C-termini are more solvent-exposed and the backbone of all residues of the mutant hexamer is more solvent-exposed.

residue now plays an important role in peptide associations. For the wild type hexamer system (Fig. 5A), compared to other residue pairs, the T253–T253 (with a contact probability of 13.1%), T253–I254 (13.3%), I254–I254 (12.6%) and T253–I255 (13.9%) pairs display high MC–MC contact probabilities. The relatively high MC–MC contact probabilities along the left diagonal of the MC–MC contact map in Fig. 5A indicate that WT peptides are aligned predominantly in the antiparallel orientation. For SC–SC interactions, the isoleucine–isoleucine pairs have the highest contact probabilities of 20.9% (I254–I254 pair), 20.1% (I254–I255 pair) and 16.5% (I255–I255 pair), reflecting strong hydrophobic interactions. When I254 is substituted by proline, the MC–MC contact probabilities (see Fig. 5B) of T253–T253, T253–I254, I254–I254 and T253–I255 pairs dramatically drop to 9.3%, 11.0%, 10.5% and 9.8%, respectively. This reduced MC–MC interaction is associated with increasing interactions between MC atoms and water molecules. For SC–SC contact probabilities (see Fig. 5E), arginine shows weak interactions

with other residues because of its positive charged side chain, leading to high contact probabilities of I255–I255 (32.1%) and T253–I255 (21.6%) pairs. The error bars of MC–MC contact probabilities in each probability map were calculated using the 150–200 ns and 200–250 ns data. Our calculations show that all the error bars are very small compared to the contact probabilities (Tables S1–S6, ESI†). The small errors reflect the reliability of the REMD simulations.

Synergistic mutual β -sheet promotion between the p53²⁵¹ILTIITL²⁵⁷ fragment and its I254R gatekeeper mutant

In order to study the cross-interaction between the aggregation prone ²⁵¹ILTIITL²⁵⁷ fragment and its I254R gatekeeper mutant, we simulated the system using a mixture of three wild type and three mutant peptides. Surprisingly, we found that there is a synergistic effect of the cross-interaction between the wild type and its I254R mutant. The β -sheet probability of the mutant in the CI hetero-hexamer is promoted to 30.6%, which is about 4.9% higher than that of the MT homo-hexamer. Even the average β -sheet probability of the WT species in the CI system is about 1.1% higher than that in the WT homo-hexamer (Fig. 1B). Comparing the WT trimer in the CI system with the WT homo-hexamer in the WT system, we found that the average β -sheet probabilities of L255 and T256 increase slightly ($\sim 3.6\%$ and $\sim 3.7\%$). Meanwhile, the probabilities of the turn, α -helix and 3-helix of both WT and MT peptides in the CI system slightly decrease. These results indicate that the interaction between the WT and MT peptides in the CI system can induce the secondary structure transition from the turn, α -helix and 3-helix to a β -sheet structure for the I254R mutant. In the CI system (Fig. 1C), β -sheet probabilities of all residues, except terminal residues of the MT peptide, distinctly increase (5.25–9.7%).

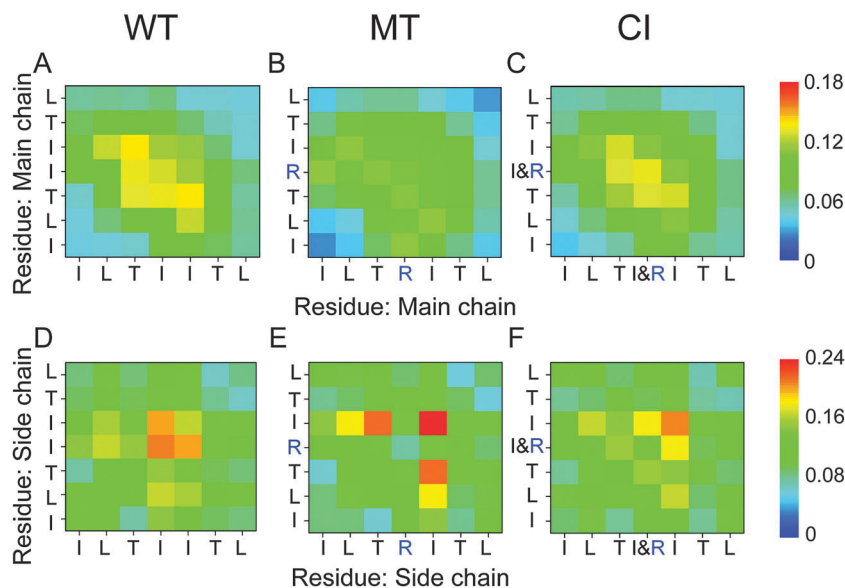


Fig. 5 Contact probability map. Main-chain–main-chain (MC–MC) and side-chain–side-chain (SC–SC) contact probabilities averaged over the 150–250 ns REMD generated conformations for the hexamers of wild type (A and D), I254R mutant (B and E), and the CI (C and F) systems at 310 K. Strong hydrophobic interactions between the isoleucines essentially stabilize oligomeric structures in the WT and CI systems.

These data show that wild type species can significantly promote β -sheet formation of its I254R mutant. In the CI system, even though the I254R mutation rendered higher contact numbers with water molecules (Fig. 4), the hydrophobic I254 in the wild type peptide still provides effective protection of backbone atoms from solvation. This trend can also be seen from the contact probability map (Fig. 5). In the CI system, the MC–MC contact probabilities of T253–T253 (11.6%), T253–I254 (12.9%), I254–I254 (13.4%) and T253–I255 (12.6%) pairs show similarly strong MC–MC interactions as the wild type system (Fig. 5C). The SC–SC contact probabilities of I/R254–I/R254 (13.6%), I/R254–I255 (18.2%) and I255–I255 (20.7%) also show strong hydrophobic interactions between the isoleucines (Fig. 5F). The maps show that the wild type peptides weaken the negative I254R mutational effect on the SC–SC interactions in the CI system. These data reveal that the oligomeric structures in the WT and CI systems are essentially stabilized by MC–MC interactions between threonine and isoleucine residues and SC–SC interactions between isoleucine and isoleucine residues.

After characterizing the secondary structure properties of WT, MT and CI hexamers, we investigated their three dimensional conformational states by first performing a RMSD-based cluster analysis for each system. With a C_{α} -RMSD cutoff of 0.3 nm, the conformations of the WT hexamer, MT hexamer and CI hexamer at 310 K were separated into 293 clusters, 443 clusters and 392 clusters, respectively. The representative conformations of the first eight most-populated clusters are shown in Fig. S3 (ESI[†]). These clusters represent 37.5%, 27.6% and 27.4% of all conformations of the WT, MT and CI hexamers, respectively. The WT homo-hexamer and the CI hetero-hexamer contain more ordered β -sheet rich conformations than the MT homo-hexamer. For example, the fourth and sixth clusters of the WT hexamer contain bi-layer β -sheet structures (three-stranded + two-stranded β -sheets) with mainly an antiparallel alignment. In the mutant system, the first cluster (with a probability of 7.6%) and the eighth cluster (1.9%) contain, respectively, five-stranded and six-stranded open β -barrels in which the side chains of positively charged arginine residues are solvent-exposed.

For the CI hexamer, similar to the WT system, bilayer β -sheet structures, such as those in the first cluster (5.3%) and the fifth cluster (3.0%), are populated. Mono-layer and bi-layer β -sheet structures are fibril-competent states, while disordered aggregates and β -barrels are difficult to form fibrils.

Our various analyses have shown that the aggregation prone $^{251}\text{ILTIITL}^{257}$ fragment effectively recruits the I254R gatekeeper mutant into its aggregation pattern. We found that the oligomeric structures of the WT homo-hexamer and the CI hetero-hexamer share fairly similar characteristics of ordered bilayer antiparallel β -sheet structures (Fig. 2B).

The CI system displays similar free energy landscape as the wild type

To have an overall view of the conformational distribution of hexamers of the three systems, we constructed the 2D free energy surface in Fig. 6 using $-RT \ln H$ (H : H-bond number, R_g) as described in Analysis methods. The locations of representative structures are labeled on the PMF plot. For the wild type hexamer, the black, light blue and green lines point to the location of the first cluster (monolayer structure, with a probability of 10.1%), the fourth cluster (bilayer β -sheet structure, 4.1%) and the sixth cluster (bilayer structure, 2.8%). For the I254R mutant system, we highlight the location of the first cluster (five-stranded open β -barrel, 7.6%), the fourth cluster (disordered coil-rich aggregate, 3.5%) and the fifth cluster (disordered coil-rich hexamer, 2.2%). For the CI hexamer, the locations of the first cluster (bilayer β -sheet structure 5.3%), the fourth cluster (four-stranded mono-layer structure 3.0%) and the fifth cluster (monolayer structure, 3.0%) are labeled. The distributions of the conformations of the WT and CI hexamers are quite concentrated on the free energy surface, while they are scattered in the mutant system. The global minimum energy basins of the WT, MT, and CI hexamers are located at (number of H-bonds, R_g) values of (25, 1.0), (23, 1.4) and (27, 0.97), respectively. Compared with the mutant hexamer, the wild type hexamer and the cross-interaction hexamer have an increased number of H-bonds and a

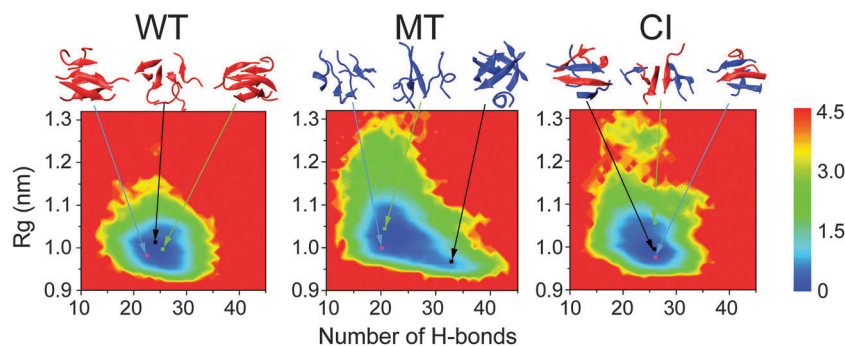


Fig. 6 Free energy surfaces (in kcal mol^{-1}) of the wild type hexamer, mutant hexamer and cross-interaction hexamer, respectively, at 310 K as functions of the total number of intra- and intermolecular H-bonds (number of H-bonds) and the radius of gyration (R_g). Their minimum energy basins are located at different (number of H-bonds, R_g) values of (25, 1.0 nm), (23, 1.4 nm) and (27, 0.97 nm), respectively, implying that the wild type can induce the mutant to more compact, ordered aggregates. Representative structures are also given, along with their probabilities: 10.1% (black line), 4.1% (light blue line) and 2.8% (green line) at WT systems; 7.6% (black line), 3.5% (light blue line) and 2.2% (green line) in MT systems; 5.3% (black line), 4.9% (light blue line) and 3.6% (green line) in CI systems.

decreased value of R_g , implying that the wild type can induce the mutant to more compact, ordered aggregates.

The origin of the synergistic effect between the wild type and the I254R mutant

To further investigate the molecular mechanism of the synergistic interaction between the wild type and the I254R mutant peptides, we calculated the contact probabilities of three different combinations of peptides (WT–WT, MT–MT and WT–MT) in the CI system. As shown in Fig. 7, WT–WT and MT–MT species still retain the same characteristics of MC–MC and SC–SC interactions as the WT and MT hexameric systems; however, the WT–MT species have the strongest MC–MC (Fig. 7C) and SC–SC (Fig. 7F) interactions, which suggest that the WT can cross-interact with its mutant. In the WT–MT peptide pairs, T253–R254 and I254–T253 pairs of main-chains have the highest probabilities, also showing preference for an antiparallel organization. As to side-chain interactions, the hydrophobic isoleucine plays an important role in the cross-interaction between WT and MT: 26.5% in the I254–I255 pair and 21.1% in the I255–I255. The error bars of SC–SC contact probabilities in each probability map were calculated using the 150–200 ns and 200–250 ns data. Our calculations show that all the error bars are very small compared to the contact probabilities (Tables S7–S12, ESI†). The small

errors reflect the reliability and good convergence of the REMD simulations.

To estimate the contribution of non-bonded interactions and reveal the prion-like behavior of the core fragment, we also calculated the inter-peptide binding free energy using the MM/PBSA method as shown in Tables 1 and 2. As can be seen in Table 1, both WT–WT and WT–MT peptide pairs have attractive interaction energies, while the I254R mutant has repulsive energy. The mixed, cross-interacting hexamer overcomes the repulsive energy and even has a small thermodynamic preference.

In Table 2 we calculated the binding free energy between three randomly selected peptides and three other peptide chains in the hexamers for both wild type and mutant systems. The positive value of electrostatic energy including both potential energy in vacuum (ΔE_{elec}) and solvation energy (ΔG_{polar}) shows that introducing arginine, a positively charged residue, is unfavorable for aggregation. However, the WT–MT binding in the cross-interaction system weakens the negative effect. The large negative values of van der Waals energy (ΔE_{vdw}) and nonpolar solvation energy ($\Delta G_{\text{nonpolar}}$) reveal that the strong hydrophobicity of the peptides is an intrinsic driving force for aggregation. A comparison of the total binding free energies of the three models, when wild type and mutant

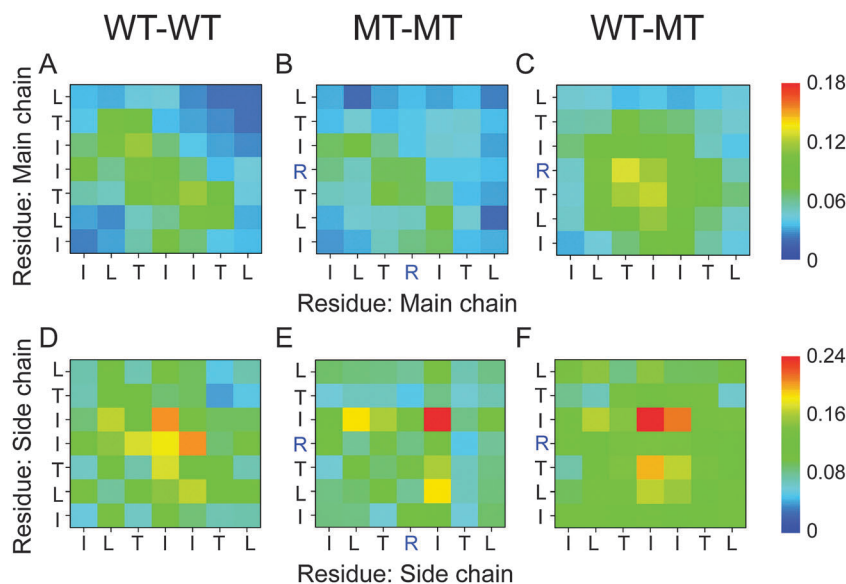


Fig. 7 Contact probability map. Main-chain–main-chain (MC–MC) and side-chain–side-chain (SC–SC) contact probabilities in the cross-interaction system, WT–WT (A and D), MT–MT (B and E) and WT–MT (C and F). The wild-type species has a higher tendency to interact with the mutant than with itself.

Table 1 Partition of total energies (in kcal mol⁻¹). The mixed, cross-interacting hexamer overcomes the repulsive energy and even has a small thermodynamic preference

Systems	E_{angle}	E_{dih}	E_{elec}	E_{vdw}	G_{polar}	G_{nonpolar}	E_{total}
WT	2126.1 ± 4.9	518.3 ± 2.0	-2651.9 ± 10.2	-274.6 ± 3.5	-221.7 ± 5.4	25.0 ± 0.5	-478.8 ± 9.7
MT	2136.4 ± 4.9	523.4 ± 2.3	-1561.0 ± 19.9	-285.2 ± 5.5	-807.6 ± 15.0	27.3 ± 0.7	33.3 ± 12.0
CI	2131.5 ± 5.7	520.7 ± 4.9	-2198.8 ± 9.2	-277.6 ± 3.4	-426.3 ± 5.9	26.4 ± 0.5	-224.1 ± 9.0
ΔH	0.3 ± 0.9	-0.2 ± 0.6	-92.4 ± 2.4	2.3 ± 0.7	88.4 ± 1.7	0.3 ± 0.1	-1.3 ± 1.8

Table 2 Binding energies (in kcal mol⁻¹). The wild type species have a higher tendency to interact with their charged mutant due to their strongest binding energy

	ΔE_{vdw}	ΔE_{elec}	ΔG_{polar}	$\Delta G_{\text{nonpolar}}$	$\Delta G_{\text{binding}}$
WT-WT	-87.9 ± 8.8	-72.4 ± 12.4	85.6 ± 12.2	-13.1 ± 1.3	-87.8 ± 10.3
MT-MT	-97.0 ± 10.3	82.6 ± 18.6	-55.3 ± 17.5	-14.9 ± 1.6	-84.6 ± 12.1
WT-MT	-101.3 ± 8.1	-102.1 ± 11.0	120.9 ± 11.4	-15.4 ± 1.3	-97.9 ± 8.9

peptides coexist, shows that the wild type species have a higher tendency to interact with its charged mutant due to their strongest binding energy.

Recently, it has been shown experimentally that proximally immobilized ions can modulate hydrophobic interactions of conformationally stable β -peptides.⁶⁴ Our results of the synergistic interaction between the wild type and the I254R mutant peptide illustrate a similar physical origin.

4. Conclusions

We have investigated the self-assembly of the aggregation-nucleating fragment ²⁵¹ILTIITL²⁵⁷ of p53 protein and its I254R gatekeeper mutant, and their cross-interaction by performing three 250 ns atomistic REMD simulations starting from a random state. Structural analyses of the three systems show that both the wild type and the cross-interaction systems mainly form a bilayer organization with three- and four-stranded antiparallel β -sheets with parallel β -strands in each sheet. The I254R mutation introduces a positive charge that is sufficient to suppress the aggregation tendency of the wild type peptides. The β -sheet rich structures are intrinsically stabilized by strong hydrophobic interactions, especially between the two isoleucines. Previously, it has been suggested that evolution tends to avoid aggregation from isoleucines.⁶⁵ Strikingly, we found that the wild type ²⁵¹ILTIITL²⁵⁷ has a higher propensity to form the β -sheet and interact with the I254R mutant, resulting in trapping the disordered peptides and extending the β -sheet rich ordered structure.

Our simulations provide insights into the difference between the full length (or core domain) p53 aggregates and the classical nucleation-growth of amyloid fibrils.²⁰ Experimental studies of p53 aggregation in cells led to the suggestion that the I254R gatekeeper mutation alleviates the interaction of aggregating mutants with wild-type p53.²³ At first sight, this conflicts with our observations of cross-interactions between the p53 core fragment and its I254R mutant. However, the disagreement raises the question of what is the difference between the aggregation behaviors of an isolated peptide and when the fragment is highly buried in the protein core. Nature has evolved a way to seal aggregation prone peptide fragments in protein cores.²¹ Mutations and unfolding perturbations often lead to the exposure of amyloidogenic fragments resulting in aggregation-related diseases. p53 can constitute such an example. The first step in p53 aggregation is unfolding and exposing core fragments like ²⁵¹ILTIITL²⁵⁷. Wild type aggregates can trap the already unfolded mutant p53, recruiting it into the aggregates.²⁰ Similarly, the A β peptides can seed the tau protein also by stretching tau and

exposing its hydrophobic core.³⁹ This is likely why the aggregation-disabled I254R mutant cannot trigger wild-type p53 aggregation.¹²

If we label the p53 core fragments like ²⁵¹ILTIITL²⁵⁷ as generic “aggregating” species, and its gatekeeper I254R mutant as “non-aggregating” species, our study well demonstrates that the “aggregating” species can trigger and recruit “non-aggregating” species in a prion-like propagation fashion. Thus, when designing peptides to inhibit the protein aggregation, the designed peptide should not only be aggregation free on its own, it should also have the ability to resist being recruited and merged into an “aggregating” species. Our results indicated that the gatekeeper mutation (I254R) does not completely eliminate the amyloid prone nature of the sequence ILTRITL. The mutation most likely eliminates or delays the kinetic formation of the nucleation core. Thus the I254R mutant could have the entropy barrier, in a similar way as some prion sequences with medium amyloid propensity.⁶⁶ With the seeding of wild type p53 fragment, the wt p53 nucleus overcomes the entropy barrier of I254R mutant. The situation is also similar to the effects of an amyloid stretch within a long sequence to trigger and recruit otherwise non-amyloidogenic sequence into amyloid filaments.⁶⁷ Nucleation and seeding of prion protein aggregates are strongly influenced by dynamic interactions between the aggregate core forming domain and its flanking regions.⁶⁸ Most prions propagate among prion-like proteins with cross-seeding barriers.⁶⁹ Still, it was recently found that prion resistant species (pig) can also be affected by bovine prions.⁷⁰ We have just become aware that a designed peptide inhibitor of p53 aggregation was actually recently shown to rescue p53 tumor suppression in ovarian carcinomas.⁷¹ We are thrilled that the peptide selected is ²⁵²LTIITL²⁵⁸, with the I254R mutation. Here we have selected ²⁵¹ILTIITL²⁵⁷ – the same peptide however with a single shifted residue, and lacking this mutation, making our peptide more hydrophobic. In the cancer cell paper, an N-terminal poly-arginine cell-penetrating tag converts R248Q p53 into soluble wild-type-like p53 and induces the cancer cell death. In the future, it will be interesting to investigate ²⁵²LTIITL²⁵⁸ and its I254R mutant and test the effect of the added polyarginine tag. We thank Dr. Soragni for letting us know of their exciting work and for discussion.

Conflict of interests

The authors declare no competing financial interest.

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