

## Supplementary Material for **Random heteropolymers preserve protein function in foreign environments**

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**Other Supplementary Material for this manuscript includes the following:**  
(available at [www.sciencemag.org/content/359/6381/1239/suppl/DC1](http://www.sciencemag.org/content/359/6381/1239/suppl/DC1))

Movies S1 and S2

## MATERIALS AND METHODS

### Section S1. Protein Analysis

#### **S1.1 Analysis of protein surface using PyMOL 1.5.0.3**

Horseradish peroxidase (HRP), glucose oxidase (GOx), green fluorescent protein (GFP), and  $\alpha$ -chymotrypsin ( $\alpha$ -CT) crystal structures – 1H55, 1CF3, 2HRW, and 1YPH respectively – were obtained from the Protein Data Bank. PyMOL was used to simplify protein surfaces as non-charged hydrophilic, hydrophobic, cationic, or anionic patches. Patches were approximated, to the best of our abilities, as circles with distinct diameters, and distances between chemically-like patches were measured using straight lines, from the center of one patch to the center of a neighboring patch. Measurements were binned to the closest whole number, and frequencies were normalized by the frequency of the mode. The histograms of diameter and distance of charged patches are plotted in **Fig. S1**.

#### **S1.2 Analysis of protein sequence based on binary hydrophobicity**

The program reads in an amino acid sequence from a protein database (PDB) file, and assigns each amino acid a value of 0 or 1. All amino acids more hydrophilic than glycine are assigned a value of 0, while all amino acids more hydrophobic than glycine including glycine are assigned a value of 1. The scale is a binary adaptation of the scale established by Monera et al. (11). Then, starting from the beginning of the chain, it counts the distance between two hydrophilic residues to determine the size of that hydrophobic block. This is repeated until the whole chain has been analyzed. The example histogram of block size distribution of horseradish peroxide (HRP) is shown in **Fig. S2**.

## Section S2. RHP Synthesis and Characterization

### S2.1. Synthesis of RHP by RAFT polymerization

Methyl methacrylate (**1**, 99%, Aldrich), ethylene glycol methyl ether methacrylate (**2**, 99%,  $M_n = 500$  g/mol, Aldrich), and 2-ethylhexyl methacrylate (**3**, 98%, Aldrich) were passed through a short column of neutral alumina to remove inhibitor before use. 3-Sulfopropyl methacrylate potassium salt (**4**, 98%, Aldrich) and solvents with the purest grades (Aldrich) were used as received. *S*-methoxycarbonylphenylmethyl dithiobenzoate (**5**) RAFT chain transfer agent was synthesized as described previously (32).

A solution of **1** (513 mg, 5.13 mmol), **2** (1280 mg, 2.56 mmol), **3** (403 mg, 2.03 mmol), **4** (123 mg, 0.50 mmol), and **5** (11 mg, 0.04 mmol) in *N,N*-dimethylformamide (3.2 mL) was degassed by three freeze-pump-thaw cycles before being sealed off under vacuum. The reaction scheme can be found in **Fig. S3**. After 20 h at 60 °C, the polymerization media was diluted in dichloromethane, precipitated in diethyl ether and dried under vacuum. The resulting polymer was dissolved in tetrahydrofuran and dialyzed against a 1:1 mixture of tetrahydrofuran and water for 3 days and against pure water for 2 days using dialysis membranes with a molecular weight cut-off of 6,000-8,000 g/mol (Spectrum Laboratories Inc.). Dialysis solutions were renewed every day. Finally, the remaining aqueous solution was freeze dried, producing a viscous pink polymer.

### S2.2. RHP sequence simulation and analysis

### S2.2.1. Simulations of RHP sequences

An in-house program was used to perform Monte-Carlo style simulations of the copolymerizations. We define the total number of monomers, their relative concentration ratios, and the monomer to RAFT ratio. Then, borrowing from the Mayo-Lewis copolymerization equation (Eq. S1, S2) (22), the reactivity ratios for each monomer with itself, along with each other monomer are inputted. Finally, a target percent conversion is chosen that determines the degree of polymerization of the polymers simulated.

Mayo-Lewis Equation: 
$$\frac{d[M_1]}{d[M_2]} = \frac{[M_1](r_1[M_1]+[M_2])}{[M_2]([M_1]+r_2[M_2])} \quad (\text{S1})$$

where  $[M_x]$  is the concentration of monomer  $x$ , and  $r_x = \frac{\kappa_{xy}}{\kappa_{xx}}$

Instantaneous Form: 
$$F_1 = 1 - F_2 = \frac{r_1 f_1^2 + f_1 f_2}{r_1 f_1^2 + 2f_1 f_2 + r_2 f_2^2} \quad (\text{S2})$$

where  $f_x$  is the mole fraction of monomer  $x$  in the feed, and  $F_x$  is the mole fraction of monomer  $x$  in the copolymer

The simulation is broken up into three steps: (i) Beginning, (ii) Growth, and (iii) Ending. A brief discussion of each step is outline below.

#### i. Beginning

- The first monomer on a given chain is determined based on the relative concentration ratio of that monomer, e.g. if MMA is 50% of the total monomer batch, there is a 50% probability of MMA being the first monomer. Each chain is then initiated by an implied idealized initiator with no preference toward any particular monomer, by randomly choosing one of the monomers taking into account the aforementioned weighted probabilities. This is done for the total number of chains simulated as defined by the monomer to RAFT ratio.

#### ii. Growth

- Each chain is then grown based on the probability for a given monomer to add to the one that precedes it in the simulated chain. First, weighting factors for each chain are established according to the equation  $w = r \times [M]$  where  $w$  is the weighting factor,  $r$  is the reactivity ratio for a given monomer pair, and  $[M]$  is the current concentration of the monomer being added. These weighting factors are normalized to produce probabilities for each type of addition event to occur. The chain is then grown randomly by one monomer based on the probabilities calculated previously, with the added monomer removed from the total monomer pool. This is completed for all chains, at which point the process loops back to the first propagation step.

#### iii. Ending

The growth process proceeds until the targeted percent conversion is achieved.

The 4-monomer simulation was performed assuming the reactivity ratios for the methacrylic methylene esters (OEGMA, EHMA, SPMA) to be equal and a reduced value for MMA, as determined by Manders et al.(21) (**Table S1**).

The validity of this assumption on reactivity was supported by the conversion of the individual monomers, as observed by  $^{13}\text{C}$ -NMR, which shows equal consumption of the methacrylic methylene esters (OEGMA, EHMA, SPMA), and a lower conversion of MMA at all global conversions as shown in **Fig. S4**. **Fig. S5** shows the calculated sequences of 25 heteropolymer chains. **Fig. S6** shows the calculated block size distribution using 2,000 simulated heteropolymers chains. Fig. S7 shows the monomer distribution pattern based on binary hydrophobicity.

### **S2.3. Characterization of RHP**

$^1\text{H}$  NMR and  $^1\text{H}$ - $^{13}\text{C}$  heteronuclear single quantum coherence (HSQC) 2D NMR spectra were carried out at 363 K with a Bruker Avance III 400 spectrometer (400 MHz for  $^1\text{H}$  and 100 MHz for  $^{13}\text{C}$ ) using a 5 mm BBFO+ probe. Quantitative  $^{13}\text{C}$  NMR distortionless enhancement by polarization transfer (DEPT135) spectra were carried out at 363 K and 100 MHz with a Bruker Avance II 400 spectrometer using a 10 mm selective  $^{13}\text{C}$  SEX probe. Polymer samples were examined as ca. 15% (w/v) solutions in *N,N*-dimethylformamide- $d_7$  (DMF- $d_7$ ). Chemical shift values ( $\delta$ ) are given in ppm in reference to residual hydrogenated solvent. Number average ( $M_n$ ) and weight average ( $M_w$ ) molar masses and dispersity ( $D = M_w/M_n$ ) of RHP were obtained from size exclusion chromatography (SEC) using a Waters 717 Plus autosampler, a 515 HPLC pump, a 2410 differential refractometer, a 2487 UV-vis detector, a MiniDawn multiangle laser light scattering (MALLS) detector (measurement angles are 44.7°, 90.0°, and 135.4°) from Wyatt Technology Inc., a ViscoStar viscosity detector from Wyatt, and five Styragel HR columns connected in the following order: 500, 103, 104, 105, and 100 Å. Tetrahydrofuran (THF) was used as the eluent at a flow rate of 1.0 mL/min at room temperature. The results were processed using the Astra 5.4 software from Wyatt Technology Inc.

$^1\text{H}$  NMR spectra of RHP were used to verify that neither organic solvents nor monomer residuals were present in the final samples. The attribution of NMR signals was performed by comparing the  $^1\text{H}$  NMR (**Fig. S8.A**), quantitative  $^{13}\text{C}$  NMR and DEPT  $^{13}\text{C}$  NMR (**Fig. S8.B**), and  $^1\text{H}$ - $^{13}\text{C}$  HSQC 2D NMR (**Fig. S9**) spectra with those of monomers **1**, **2**, **3**, **4** and 2-monomer random copolymers model analogues obtained by the copolymerization of 80 mol% of **1** and 20 mol% of either **2**, **3** or **4**.

Calculating experimental monomer ratios was performed by comparing the quantitative  $^{13}\text{C}$  NMR spectra integrations of **1** at 51.86 ppm ( $I_1$ ), **2** at 58.43 ppm ( $I_2$ ), **3** at 68.74 ppm ( $I_3$ ) and **4** at 48.77 ppm ( $I_4$ ). Molar ratios of monomers **1** ( $i$ ), **2** ( $j$ ), **3** ( $k$ ) and **4** ( $l$ ) were calculated using equations S3-S6.

$$i = I_1 / (I_1 + I_2 + I_3 + I_4) \quad (\text{S3})$$

$$j = I_2 / (I_1 + I_2 + I_3 + I_4) \quad (\text{S4})$$

$$k = I_3 / (I_1 + I_2 + I_3 + I_4) \quad (\text{S5})$$

$$l = I_4 / (I_1 + I_2 + I_3 + I_4) \quad (\text{S6})$$

RHP composition and SEC data (**Fig. S10**) can be found in **Table S2**.

### **S2.4. Analysis of compositional drift by varying conversion**

A set of 12 individual batches of RHPs were made and brought to various conversions to examine the possibility of compositional drift. All reagents were purchased from Sigma Aldrich and used without further purification unless specified otherwise. Azobisisobutyronitrile (AIBN) was recrystallized in ethanol prior to use. To remove inhibitors prior to polymerization, methyl methacrylate (**1**) and 2-ethylhexyl methacrylate (**3**) were cryodistilled and ethylene glycol methyl ether methacrylate (**2**) was passed over a column of basic alumina. 3-Sulfopropyl methacrylate potassium salt (**4**) was used without further purification. Ethyl-2(phenylcarbanothioylthio)-2-phenylacetate (98%) was purchased from Aldrich and used as received. Three different ratios of monomers to RAFT agent were tested: 572:1, 286:1, 191:1, with four duplicates of each. Reaction times were varied in order to target approximate conversions.

1. 25% Conversion Target (A1, A2, A3, A4) 4 clean 20 mL glass ampules were each charged with **1** (0.501 g, 5.00 mmol), **2** (1.25 g, 2.50 mmol), and **3** (0.397 g, 2.00 mmol) (all added as 2.14 mL of a 5.00:2.50:2.00M **1: 2: 3** solution), **4** (123 mg, 0.500 mmol, added as 0.760 mL of a 660 mM suspension in DMF), ethyl-2(phenylcarbanothioylthio)-2-phenylacetate (5.5 mg, 0.018 mmol) and AIBN (0.8 mg, 0.005 mmol) (added as 0.38 mL of a 46 mM RAFT:13 mM AIBN solution in DMF), trioxane (55 mg, 0.61 mmol, added as 0.090 mL of a 6.66 M solution in DMF), and 1.63 mL of DMF to a final reaction volume of 5.0 mL.

2. 50% Conversion Target (B1, B2, B3, B4) 4 clean 20 mL glass ampules were each charged with **1** (0.501 g, 5.00 mmol), **2** (0.125 g, 2.50 mmol), and **3** (0.397 g, 2.00 mmol) (all added as 2.14 mL of a 5.00:2.50:2.00M **1: 2: 3** solution), **4** (123 mg, 0.500 mmol, added as 0.760 mL of a 660 mM suspension in DMF), ethyl-2(phenylcarbanothioylthio)-2-phenylacetate (11 mg, 0.035 mmol) and AIBN (1.6 mg, 0.010 mmol) (added as 0.76 mL of a 46 mM RAFT:13 mM AIBN solution in DMF), trioxane (55 mg, 0.61 mmol, added as 0.090 mL of a 6.66 M solution in DMF), and 1.25 mL of DMF to a final reaction volume of 5.0 mL.

3. 75% Conversion Target (C1, C2, C3, C4) 4 clean 20 mL glass ampules were each charged with **1** (0.501 g, 5.00 mmol), **2** (0.125 g, 2.50 mmol), and **3** (397 mg, 2.00 mmol) (all added as 2.14 mL of a 5.00:2.50:2.00M **1: 2: 3** solution), **4** (123 mg, 0.500 mmol, added as 0.760 mL of a 660 mM suspension in DMF), ethyl-2(phenylcarbanothioylthio)-2-phenylacetate (16.6 mg, 0.052 mmol) and AIBN (2.5 mg, 0.015 mmol) (added as 1.14 mL of a 46 mM RAFT:13 mM AIBN solution in DMF), trioxane (55 mg, 0.61 mmol, added as 0.09 mL of a 6.66 M solution in DMF), and 0.87 mL of DMF to a final reaction volume of 5.0 mL.

Each reaction mixture was degassed by 4 freeze-pump-thaw cycles and the ampules were flame sealed at 30 mtorr. The ampules were held in an 80 °C oven for the total number of hours as listed: (A1: 18.25h A2; 18.25h A3: 2h A4: 5.8h B1: 18.25h B2: 20.3h B3:23.3h B4: 5.8h C1: 5.8h C2: 14.5h C3: 14.5h C4: 14.5h). Each viscous mixture was cooled in liquid nitrogen and then cracked open. The polymer was precipitated by dropwise addition of the diluted mixture into 250 mL of stirring pentane. The pink-purple precipitate was then dissolved in 10 mL of water and transferred to a 3500 MWCO centrifuge filter. Each sample was washed 5 times via spinning the solution down to a total volume of 2 mL and bringing it back up to 10 mL via addition of fresh water. The resulting solution was then transferred to a vial and dried under vacuum overnight.

Individual monomer conversion was calculated from <sup>13</sup>C NMR using the polymerization solvent (DMF carbonyl at 162.37 ppm) as the internal standard. Monomers MMA (136.25 ppm) and OEGMA (136.33 ppm) could be resolved individually. Monomers 2-EHMA and SPMA

could not be resolved from each other (136.55 ppm) and are grouped in the calculations of the conversion (**Fig. S4**).

Global monomer conversion was calculated from the individual monomer conversions together with the total monomer composition. Plotting the individual conversions against these total conversions for all samples reveals a slower consumption of MMA in comparison to the other monomers, as expected with the lower reactivity reported for MMA in copolymerizations with methylene methacrylate esters, including OEGMA. This suggests a slight drift from the starting monomer composition of 50% to 54-56% MMA, remaining at that level to at least 70% conversion. This indicates a minimal drift in copolymer composition and is further corroborated by the near-perfect overlapping in the superimposition of the <sup>1</sup>H NMR of the purified polymers at all conversions (**Fig. S11**). It should also be noted that the signal to noise increases with the conversions yielding less exact measurements at high conversions.

### **Section S3. All-Atom Molecular Dynamics Simulations**

#### **S3.1. Simulations methodology**

The classical molecular dynamics (MD) simulations were performed at the all-atom resolution using the package GROMACS (version 5.0) (33). The CHARMM 36 force field (34, 35) and the compatible CHARMM General Force Field (CGenFF) (36, 37) were used for all the molecules investigated, which has already been implemented in GROMACS (34). The CHARMM 36 force field has been well established in investigating macromolecule (both natural and synthetic) systems (38).

The structure of the protein HRP was downloaded from the RSCB Protein Data Bank with the protein ID 1H55. The degree of polymerization (*DP*) of 80 is employed for the RHP, which is close to the experimental molar weight of 20 kDa (*DP* is around 88). See **Fig. S12** for the structure of the RHP. In building a single RHP chain, random seeds were varied to match the experimental composition ratio of MMA : 2-EHMA : OEGMA : 3-SPMA = 10 : 4 : 5 : 1. Therefore, each polymer chain contained 40 MMA, 16 2-EHMA, 20 OEGMA and 4 3-SPMA monomers in a random order. In total, 12 different RHP chains were built using different random seeds so that the ratio of the concentration of polymer to that of protein matched the experimental value. To investigate the influence of solvent on the encapsulation of proteins by RHPs two different solvent conditions, aqueous solution and organic (toluene) solution were studied. The compositions of the components are provided in **Table S3**.

Due to the nature of long polymer chains, the encapsulation kinetics is beyond the capability of atomistic MD simulations. Therefore, some approaches were employed to speed up the polymer-protein aggregation process. In specific, the simulation was first performed in vacuum condition, followed by the simulations at an elevated temperature of 400 K before the production simulation at room temperature (298 K).

The initial structures of the polymer-protein complex were built using the package PACKMOL (39). The protein molecule was initially put in the center of the box with the edge length of 15 nm in all dimensions. The 12 polymer chains were subsequently arranged surrounding the protein with one out of the four -SO<sub>3</sub><sup>-</sup> groups distributed within the distance of 4 nm from the center of mass of the protein molecule. Note the size of the protein molecule is around 6 × 4 × 4 nm<sup>3</sup>. See **Fig. S13.V<sub>b</sub>** for the initial structure under the vacuum condition. After a short energy minimization using the steepest descent algorithm, the polymer-protein complex

was simulated under vacuum conditions for a duration of 1 ns. At the end of the simulation, all the polymer chains were aggregated surrounding the protein molecule. See **Fig. S13.V<sub>e</sub>** for the final snapshot. In the simulation under the vacuum condition, the NTV ensemble (constant number of particles, temperature and volume) was employed via the V-rescale thermostat (the reference temperature 400 K and the characteristic time of 0.5 ps). The backbone atoms of the protein molecule were constrained at their initial coordinate using a force constant of 1000 KJ/mol/nm<sup>2</sup> to maintain the structure of the protein molecule.

Based on the final frame of the simulation under vacuum condition, the polymer-protein complex was subsequently embedded in organic toluene solution or water solution. See **Table S3** for the number of solvent molecules added, and **Fig. S13.T<sub>b</sub>** and **Fig. S13.W<sub>b</sub>** for the obtained structures. They were used as the initial structures for the following simulations under the organic or the aqueous solutions. In the following simulations, the periodic boundary conditions were imposed in all the three dimensions; neighbor searching was performed for a distance up to 1.2 nm, and was updated every 10 time steps; the short-range van der Waals interactions using the Lennard-Jones (LJ) 12-6 potential were truncated at 1.2 nm with the long-range dispersion corrections applied for both the energy and the pressure; the short-range Coulomb interactions was also truncated at 1.2 nm with the long-range interactions calculated using the smooth Particle Mesh Ewald algorithm (40, 41). Moreover, to speed up the simulations, the simulation time step of 2 fs was employed by constraining the covalent bonds involving hydrogen atoms using the LINCS algorithm (42, 43). To speed up the process for the optimal polymer-protein aggregate structure, the simulations were first simulated at a temperature of 400 K, which served as a simple means to enhance the simulation sampling of protein-polymer aggregation. The NTP ensemble (constant number of particle, temperature and pressure) was applied with the temperature coupled via V-rescale algorithm and the isotropic Berendsen barostat (the reference pressure 1 bar). The simulation duration of 40 ns was performed, during which the protein backbone atoms were again constrained to maintain the structure.

The production simulations were subsequently performed at room temperature of 298 K. The position restrain was switched off for the protein molecule, allowing it to fully relax for the optimal structure. The NTP ensemble was imposed with the pressure coupled via the Parrinello-Rahman algorithm (reference pressure 1 bar, characteristic time 4 ps, compressibility of  $4.5 \times 10^{-5}$  bar<sup>-1</sup>). The temperatures of protein, polymer and solvent (toluene, or water) were separately coupled via the Nose-Hoover algorithm (reference temperature 298 K, characteristic time 0.5 ps). The frames were saved at a frequency of 50 ps. The toluene system was simulated for a duration of 600 ns with the last 400 ns employed for the data analysis. The water system was simulated 200 ns. The snapshots of the last frames were plotted in **Fig. S13.T<sub>e</sub>** and **Fig. S13.W<sub>e</sub>**, respectively.

To justify the convergence of the all-atom MD simulations, we calculated the radius of gyration ( $R_g$ , **Fig. S14.A**) and the root mean square deviation (RMSD, **Fig. S14.B**) of the backbone atoms of both protein HRP and the 12 polymer chains in the toluene solution simulations at 400 K and 298 K. Similar calculations were performed for the water solution simulations and presented in **Fig. S14(C, D)**. Furthermore, we calculated the system potential energies and densities in the two systems as a function of the simulation time (**Fig. S15**). The obtained results in **Fig. S14** and **S15** roughly support the convergence of protein-polymer aggregation behavior in the toluene and water solution simulations. Note that longer simulations at the microsecond or beyond, or enhanced sampling approaches (e.g., metadynamics (44, 45),

replica-exchange (46)), could provide further evidences in terms of the convergence of the all-atom simulations. Nevertheless, these methods are highly computationally expensive for complex, large-length-scale simulations, like the protein-polymers aggregates investigated here.

### **S3.2. Control all-atom simulation justifying the CHARMM36 force field**

In addition to the three systems aforementioned, a control simulation was performed. In the control system, one protein molecule was solved in the water solution in addition to the ligand molecules of HEME-Fe, acetate ion,  $\text{Ca}^{2+}$  ions and 403 water molecules. See also **Table S3**. All the simulation parameters were the same as those in the production simulations at 298 K presented above. The convergences of the calculated secondary structures and the RMSD of the protein backbone atoms, which are plotted in **Fig. S16**, justify the employment of the CHARMM36 force field.

### **S3.3. Protein encapsulation**

The protein-polymer complexes are forming core-shell structures, where protein 1H55 is encapsulated by the polymers (**Fig. S17**).

### **S3.4. Secondary structures of protein HRP**

The secondary structures of the protein 1H55 residues (**Table S4**) are calculated based on the STRIDE method under the VMD package (47).

### **S3.5. Correlation between protein surface residues and their nearest polymer neighbors in toluene**

To quantify the correlations between the protein surface and the surrounding polymer monomers, the protein surface residues, as well as the surrounding polymer monomers, are classified to be hydrophilic and hydrophobic. For the protein, the following residues are considered as hydrophobic: ALA, GLY, ILE, LEU, MET, PHE, PRO, TRP, TYR and VAL, while all other protein residues are considered hydrophilic. For the polymer, the MMA and 2-EHMA monomers are considered hydrophobic, and OEGMA and 3-SPMA monomers are hydrophilic (**Fig. S12**).

- 1) First, the protein backbone atoms on the surface are labeled by assuming that the protein is in the roughly spherical shape. To that end, all the protein backbone atoms are searched. An imaginary spherical shell of the radius of 5 nm (larger than the protein radius) is implicitly included based on the center of mass (COM) of the protein backbone atoms ( $x_c$ ,  $y_c$ ,  $z_c$ ). For each polymer backbone atom located at ( $x$ ,  $y$ ,  $z$ ), the nearest position ( $x_s$ ,  $y_s$ ,  $z_s$ ) on the spherical shell is obtained based on the vector from the protein COM ( $x_c$ ,  $y_c$ ,  $z_c$ ) to the protein backbone atom ( $x$ ,  $y$ ,  $z$ ). Then all the protein backbone atoms are searched so as to find the nearest one to the position on the shell ( $x_s$ ,  $y_s$ ,  $z_s$ ), which is consequently labeled as a protein surface atom. Based on all the labeled protein surface atoms, the hydrophilic/hydrophobic features of the protein surface are counted quantitatively. See **Fig. S18** for the distribution of the hydrophilic (in red) and hydrophobic (in green) protein surface.
- 2) Based on the labeled protein surface atoms, those polymer tail atoms are marked which are the nearest-neighbors of the protein surface atoms labeled in the previous step. In these calculations, the last hydrocarbon atoms on the tails of MMA and 2-EHMA monomers, the last oxygen atoms on OEGMA and the sulfur atom on 3-SPMA monomers are employed

(**Fig. S12**). With MMA and 2-EHMA defined as hydrophobic, OEGMA and 3-SPMA defined as hydrophilic, the hydrophilic and hydrophobic features of the polymer neighbors to the labeled protein surface atoms are consequently defined. See **Fig. S18** for the distribution of the hydrophilic (in magenta) and hydrophobic (in blue) polymer neighbors.

- 3) Based on the hydrophilic and hydrophobic nature of the protein surface atoms and of their nearest-neighbor polymer tail atoms, the numerical probabilities of the correlations between the hydrophilic and hydrophobic protein surface and the hydrophilic and hydrophobic polymer neighbors are quantitatively calculated, and presented in **Fig. S19.A**.
- 4) Similar to Step 3, the intermolecular (Coulomb and LJ 12-6) interaction energies between the protein surface residues and their polymer neighbor residues are thus calculated. The obtained results are provided in **Fig. S19.B**.

#### **Section S4. Coarse-Grained Model and Simulations**

We developed a coarse-grained model and simulation based on the all-atom model and results obtained from Section S4. Specifically, we took the final HRP configuration equilibrated in toluene (**Fig. S13.Te**) and created a shape-based coarse-grained model using the CGBuilder plugin in the VMD software package (48, 49). The atoms in the hydrophobic residues are coarse-grained into 98 spherical beads and those in the hydrophilic residues into 107 spherical beads.

The random copolymers are modeled as linear chains consisting of  $P = 20$  beads of 4 types A, B, C and D, corresponding to monomers MMA, 2-EHMA, OEGMA and 3-SPMA, respectively (**Fig. S12**). The bonds between consecutive beads in a chain are connected via finitely extensive nonlinear elastic (FENE) springs. Despite its simplicity, the linear chain model proves to be sufficient for qualitatively capturing the adsorbing behavior of the polymers on the protein surface observed in our all-atom MD simulations and experiment.

In developing the present CG model, we choose to focus on the competition between the short-ranged attractions between polymer beads and between polymer beads, leaving electrostatic interactions for future studies. This is reasonable for the present study because the fraction of the charged residues on the HRP surface under experimental conditions is presumably small. As such, the interaction between the adsorbing polymer beads (i.e. of types A and B) and polymer attractive sites is modeled by the 12-6 Lennard-Jones (LJ) potential, of which the energy well depth,  $\epsilon_{\text{Hh}}$ , characterizes the adsorption strength. The solvent selectivity is modeled by the effective attraction between the polymer beads of types A and B themselves via the LJ potential with the energy well depth  $\epsilon_{\text{hh}}$ . The LJ potentials are truncated and shifted to zero at the cutoff distance of  $3.0\sigma$ . The effective interactions between other bead types are modeled by the purely repulsive Weeks-Chandler-Andersen (WCA) potential.

The CG simulations are performed at constant temperature and volume using the Langevin thermostat. The solvent molecules and counterions are treated implicitly and their effects are modeled by the random and drag forces applied to individual beads by the Langevin thermostat. The equation of motion of individual CG beads is governed by:

$$ma = \mathbf{F}_C + \mathbf{F}_D + \mathbf{F}_R$$

where  $m$  and  $\mathbf{a}$  are the bead mass and acceleration, respectively;  $\mathbf{F}_C$  is the conservative force,  $\mathbf{F}_D = -\gamma\mathbf{v}$  is the drag force, and  $\mathbf{F}_R$  is the random force, whose magnitude is proportional to

$(k_B T \gamma / \Delta t)^{1/2}$  according to the fluctuation-dissipation theorem. The simulation time step is  $\Delta t = 0.005\tau$ , where  $\tau$  is the dimensionless time unit  $\tau = \sigma(m/\varepsilon)^{1/2}$ , where  $m = 1$  is the mass of the CG beads and  $\varepsilon = 1$  is the energy well depth of the WCA potential. The dimensionless set temperature is  $T^* = k_B T / \varepsilon = 1.0$ . All the CG simulations were performed using LAMMPS version 20 Sep 2016 (50).

The protein is located at the center of the cubic simulation box whose dimension is  $L = 65\sigma$  with periodic boundary conditions. The protein beads are constrained to their equilibrated configurations via harmonic springs to ensure that the global structure of the protein is maintained while taking into account the effects of thermal fluctuations due to implicit solvent and counterion bombardments. We vary the number of polymer chains  $N_C = 12, 50$  and  $100$ , to cover the molar ratios used in experiment and that in all-atom MD simulation ( $N_C = 12$ ). The total number of CG beads varies from  $N = 445$  to  $2205$ . The random order of the monomers within the chains is generated by two methods: 1) the bead types are shuffled within individual chains and 2) the bead types are shuffled within the total number of polymer beads. The first method is used when the number of chains  $N_C$  is sufficiently large, e.g.  $N_C = 50$  and  $100$ . The second method is used when the number of chains is small, i.e.  $N_C = 12$ , so as to improve the randomness in the monomer order. The compositions of beads of types B and D are always fixed at  $0.2$  and  $0.05$ , respectively, consistent with those in full-atom MD simulations. When the composition of beads of type A,  $\phi_A$ , is varied, the composition of beads of type C is adjusted accordingly to ensure that  $\phi_A + \phi_B + \phi_C + \phi_D = 1$ .

To accelerate the sampling of the adsorption process by MD simulations, we constrain all the polymer chains within a spherical volume of radius  $R_c$  using harmonic springs (i.e. using the command fix indent in LAMMPS). The value of  $R_c$  is chosen so that the number density of the polymer beads in the spherical confinement is fixed at  $0.1\sigma^{-3}$  for all the polymer-protein molar ratios studied. The choice of such a number density is motivated by our all-atom MD simulations and experimental conditions.

To ascertain that the simulation results are not biased by initial configurations, we vary the random seeds used to generate the random polymer chains and the initial velocity profiles of the polymer beads. To further improve the statistics of the simulation results, we perform annealing/equilibrating cycles for each data point (Fig. S20A). During the annealing period, the interaction between the polymer attractive sites and polymer adsorbing beads is switched to the purely repulsive WCA potential so that the system is annealed at the athermal condition before being equilibrated again. In addition to monitoring the system potential energy, we also measure the average number of adsorbing polymer sites in the vicinity of each attractive site (Fig. S20.B) to justify if the equilibration reaches steady state. At the end of the equilibration period, the final configuration is then collected for analyzing surface coverage, pair correlation functions and radial density profiles. By using annealing/equilibrating cycles, we attempt to avoid kinetically trapped configurations with strong attractions, as well as to obtain statistically uncorrelated final configurations.

We have also run additional Hamiltonian exchange simulations to assure the system is equilibrated at equilibrium. In the Hamiltonian exchange simulations, we used 16 replicas each assigned with an adsorption strength  $\varepsilon_{Hh}$  in the range of  $0.3$ - $1.8$   $k_B T$ . A trial swap in  $\varepsilon_{Hh}$  between adjacent replicas is attempted every  $10^4$  time steps. The swap is accepted or rejected using a Metropolis-like criterion (1). A replica-exchange simulation for a given value of  $\phi_A$  is performed

for 50 million time steps. The configurations obtained at each value of  $\varepsilon_{\text{Hh}}$  are then collected at the end of the run from the replicas for surface coverage analysis. The results from the Hamiltonian exchange runs are indeed consistent with those obtained from the heating/cooling procedure, as already reported.

We estimate surface coverage  $\Gamma$  from particle-based molecular simulation data as follows. First, the protein beads that interact with the polymer beads are identified. The maximum distance from the protein beads to the center of mass of the protein  $r_{\max}$  is then calculated. Next, we construct a spherical volume of radius  $(r_{\max} + r_c)$  centered at the protein center of mass, and divide the spherical surface into  $N_{\text{cells}}$  with equal areas so that the dimension of each cell is approximately  $\sigma$ . The polymer beads that are within the spherical volume are then binned into the cells based on their coordinates. The ratio between the number of the occupied cells and the total number of cells  $N_{\text{cells}}$  gives the approximate measure of surface coverage  $\Gamma$ .

**Fig. S21** shows the representative final configurations for different adsorption strengths,  $\varepsilon_{\text{Hh}}$ , when the attraction between the beads of type A and B is fixed at  $\varepsilon_{\text{hh}} = 0.8k_B T$  and the fraction of A is  $\phi_A = 0.5$ .

**Fig. S22.A** demonstrates that the pair correlation functions between adsorbed polymer beads for different fractions of the adsorbing components  $\phi_A$  and between the attractive sites of the protein. The increase in the peak height of the pair correlation function with  $\phi_A$  strongly correlates with the surface coverage, suggesting that the polymer adsorbed beads are more likely to match the distribution of protein attractive sites upon increasing the composition of the adsorbing components.

Our simulation results reveal that surface coverage increases with the polymer-ratio molar ratio (**Fig. S22.B**) for all the adsorption strength studied. This is because  $N_C$  is proportional to the number of adsorbing beads, which to a certain degree have similar effects to increasing  $\phi_A$ .

The structural properties of the polymer chains are investigated and shown in **Fig. S23**. The variation in the number of adsorbed polymer chains,  $n_{\text{ads}}$ , as functions of  $\phi_A$  and adsorption strength  $\varepsilon_{\text{Hh}}$  is shown in **Fig. S23.A**. Upon increasing  $\phi_A$ ,  $n_{\text{ads}}$  increases because there are more adsorbing polymer beads, as expected. Nevertheless, at higher values of  $\phi_A$ ,  $n_{\text{ads}}$  is reduced for strong attractions between polymer beads of types A and B. The strong correlation between  $n_{\text{ads}}$  and surface coverage  $\Gamma$  (**Fig. 2.H**) can be explained as follows. In the one hand,  $n_{\text{ads}}$  is proportional to the energetically favored polymer-protein contacts governed by the adsorption strength. Consequently,  $n_{\text{ads}}$ , and hence  $\Gamma$ , increases with  $\phi_A$  and the density and distribution of the protein attractive sites. On the other hand,  $n_{\text{ads}}$  characterizes with the entropic loss due to the surface confinement of the chains, compared to those that are not adsorbed. The entropic loss is reduced when  $\varepsilon_{\text{hh}}$  increases because the polymer beads of types A and B that are not non-adsorbed tend to aggregate. The optimal value of  $\Gamma$  and the maximum value of  $n_{\text{ads}}$  are then attributed to the competition between the tendency of the polymers to avoid entropic loss due to confinement to the surface and their tendency to maximize their energetically favored contacts.

**Fig. S23.B** shows the pair correlation functions of the adsorbing beads and polymer attractive sites for different adsorption strengths, which are consistent with the findings from our all-atom MD simulations in Section S3. **Fig. S23.C** further shows the radial density profiles of the polymer beads from the protein center of mass for different adsorption strengths. The peak of the density profiles shifts to higher values and gets closer to the protein surface as  $\varepsilon_{\text{Hh}}$  increases.

## **Section S5. RHP-Mediated Cell Free Synthesis and Protein-RHP Complexes**

### **S5.1. Chemicals**

Chemical reagents were purchased from Sigma-Aldrich Chemical Co. or Fisher Scientific International Inc. unless otherwise noted. Cell-free transcription/translation system (PURExpress®) was purchased from New England BioLabs Inc. Plasmid pWaldo-GFPe\_PepTso was a gift from So Iwata & Simon Newstead (Addgene plasmid # 58334). The GFP tag in the plasmid was removed to generate a plasmid pWaldo-PepTso for protein function assay by QB3 MarcoLab (University of California, Berkeley). Plasmid Aqpz-GFP was a gift from Prof. Daniel L. Minor, Jr. (University of California, San Francisco). DOPC, POPE and POPG phospholipids were purchased from Avanti® Polar Lipids Inc. Sephadex G-25 i(PD-10) desalting columns were purchased from GE Healthcare. Anti-GFP antibody (B2) was purchased from Santa Cruz Biotechnology. Goat anti-mouse alkaline phosphatase secondary antibody conjugate was purchased from Bio-Rad Laboratories.

**Cell-free protein synthesis.** Cell-free protein synthesis was performed according to PURExpress manual with modifications. Briefly, in a 25  $\mu$ l reaction, **RHP** was added to the ribosome solution in a ratio of 50/1 (mole/mole), and incubated on ice for 30 min. The polymer/ribosome samples were mixed with solution A and B containing all other required enzymes, RNAs, energy and nutrient molecules. 20 units RNase inhibitor and 200 ng plasmids for PepTso-GFP, PepTso, or AqpZ-GFP were added, and the mixtures were incubated at 37 °C for 3 hours to complete membrane protein synthesis. The resulting samples were stored at -20 °C.

**Kinetics of protein synthesis and Western blot analysis.** Kinetics of PepTso-GFP and AqpZ-GFP cell-free synthesis were monitored through measuring the fluorescence intensity of the GFP tag on a Tecan I-control infinite 200 plate reader. Cell-free protein synthesis mixtures were incubated in 384-well plate at 37 °C, and the GFP fluorescence (Ex/Em 480 nm/520 nm) was recorded over 3 hours. For the western blot, proteins in the cell-free protein synthesis samples were first separated by sodium dodecyl sulfate polyacrylamide gel electrophoresis (SDS-PAGE). Lane 1 corresponds to protein mass standards. (Spectra multicolor broad range protein ladder, (ThermoFisher)) The proteins were then electrophoretically transferred to a polyvinylidene difluoride (PVDF) membrane. Multicolor protein mass standards were visible on the membrane after successful transfer. The membrane was then blocked with 3 % BSA, incubated with 1:2000 mouse anti-GFP antibody, and washed. Following incubation with 1:4000 alkaline phosphatase conjugated goat anti-mouse secondary antibody, band colors were developed by using 5-bromo-4-chloro-3-indolyl phosphate (BCIP) and nitroblue tetrazolium (NBT).

**Reconstitution of membrane protein and transport assay.** The preparation of POPE/POPG (3/1) proteoliposomes containing PepTso was performed following the reported method with modifications (26). Briefly, to prepare liposomes, chloroform in lipid was evaporated, and resulting lipid film was further dried under vacuum overnight. Dried lipid film was solubilized at 10 mg/mL in an inner solution (5 mM HEPES pH 6.8, 1.2 mM NaCl, 120 mM KCl, 2 mM MgSO<sub>4</sub>, 1 mM pyranine) at 37 °C for 1 hour. 10  $\mu$ L of the cell-free PepTso synthesis mixtures was added to 500  $\mu$ L of the liposomes solutions, which was then subjected to 6 cycles of freeze/thaw using liquid nitrogen and 37 °C water bath. The resulting proteoliposomes were

extruded 10 times through 0.4  $\mu$ m membranes at 37 °C. Excess pyranine was removed through a sephadex G-25 desalting column using the inner solution without pyranine. Before proton and oligopeptide transport assay, samples containing the proteoliposomes were diluted 20 folds in the outer solution (5 mM HEPES pH 6.8, 1.2 mM KCl, 120 mM NaCl, 2 mM MgSO<sub>4</sub>) containing 0.2 mM dipeptide Ala-Ala. 1  $\mu$ M valinomycin was added to the outer solution to initiate the proton transport. Proton transport was measured by reading the ratio of pyranine fluorescence at 510 nm when excited at 460 and 415 nm (52) in a cuvette with a stirring magnetic flea, on a Perkin Elmer LS55 Fluorescence spectrometer.

Horseradish peroxidase type II (HRP, Sigma-Aldrich) and glucose oxidase type II (GOx, Sigma-Aldrich) were dissolved in Milli-Q water (Millipore) to a concentration of 10 mg/mL.  $\alpha$ -chymotrypsin ( $\alpha$ -CT, Sigma-Aldrich) was dissolved in 0.1 M Tris-HCl with 1 mM CaCl<sub>2</sub>, pH 7.75, to a concentration of 10 mg/mL. Upon arrival of green fluorescent protein (GFP, Millipore), GFP was centrifuge filtrated to remove salts and glycerol using Amicon Ultra-0.5 mL centrifugal filters (Millipore) with a cutoff molecular weight of 10 kDa. GFP was then dissolved in Milli-Q water to a concentration of approximately 1 mg/mL. Organophosphorus hydrolase (OPH) was expressed, based on established procedure. The enzyme was stored in a 50 mM Tris-HCl buffer, pH 9, at a concentration of 1.5 mg/mL. RHPs were dissolved in Milli-Q water at a concentration of 1 mg/mL. Sodium bis(2-ethylhexyl) sulfosuccinate (AOT, Aldrich) was dissolved in toluene at a concentration of 200 mM. Polymeric surfactant, polystyrene(22kDa)-block-poly(ethylene oxide) (21.5kDa) (PS-*b*-PEO), was purchased from Polymer Source and was dissolved in toluene at a concentration of 1 mg/mL.

To prepare reverse micelles, HRP solutions were injected into AOT solution at a [H<sub>2</sub>O]/[AOT] ratio of  $w_0=13$  or injected into PS-*b*-PEO at a volumetric ratio of 1:50 HRP solution to polymer solution. The suspensions were sonicated, mixed, and partially evaporated using a constant stream of N<sub>2</sub> gas until optically clear. Additional toluene was added to compensate for evaporated solvent in order to reobtain original concentrations. Protein/RHP complexes were obtained by combining the aqueous RHP solution and protein solution at a volume ratio of 50:1. RHP/protein mixtures were lyophilized overnight, resuspended with toluene to the original concentration, and subsequently sonicated. Other than GFP, all materials were used as received.

## **S5.2. Structural characterization**

**Transmission Electron Microscopy (TEM)** RHP-HRP complexes were drop casted on TEM carbon grids (Ted Pella) and dried for 5 minutes. Subsequently, several drops of water were used to wash off excess free polymer. Samples were stained using a 2 w/v% solution of phosphotungstic acid in water for 2 minutes. TEM images were taken on a JEOL 1200EX TEM at an accelerating voltage of 80 kV. In toluene HRP/RHP forms nanoparticles, ~50 nm in size.

**Fourier-Transform Infrared Spectroscopy (FT-IR)** To enhance the signal-to-noise ratio, solutions of HRP/RHP complex were concentrated to reach a HRP concentration of ~8-10 mg/mL. Solutions were deposited in a liquid cell composed of CaF<sub>2</sub> windows with a path length of 1 mm. Spectra were collected at 0, 2, 4, 8, and 24 hours after resuspension in toluene on a Thermo Scientific Nicolet 6700. Measurements were conducted at room temperature and the amide I band between 1700 cm<sup>-1</sup> and 1620 cm<sup>-1</sup> was monitored. Spectral analysis was performed using the built-in OMNIC Spectra Software.

**UV-Visible Spectroscopy (UV-Vis)** UV-visible spectroscopy was performed on a Hewlett-Packard 8453 Spectrophotometer. HRP/RHP complex solutions in toluene were sealed in a 1-cm path length quartz cuvette. Measurements were conducted at room temperature and the location of the hemin peak was monitored between 350 nm and 800 nm at 0, 2, 4, 8, and 24 hours after resuspension.

**Fluorescent Spectroscopy** Retention of GFP fluorescence was evaluated using a Perkin Elmer LS-55 fluorescence spectrometer. Immediately after dissolving GFP/RHP in toluene, solutions were sealed in a 1-cm path length quartz cuvette and excited at an excitation wavelength of 450 nm. Emission wavelength was monitored between 450 nm and 600 nm. Measurements were conducted at room temperature and taken at 0, 2, 4, 8, and 24 hours after resuspension in toluene.

### S5.3. Assay of protein activity in aqueous buffer

Solutions were left in ambient room temperature conditions for 0, 2, 4, 8, and 24 hours in toluene. After specified times, aliquots were taken and diluted in 100 mM KH<sub>2</sub>PO<sub>4</sub>/K<sub>2</sub>HPO<sub>4</sub> phosphate buffer, pH 6, to disperse the protein and RHP. After thorough mixing, the assay solution is applied. Activity was quantified using UV-visible spectroscopy by monitoring the conversion of the colorimetric assay.

**HRP activity** was assessed using a TMB Peroxidase EIA Substrate Kit (Bio-Rad). Baseline HRP activity was determined by preparing a stock protein solution in 100 mM KH<sub>2</sub>PO<sub>4</sub>/K<sub>2</sub>HPO<sub>4</sub> phosphate buffer at pH 6 and applying a prepared TMB assay solution, as outlined by the manufacturer. Solution was thoroughly mixed and UV-visible spectroscopy was performed on a Hewlett-Packard 8453 Spectrophotometer. A 1-cm path length cuvette was used and absorbance at 370 nm was monitored.

**GOx activity** was assessed using an assay containing glucose, phenol, 4-aminoantipyrine, and HRP. Baseline GOx activity was determined by preparing a stock protein solution in 100 mM KH<sub>2</sub>PO<sub>4</sub>/K<sub>2</sub>HPO<sub>4</sub> phosphate buffer at pH 6 and applying the colorimetric assay. A 1-cm path length cuvette was used and absorbance at 505 nm was monitored.

**OPH activity** was assessed using an assay containing methyl parathion. Baseline OPH activity was determined by preparing a stock protein solution in 50 mM TRIS-HCl buffer and applying the colorimetric assay. OPH activity was monitored using a Thermo Fisher Scientific NanoDrop 2000 at 405 nm.

### S5.4. Evaluation of RHP-assisted protein dispersion and stabilization for other proteins

**GFP** Similar experiments were performed using GFP and fluorescence was monitored as a function of incubation time in toluene (**Fig. S27**).

**GOx PEGylation** was performed following the procedure reported previously. (28) The feeding ratio of PEG was adjusted to increase PEGylation.

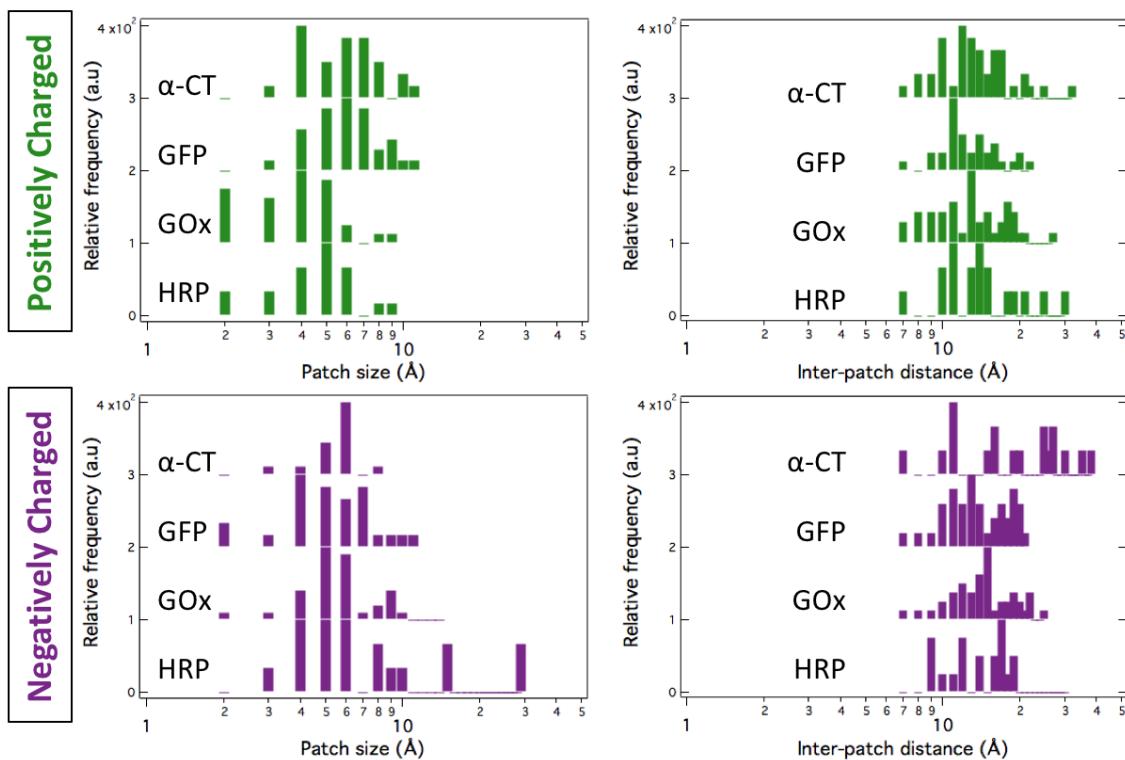
### S5.5. Protein-containing fiber mats

**Electrospinning** For PMMA based electrospun fiber mats, poly(methyl methacrylate) (PMMA, Sigma-Aldrich) ( $M_w=350,000$  g/mol) was first dissolved in chloroform to achieve a 7.5 wt% solution concentration. 200  $\mu$ L of toluene containing the protein/RHP complex was mixed

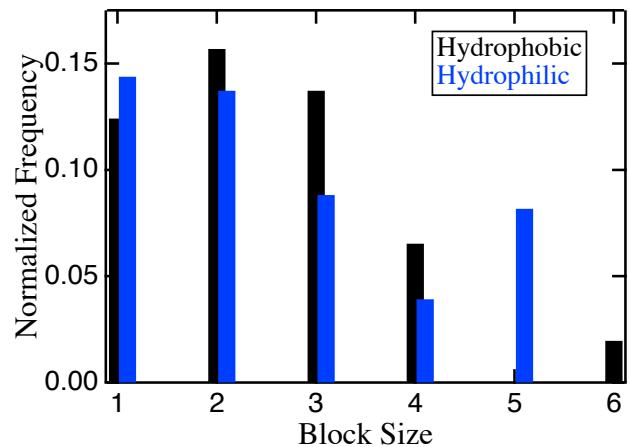
with 800  $\mu$ L of the PMMA/chloroform solution. For present studies, the protein concentration in the fiber mat is typically in the range of 0.3-0.5 wt%. The mixture was stirred for 5 minutes at 350 RPM immediately prior to electrospinning. The solution was then electrospun from a 1 mL syringe and a 20 gauge needle. An aluminum collector plate was placed 18 cm away from the needle tip in a horizontal setup. The solution was pumped at a rate of 0.45 mL/hr. A voltage of 8 kV was applied to the solution. For PEO-based fiber mats, OPH-RHP was lyophilized and resuspended directly into a 5.5 wt% PEO ( $M_w = 900,000$  g/mol) in MilliQ water solution. A flat aluminum collector plate was placed 20 cm from the needle tip. An accelerating voltage of 9 kV was applied to the solution, which was pumped at a flow rate of 0.35 mL/hr.

**Scanning Electron Microscopy (SEM)** A Hitachi TM-1000 scanning electron microscope was used to acquire images of the electrospun fibers. The samples were mounted on conductive carbon tape. A 15 kV accelerating voltage was used to acquire the SEM images.

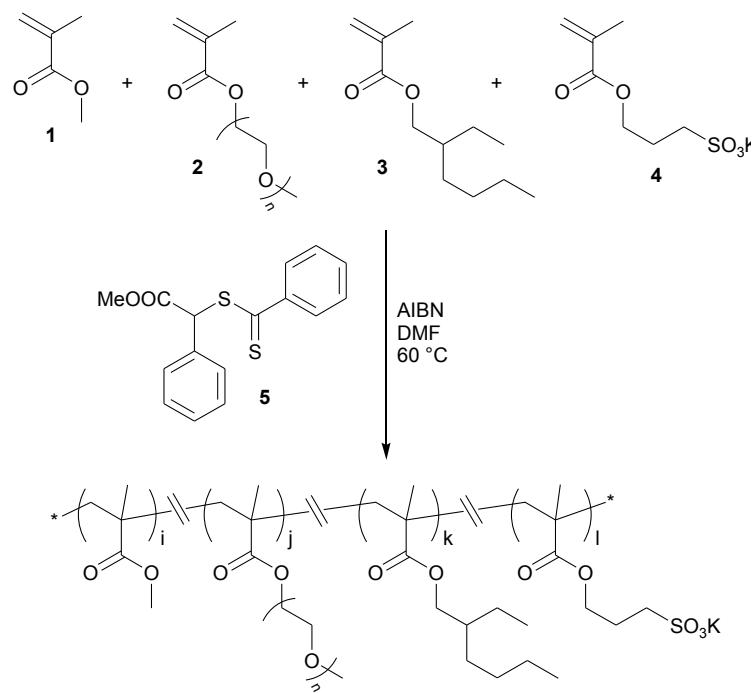
**Assay of OPH Activity in Electrospun Fibers** Protein/RHP-loaded fiber mats were tested in the same aqueous buffer used to measure activity of free proteins. The activity was monitored visually by the solution color change. A protein loading of approximately 1  $\mu$ g of OPH per 1 mg of polymer was achieved.



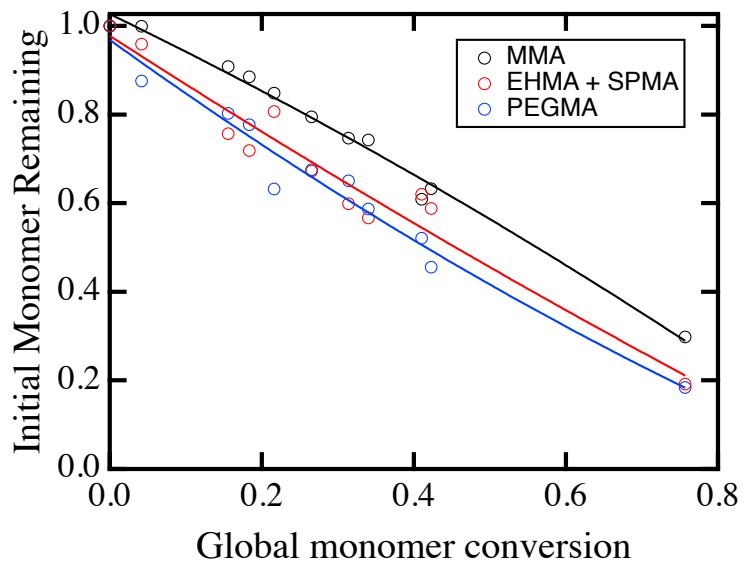
**Fig. S1. Histograms of the diameters and distances of (top) positively and (bottom) negatively charged patches.** The average patch diameter is less than 1 nm and the inter-patch distance is typically less than 2 nm. The characteristic length scales are used to guide the design of random heteropolymers.



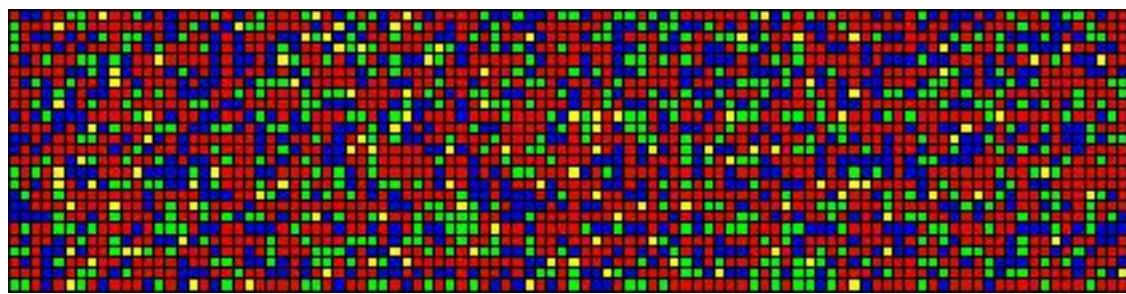
**Fig. S2. Protein sequence analysis of HRP (1H55).** The histogram of the block size along the HRP sequence is determined using binary hydrophobicity assignment. The average length of consecutive amino acids in HRP with the similar hydrophobicity ranges from 1-6.



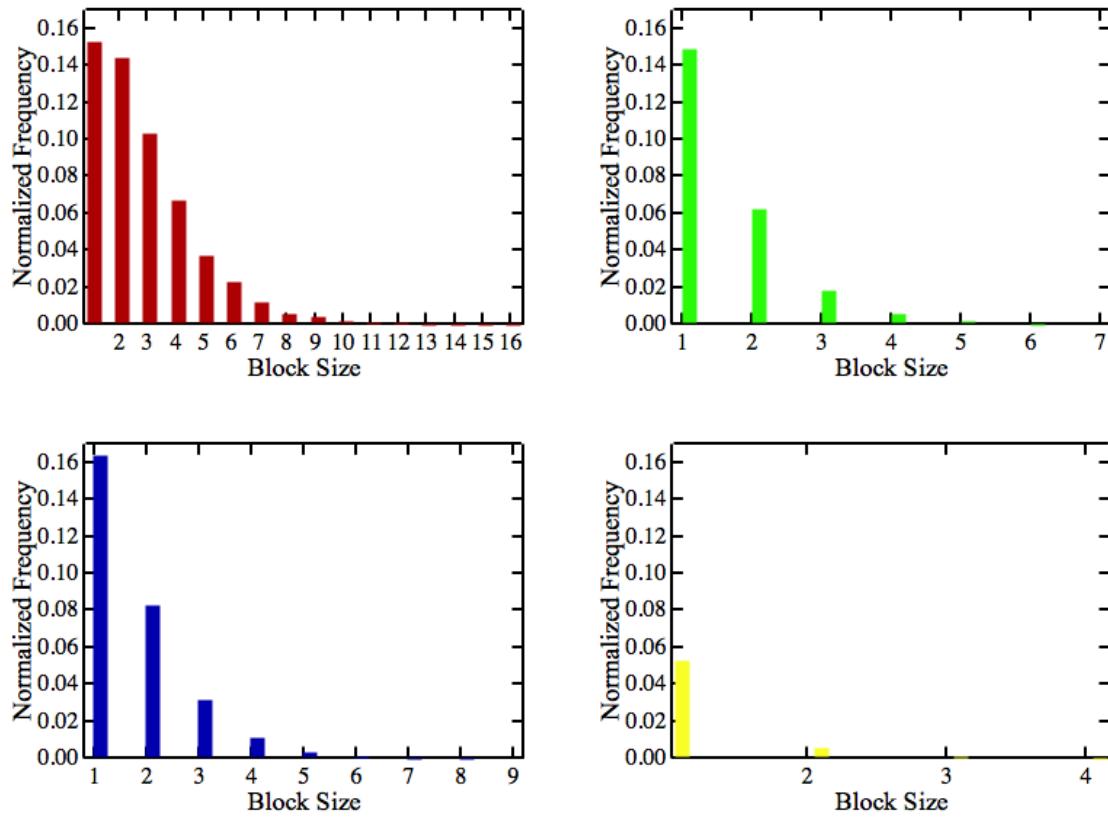
**Fig. S3. Synthesis of RHP by RAFT polymerization.**



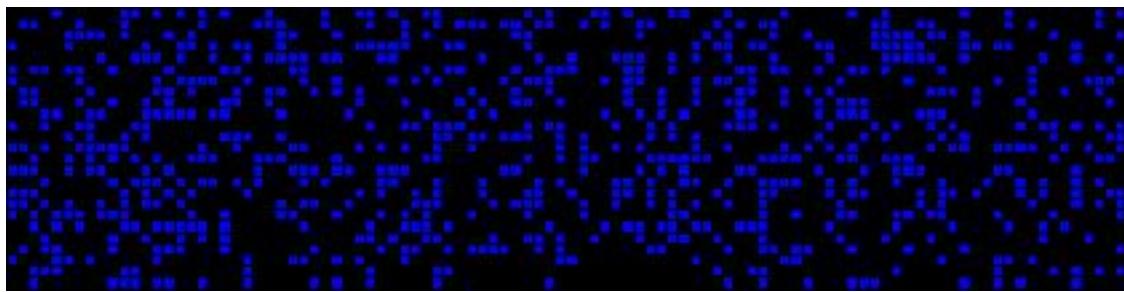
**Fig. S4. Individual monomer conversion as a function of global conversion.**



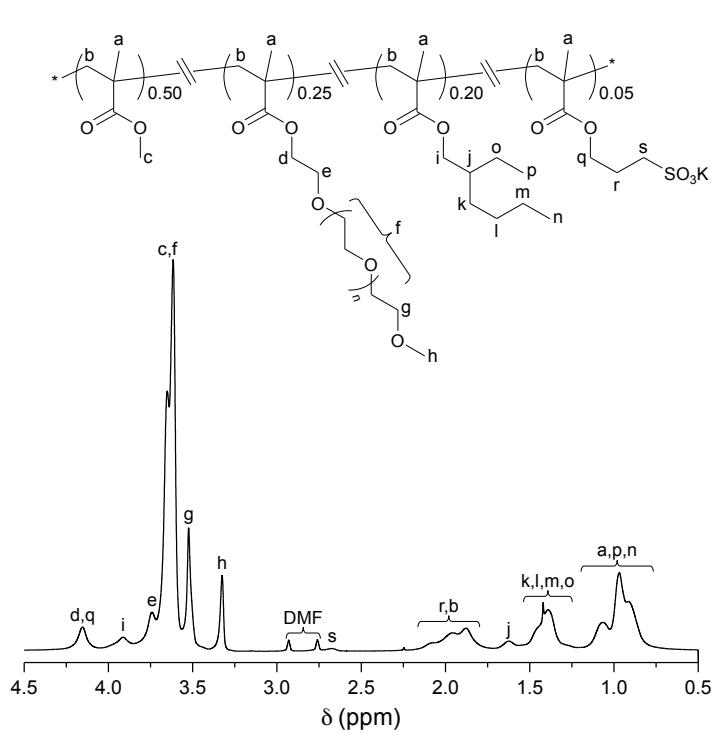
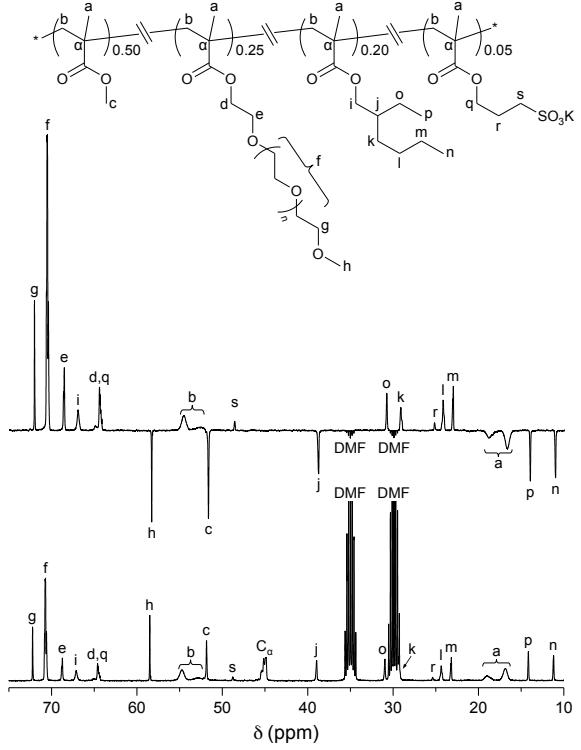
**Fig. S5. Graphical representation of 25 statistically random heteropolymer chains of DP 100.** Each row represents a single chain with 4 monomers, generated using Monte Carlo simulations based on the Mayo-Lewis equation. (Red: MMA, Green: OEGMA, Blue: EHMA, Yellow: SPMA)



**Fig. S6. Characteristic block sizes for statistically random 4-monomer heteropolymer.**  
Calculated from 2000 chains Monte Carlo simulation based on Mayo Lewis equation. (Red: MMA, Green: OEGMA, Blue: EHMA, Yellow: SPMA)



**Fig. S7. Graphical representation of 25 random heteropolymer chains of  $DP = 100$ .** Each row represents a single chain with 4 monomers, showing hydrophobic (MMA, EHMA) (black) and hydrophilic (OEGMA, SPMA) (blue) block sizes, generated using Monte Carlo simulations based on Mayo-Lewis equation.

**A****B**

**Fig. S8. NMR of RHP.** (A)  $^1\text{H}$  NMR ( $\text{DMF}-d_7$ , 400 MHz) of RHP. (B) (top) DEPT  $^{13}\text{C}$  NMR and (bottom) quantitative  $^{13}\text{C}$  NMR ( $\text{DMF}-d_7$ , 100 MHz) of RHP.

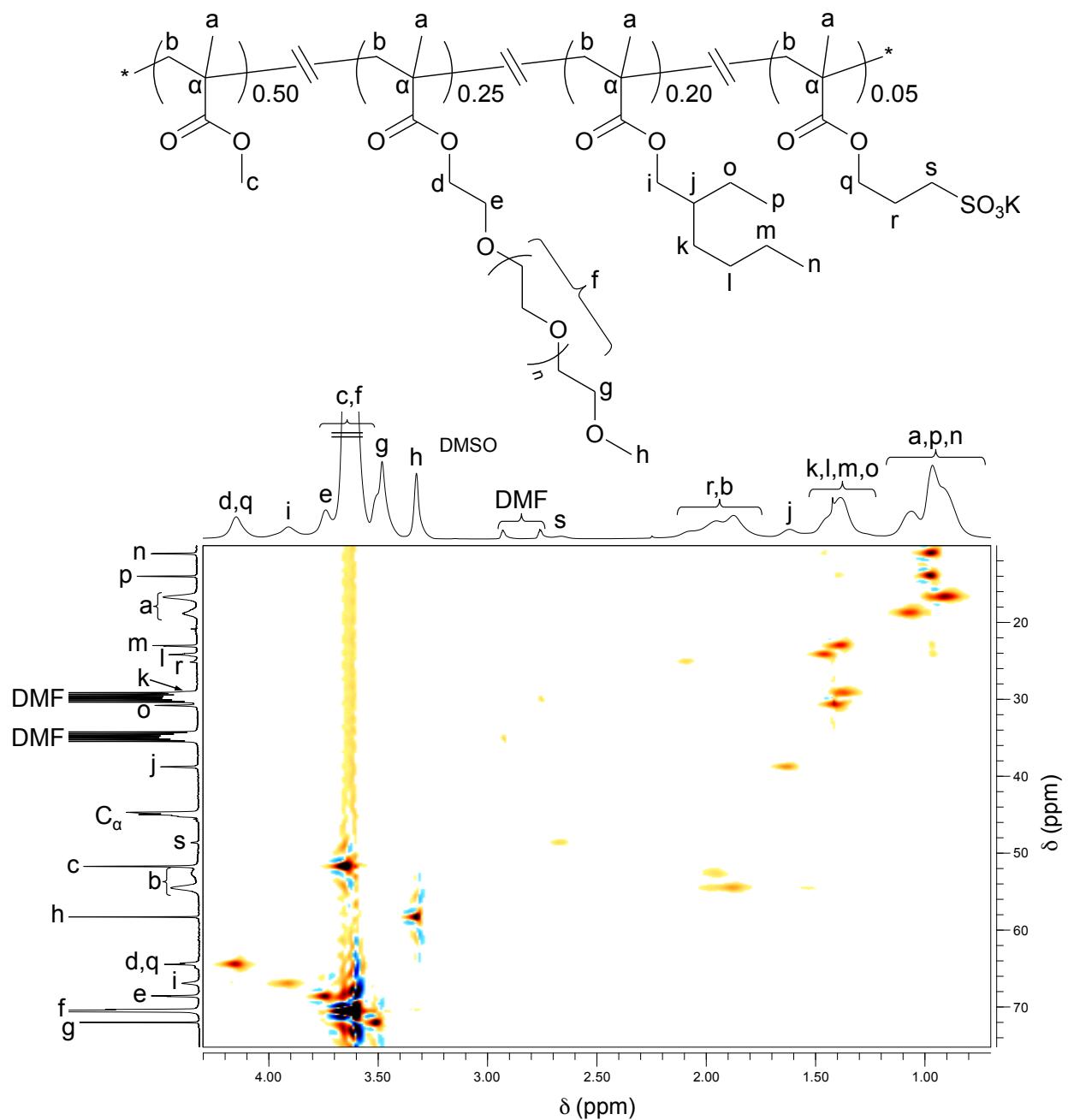
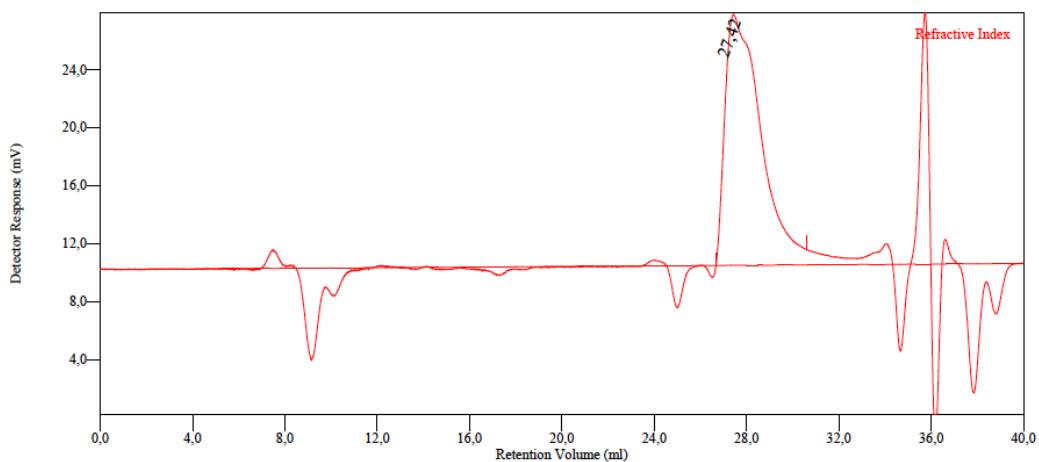
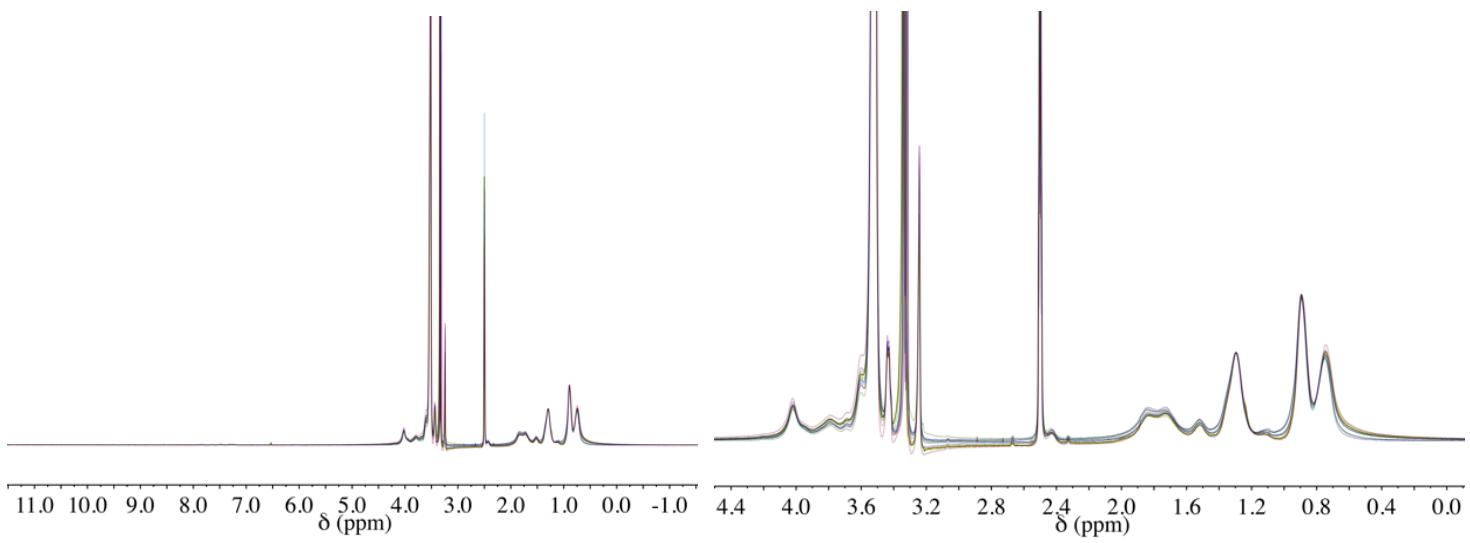


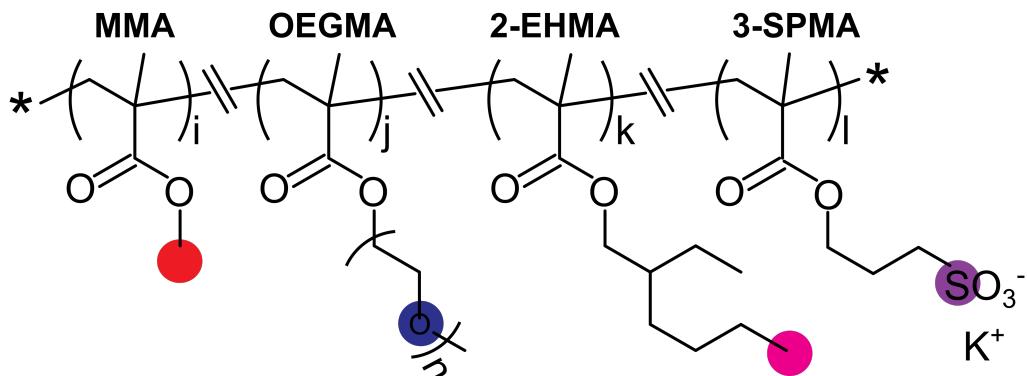
Fig. S9.  $^1\text{H}$ - $^{13}\text{C}$  HSQC 2D NMR (DMF- $d_7$ , 100 MHz) of RHP.



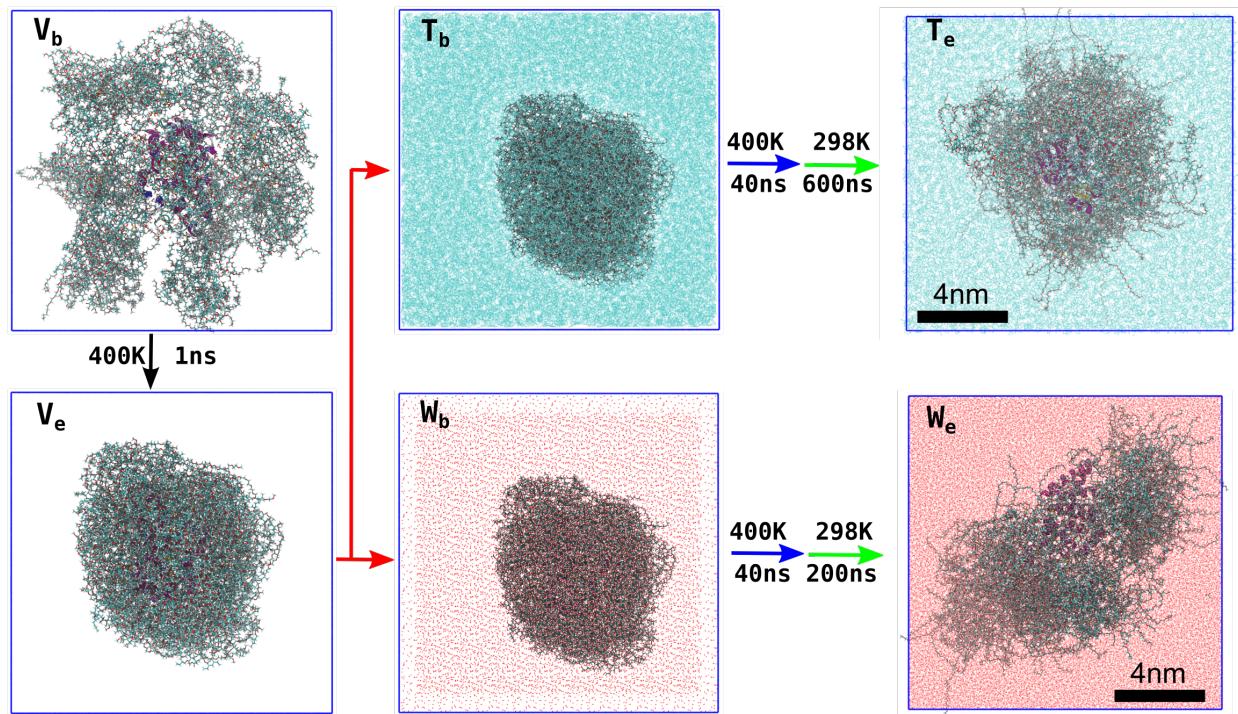
**Fig. S10. SEC trace of RHP.**



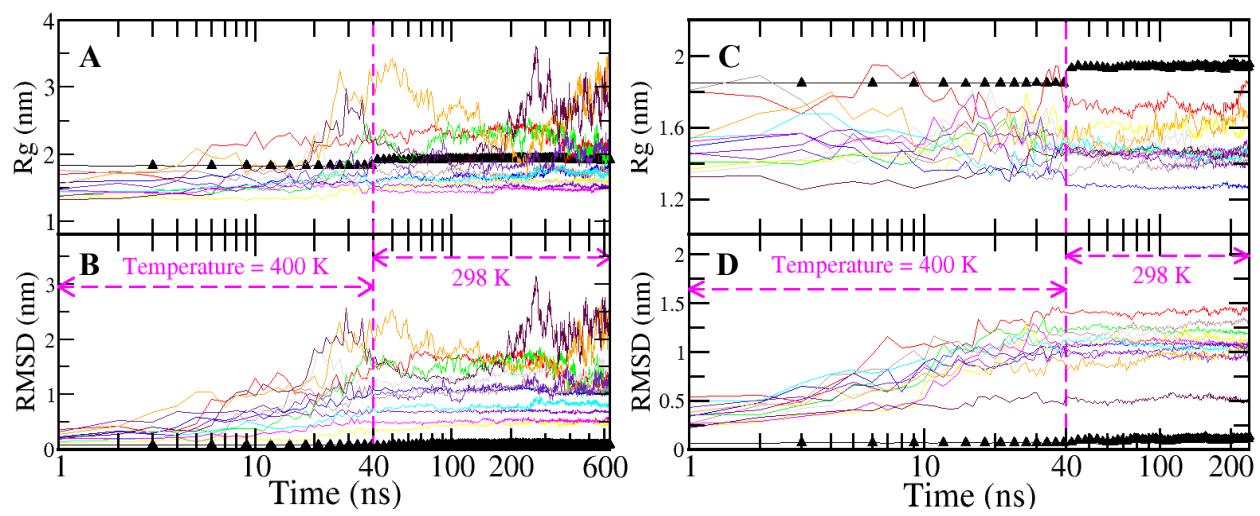
**Fig. S11.**  $^1\text{H}$ -NMR (DMSO- $d_6$ , 400 MHz) superimposition of 12 purified RHPs with the same targeted composition.



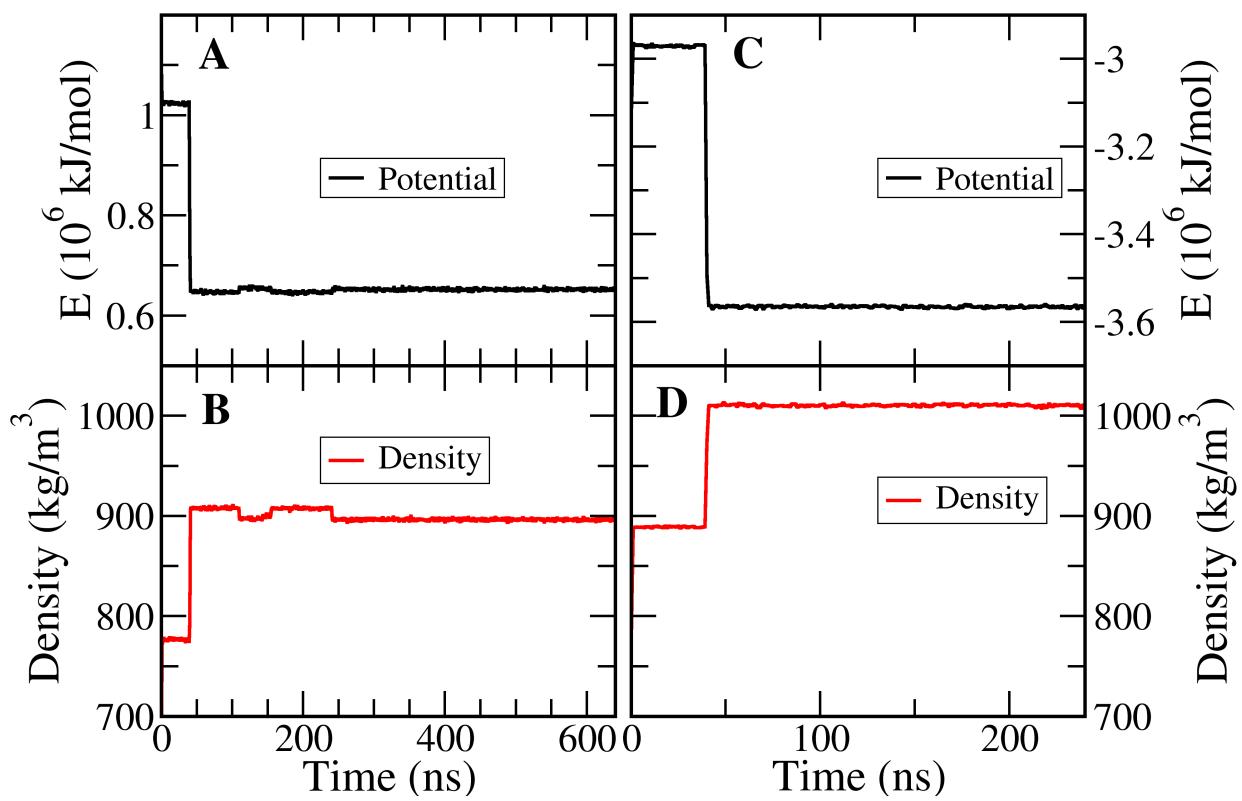
**Fig. S12. Schematic representation of polymer HRP employed in the all-atom MD simulations.** The four types of monomers are (MMA) methyl methacrylate, (2-EHMA) 2-ethylhexyl methacrylate, (OEGMA) oligo(ethylene glycol) methacrylate and (3-SPMA) potassium 3-sulfopropyl methacrylate. The total degree of polymerization  $i + j + k + l = 80$ , with  $i = 40$ ,  $j = 16$ ,  $k = 20$  ( $n = 9$ ) and  $l = 4$ .



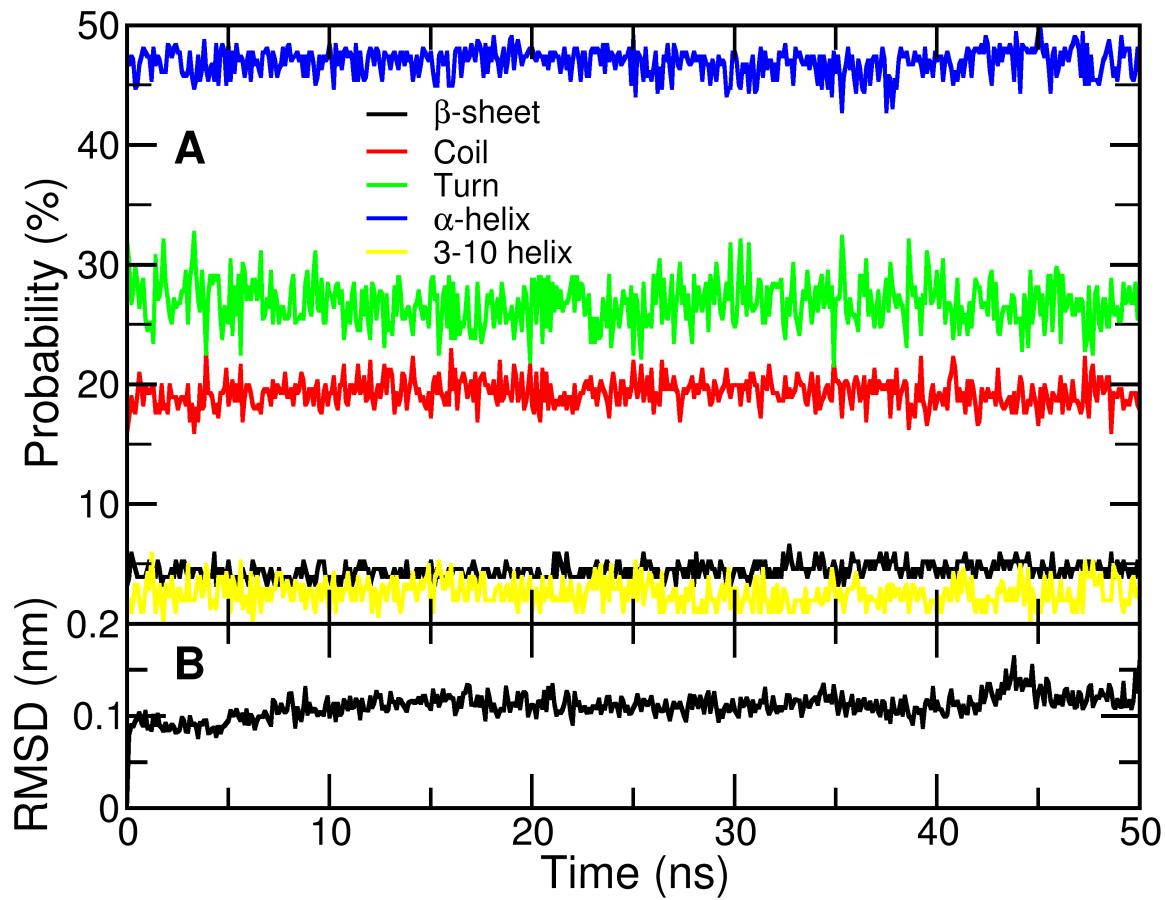
**Fig. S13: All-atom MD simulation snapshots in different simulation processes.** ( $V_b$ ,  $V_e$ ) stand for vacuum simulation box, ( $T_b$ ,  $T_e$ ) for toluene solution, ( $W_b$ ,  $W_e$ ) for water solution, where subscripts b and e correspond to the beginning and the end of the given simulation, respectively. Blue solid lines represent the boundary of the simulation boxes. The rotation movies of the last frames ( $T_e$  and  $W_e$ ) are provided as **Movie S1(A, B)**.



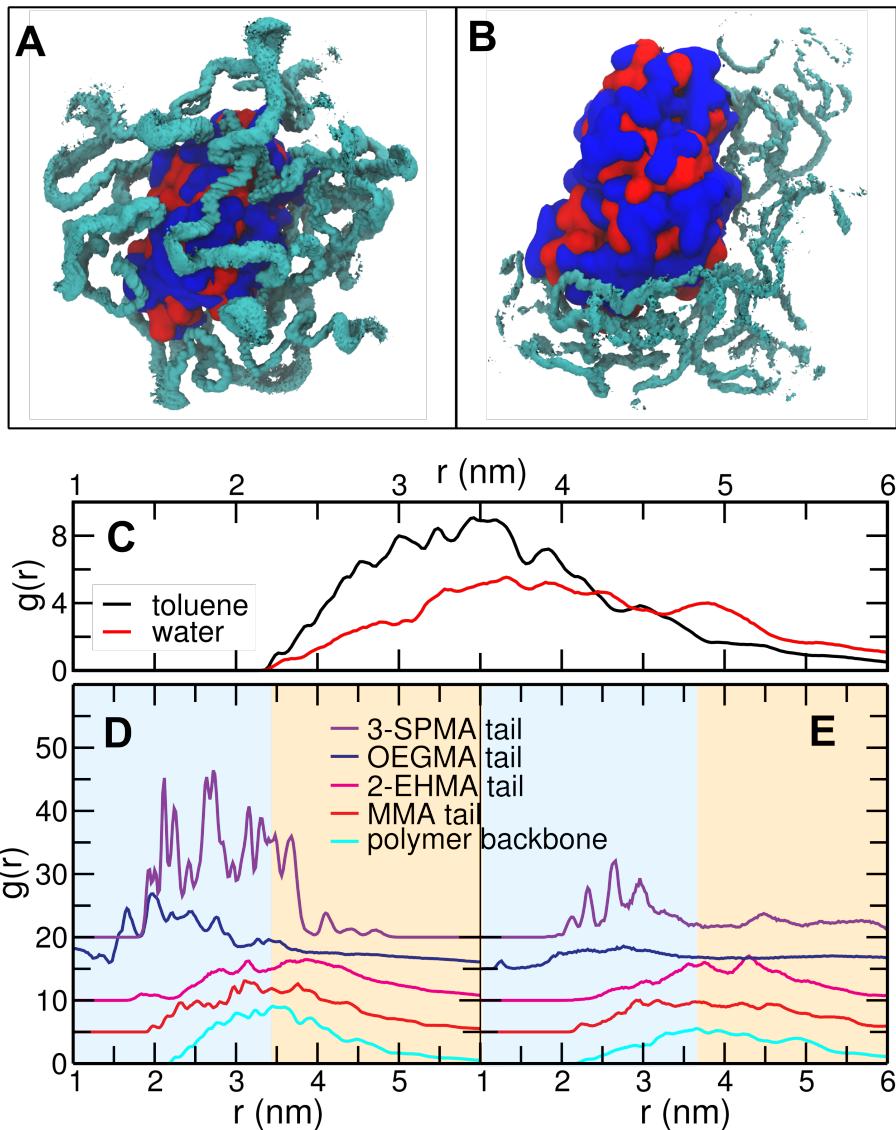
**Fig. S14. Temporal structural properties.**  $R_g$  and RMSD of the backbone atoms of protein (HRP) and the 12 polymer chains in (A, B) toluene and (C, D) water solution all-atom simulations. The results of protein are highlighted using filled triangles; the results for the 12 polymers are colored differently.



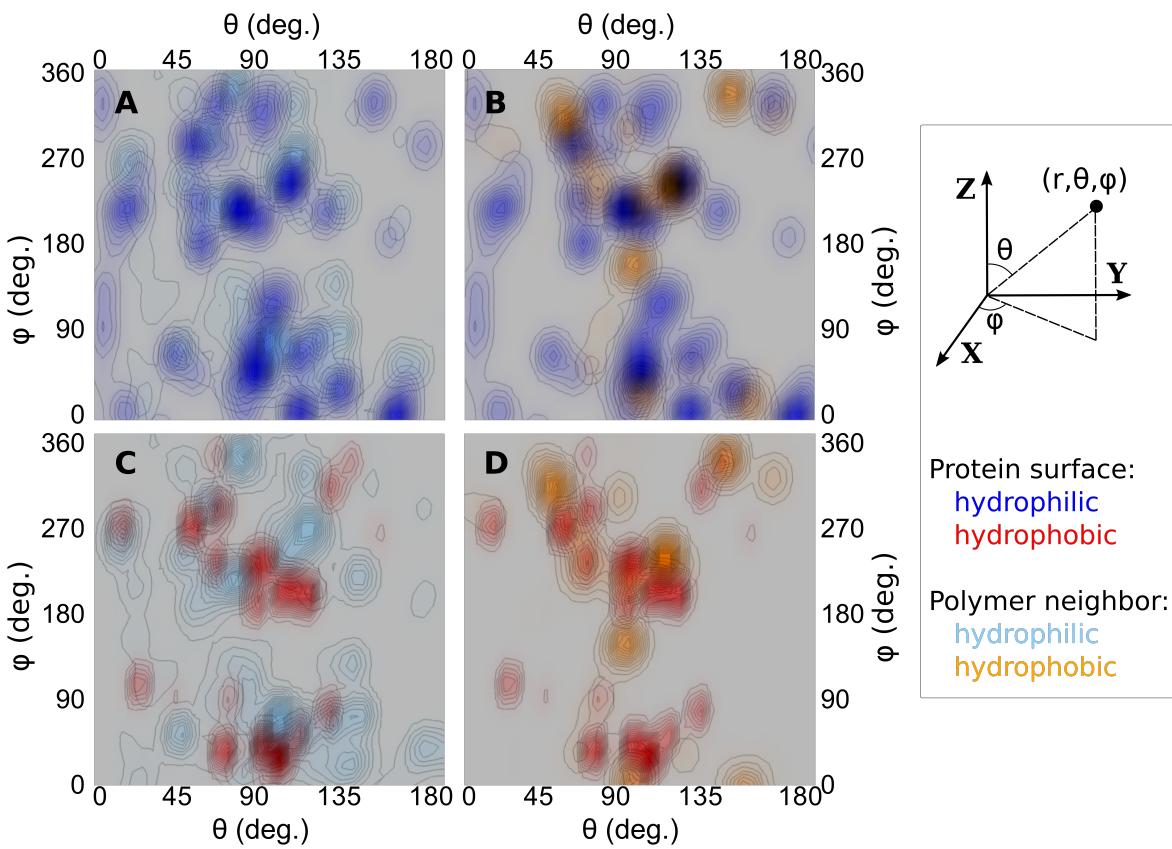
**Fig. S15. Temporal potential energy and density.** (A) System potential energy and (B) system density in the toluene solution simulation. The corresponding data in the water phase simulation are presented in panels (C, D), respectively. Note that the first 40 ns simulations were performed at 400 K for both the toluene and water solution simulations.



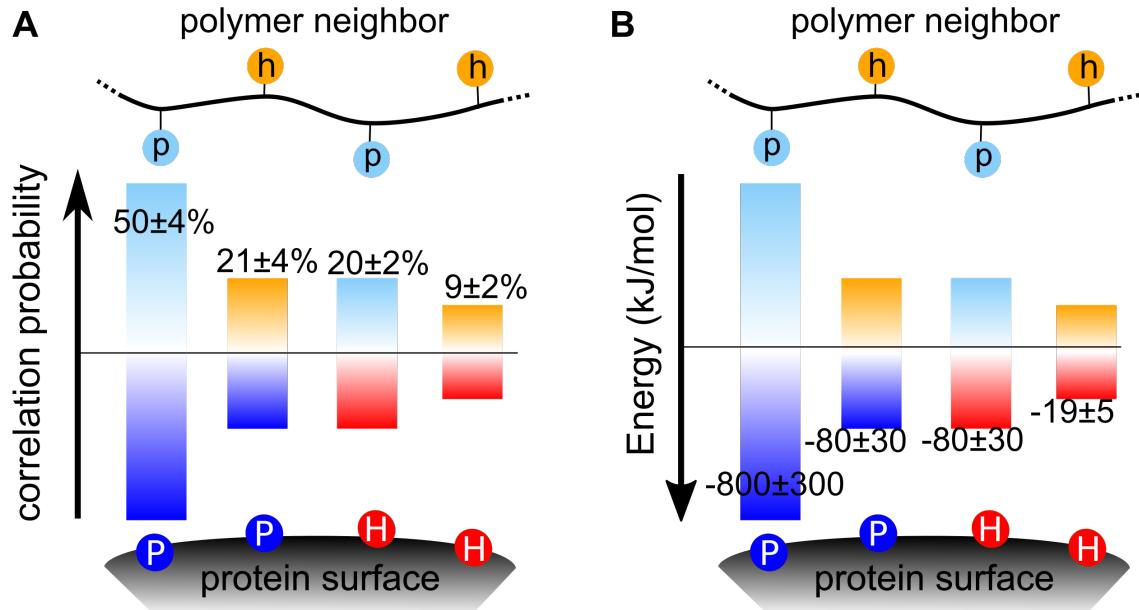
**Fig. S16. Convergence in the control all-atom MD simulation.** (A) The probability of secondary structures and (B) the RMSD of the protein backbone atoms as a function of the simulation time in the control system (see compositions in Table S3).



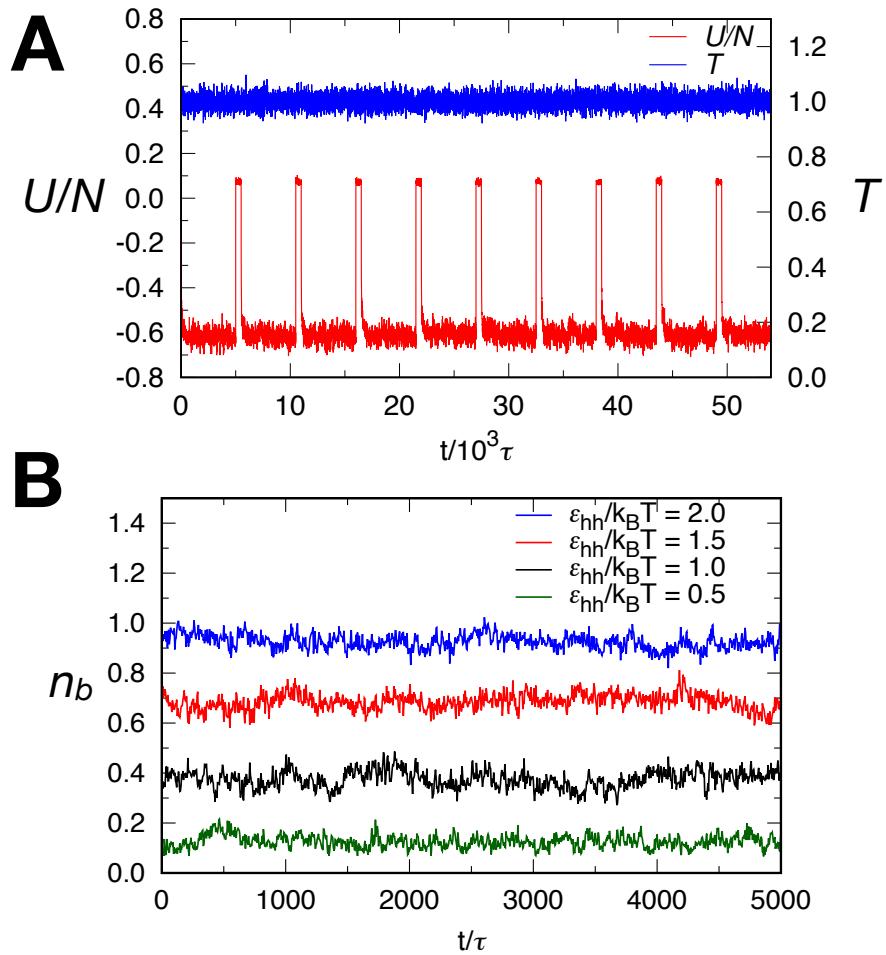
**Fig. S17: Polymers-encapsulated protein.** 3D spatial distribution function of polymer backbone atoms (in cyan) around the protein molecule in (A) toluene and (B) water solutions. The isosurfaces are plotted at 50 times the average density. The hydrophilic/hydrophobic protein surfaces are plotted in blue/red, respectively. Other atoms than the polymer backbone carbons are omitted for display. The corresponding rotation movies are provided as **Movie S2 (A, B)**. (C) Radial density profiles of the polymer backbone around the center-of-mass of HRP backbone in toluene and water. (D) Radial density profiles of the polymer backbone atoms and the last tail atoms of the four monomers (**Fig. S12**) in toluene. The curves are shifted just for the clarity. The areas in light blue and orange are to highlight the inner core and the shell regions based on the primary peak on the polymer density profile in (C). (E) The corresponding radial density profiles in water.



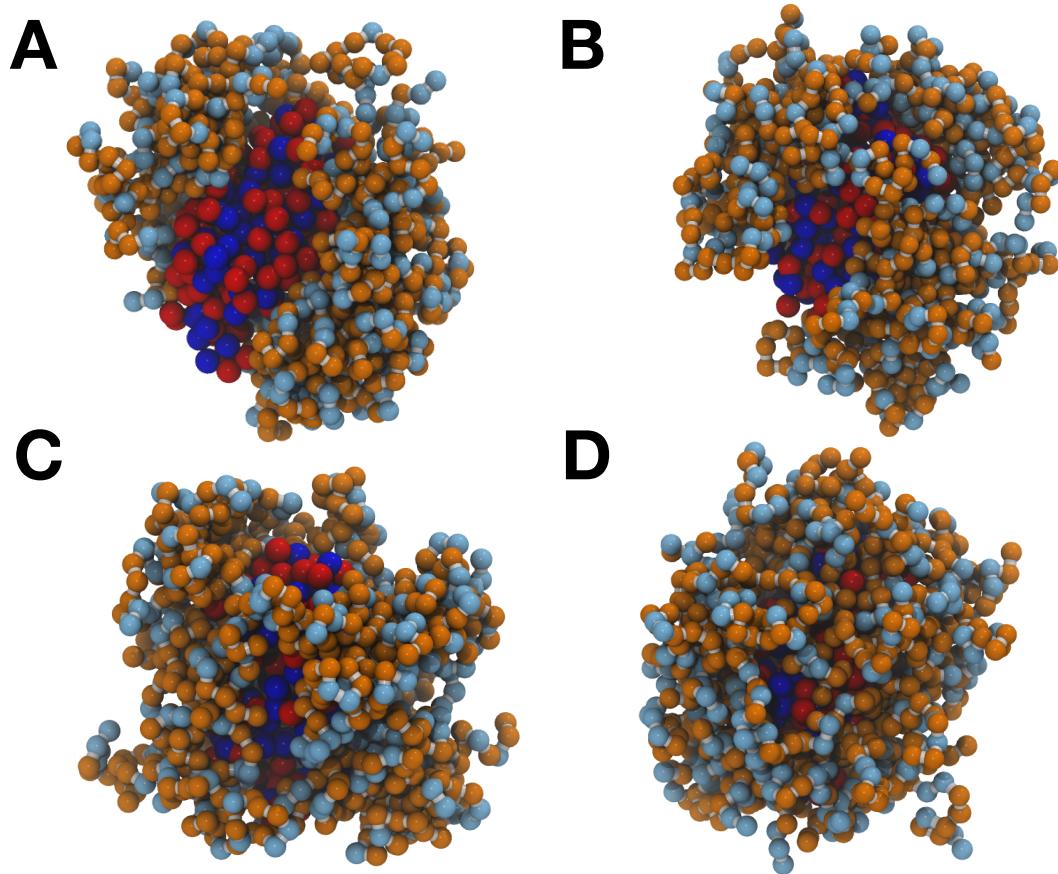
**Fig. S18. Correlations of the hydrophilic/hydrophobic features of the protein surface atoms and those of their nearest polymer neighbors.** (A) The hydrophilic protein surface atoms (in blue) and the hydrophilic polymer neighbors (in light blue) are superimposed to visualize their correlations. Similarly, in (B) the hydrophilic protein surface (in blue) and the hydrophobic polymer neighbors (in orange) are superimposed; in (C) the hydrophobic protein surface (in red) and the hydrophilic polymer neighbors (in light blue); and in (D) the hydrophobic protein surface (in red) and the hydrophobic polymer neighbor (in orange). In these plots, the spherical coordinate is employed based on the center of mass of the protein backbone atoms (inset). For the definition of hydrophilic/hydrophobic protein surface atoms and the polymer neighbors, see the text for the details.



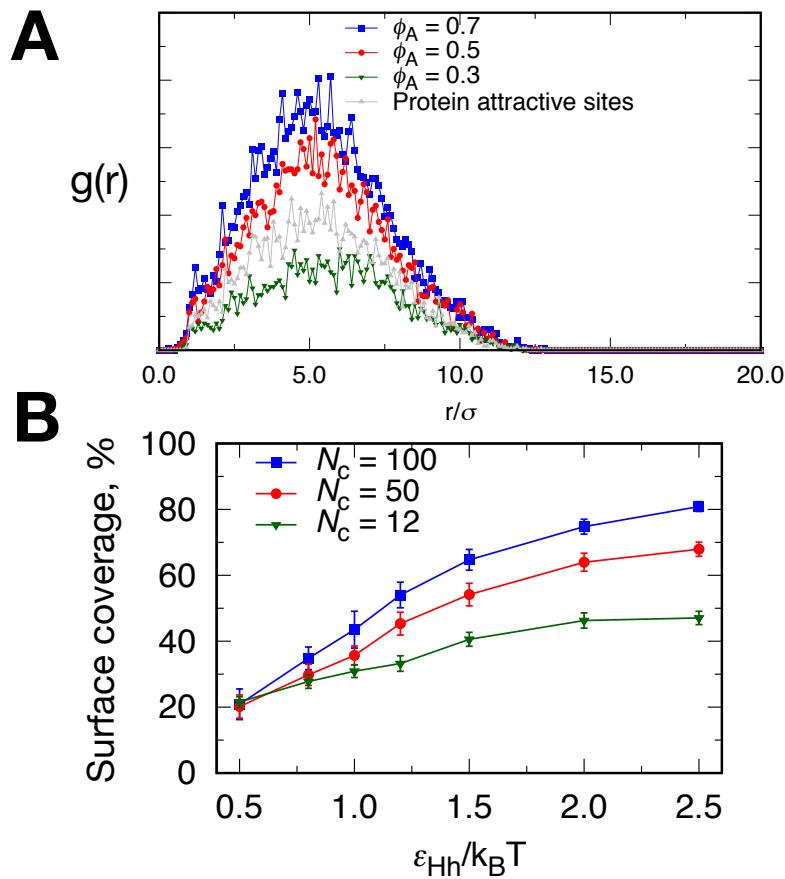
**Fig. S19. Correlations between protein surface and their polymer neighbors.** (A) The correlation probabilities (in %) between hydrophilic (P, blue) and hydrophobic (H, red) protein surface and hydrophilic (p, light blue) and hydrophobic (h, orange) polymer neighbors. (B) The intermolecular interaction energies between the protein surface residues and their polymer neighbor residues. In both panels, the standard deviations are presented as the uncertainties.



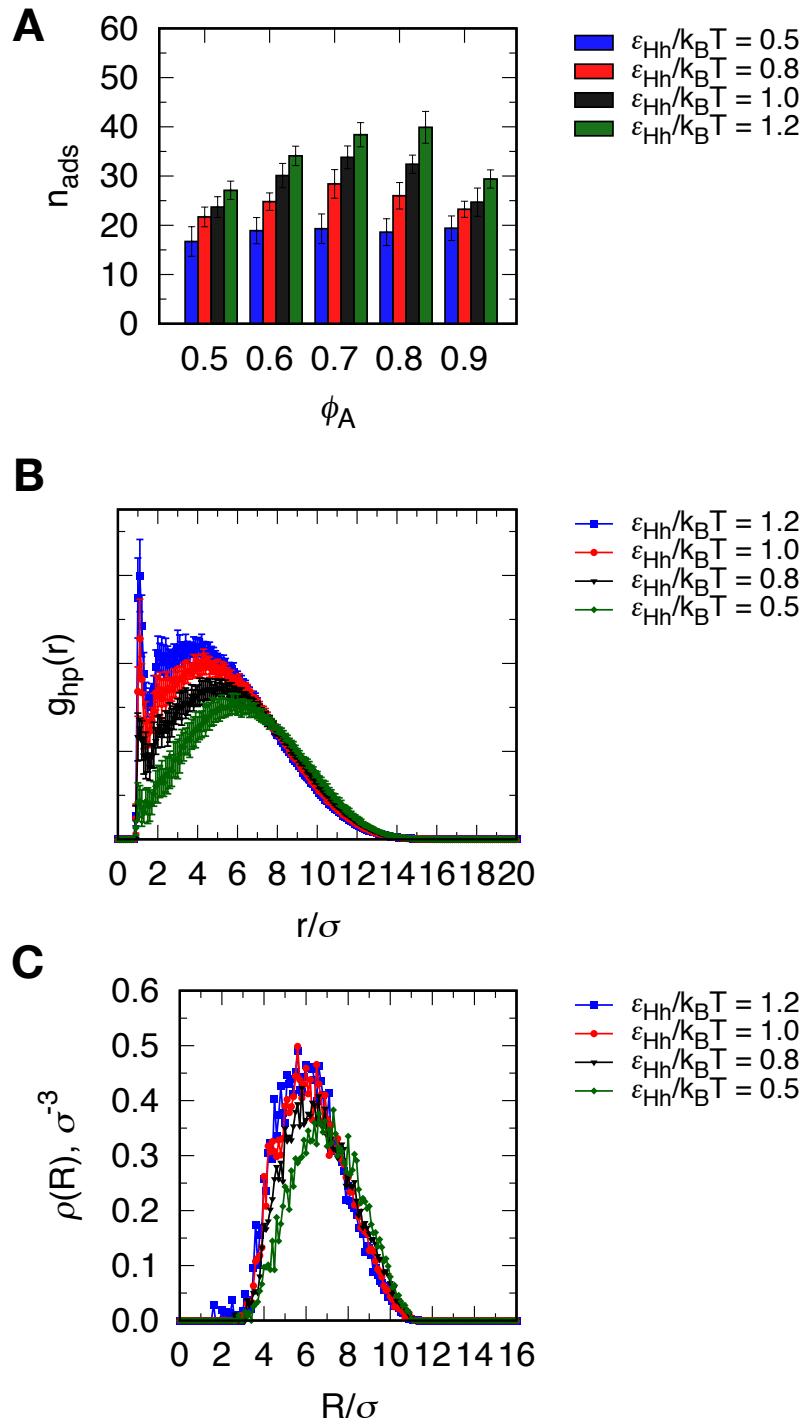
**Fig. S20. Equilibration procedure in coarse-grained simulations.** **(A)** Representative time evolution of the system potential energy (red) and instantaneous temperature (blue) during annealing/equilibration cycles. **(B)** Average number of neighbor per protein attractive sites over time.



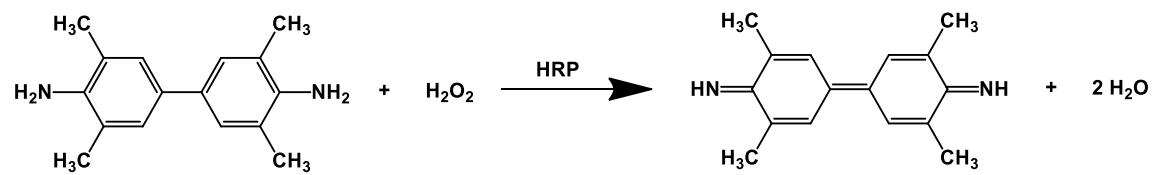
**Fig. S21. Representative snapshots for different adsorption strengths.** (A)  $\varepsilon_{Hh} = 0.5k_B T$ , (B)  $\varepsilon_{Hh} = 0.8k_B T$ , (C)  $\varepsilon_{Hh} = 1.0k_B T$ , (D)  $\varepsilon_{Hh} = 1.5k_B T$ . In these cases,  $\phi_A = 0.5$  and the attraction between polymer adsorbing beads is  $\varepsilon_{hh} = 0.8k_B T$ .



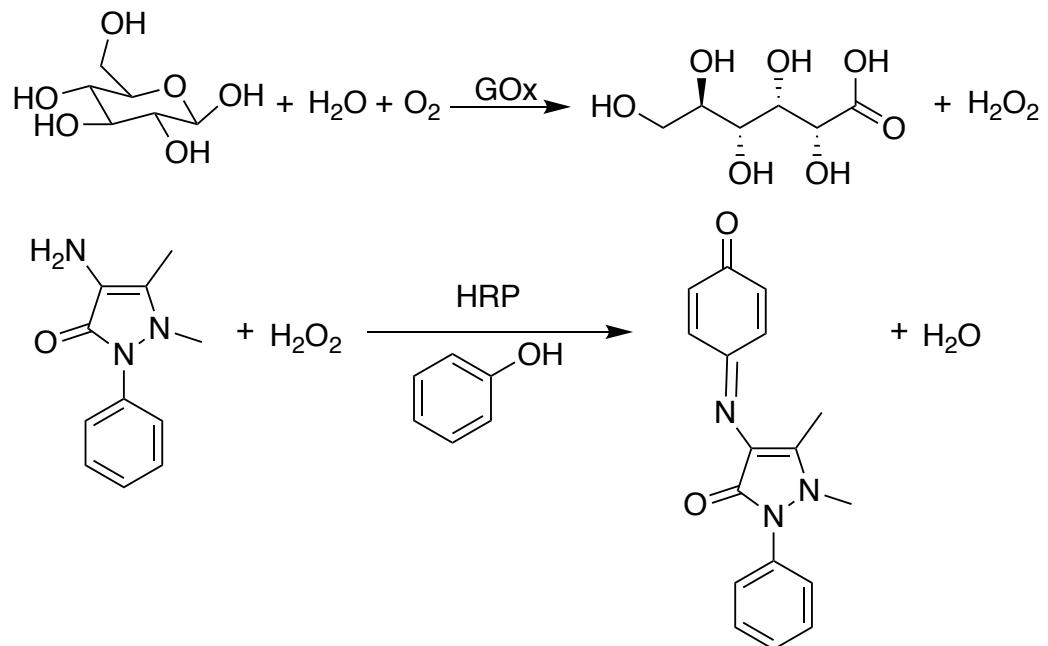
**Fig. S22. Roles of the fraction of the adsorbing polymer components and of the polymer-protein molar ratios.** (A) Pair correlation functions between adsorbed polymer beads for different fractions of the adsorbing component and between protein attractive sites. (B) Surface coverage as functions of the adsorption strengths for different number of polymer chains. Error bars are the standard deviations from ensemble-averaging over 10 independent configurations.



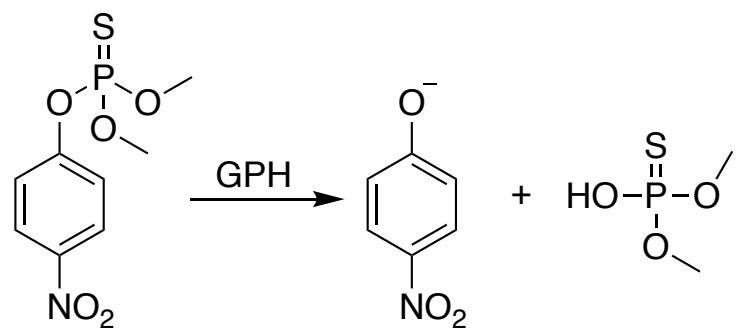
**Fig. S23. Structural analysis of the adsorbed polymer chains.** (A) Number of adsorbed polymer chains as function of the fraction of the adsorbing polymer beads; the number of polymer chains in total is  $N_C = 50$ , (B) the pair correlation functions of the adsorbing beads and polymer attractive sites for different adsorption strengths; (C) the radial density profiles of the polymer beads from the protein center of mass for different adsorption strengths.



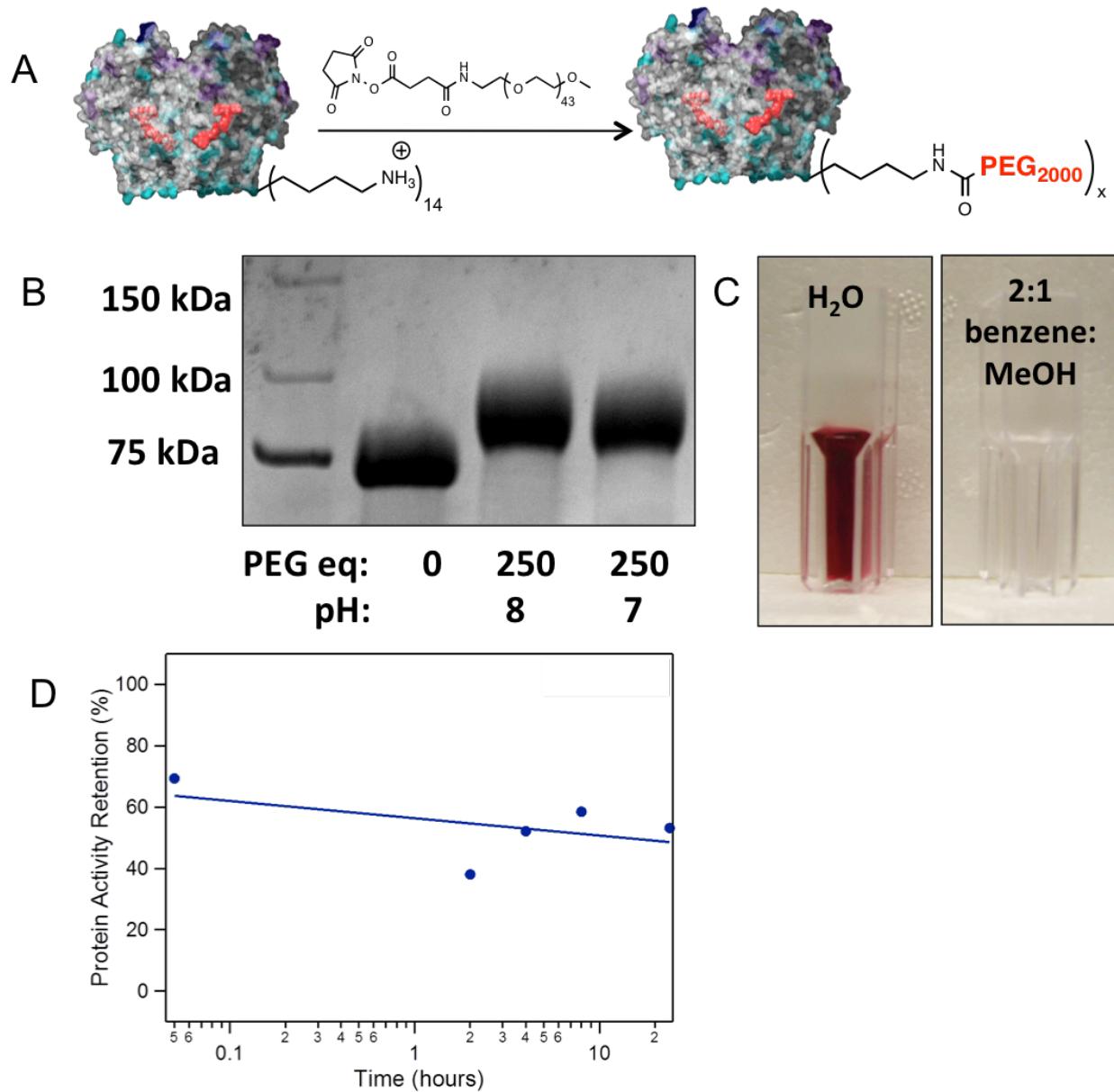
**Fig. 24. Colorimetric assay to evaluation HRP activity.**



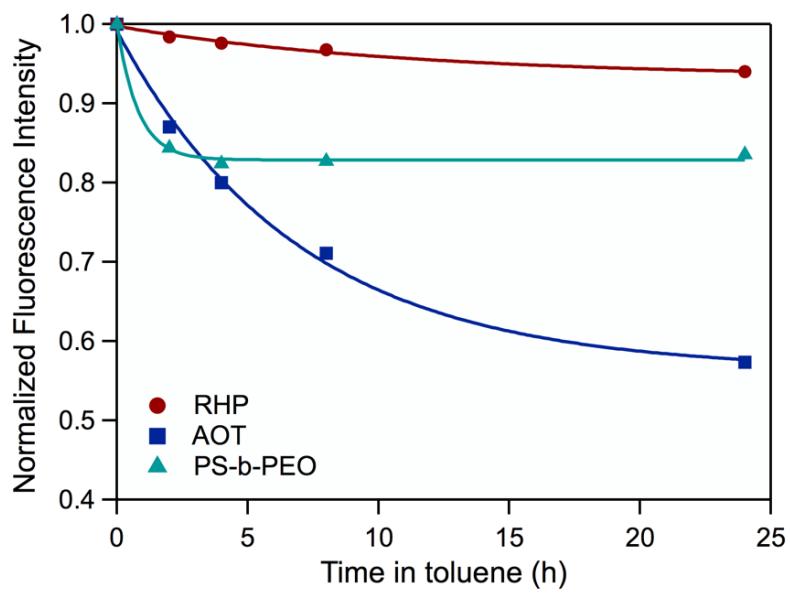
**Fig. S25.** Colorimetric assay to evaluation GOx activity.



**Fig. S26. Colorimetric assay to evaluation OPH activity.**



**Fig. S27. Comparison of PEGylation and RHP approach for protein dispersion and stabilization in solvents.** (A, B) Only limited copies of PEG can be attached to the GOx surface. (C) There is significant GOx activity loss when dispersed in benzene/methanol (2:1) solvent mixture. (D) The GOx/RHP can retain ~50% of native activity after 24 hr dispersion in toluene. The solvent selection is governed by the PEG and RHP solubility.



**Fig. S28.** The fluorescence spectra of GFP in toluene over 24 hours with the presence of RHP, suggesting minimal changes in GFP structure or toluene penetration. Similar experiments were performed using AOT and PS-*b*-PEO polymeric surfactants.

**Table S1.** Reactivity ratios used in 4-monomer copolymerization simulation.

	MMA	OEGMA	EHMA	SPMA
MMA	0.89	1	1	1
OEGMA	1	1.09	1.09	1.09
EHMA	1	1.09	1.09	1.09
SPMA	1	1.09	1.09	1.09

**Table S2.** Physico-chemical properties of RHP.

mol% of <b>1</b>	mol% of <b>2</b>	mol% of <b>3</b>	mol% of <b>4</b>	$M_n^a$ (kg/mol)	$D^a$
50/46 <sup>b</sup>	25/32 <sup>b</sup>	20/18 <sup>b</sup>	5/4 <sup>b</sup>	30.2	1.38

<sup>a</sup>Determined by SEC.

<sup>b</sup>Theoretical/experimental mol% determined by gravimetry and quantitative  $^{13}\text{C}$  NMR, respectively.

**Table S3.** Compositions of HRP/RHP Mixtures in the Different All-Atom MD Simulations.

	protein <sup>a</sup>	polymer <sup>b</sup>	toluene <sup>c</sup>	water <sup>c</sup>
<b>vacuum</b>	1	12	-	-
<b>aqueous solution</b>	1	12	-	92151
<b>organic solution</b>	1	12	15430	-
<b>control</b>	1	-	0	31246

a) Protein contains one 1H55 molecule and its ligands of HEME-Fe, acetate ion, Ca<sup>2+</sup> ions and 403 water molecules.

b) 12 polymer chains with *DP* = 80, and 48 K<sup>+</sup> counterions.

c) Solvent: toluene or water.

**Table S4.** Probabilities (%) of Secondary Structures in the Three Different All-Atom Simulations

	$\beta$ -sheet	Coil	Turn	$\alpha$ -helix	3-10 helix
<b>Control simulation</b>	4.4 ±0.6	19±1	27±2	47±1	3±1
<b>toluene solution</b>	4.4±0.6	19±1	26±2	47±1	4±1
<b>aqueous solution</b>	4.4±0.7	19±1	28±2	47±1	2±1

**Movie S1(A, B):** 3D rotation movie of the last all-atom MD simulation frames in (A) toluene and (B) water solutions. Only polymer (in black) and protein atoms are displayed for clarity. See also **Fig. S13(W<sub>e</sub>, T<sub>e</sub>)**.

**Movie S2(A, B):** 3D spatial distribution functions of the polymer backbone atoms around the protein in the all-atom MD simulations on (A) toluene and (B) water solutions. The isosurfaces are plotted at 50 times the average density. Polymer backbone atoms are colored in cyan and hydrophilic/hydrophobic protein surface in blue/red, respectively. Only polymer backbone and protein atoms are displayed for clarity. See also **Fig. S17(A, B)**.

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