

# Nucleation limited assembly and polarized growth of a *de novo*-designed allosterically modulatable protein filament

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## Abstract

The design of inducibly assembling protein nanomaterials is an outstanding challenge. Here, we describe the computational design of a protein filament formed from a monomeric subunit which binds a peptide ligand. The cryoEM structure of the micron scale fibers is very close to the computational design model. The ligand acts as a tunable allosteric modulator: while not part of the fiber subunit-subunit interfaces, the assembly of the filament is dependent on ligand addition, with longer peptides having more extensive interaction surfaces with the monomer promoting more rapid growth. Seeded growth and capping experiments reveal that the filaments grow primarily from one end. Oligomers containing 12 copies of the peptide ligand nucleate fiber assembly from monomeric subunit and peptide mixtures at concentrations where assembly occurs very slowly, likely by generating critical local concentrations of monomer in the assembly competent conformation. Following filament assembly, the peptide ligand can be exchanged with free peptide in solution, and it can be readily fused to any functional protein of interest, opening the door to a wide variety of tunable engineered materials.

## Main text

Protein filaments in nature exhibit complex and tunable assembly and disassembly dynamics. For example, microtubules and actin exhibit dynamic instability, where filaments form upon binding of GTP or ATP. Glk1 (Glucose kinase 1) binding to glucose and CTP synthase binding to its substrate (ATP, UTP and glutamine) both allosterically regulate assembly by promoting a filament-competent conformation. IMPDH (Inosine-5'-monophosphate dehydrogenase) binds to regulator adenine and guanine nucleotides directly at an oligomerization interface. PRS1(5-phospho-ribosyl-1(alpha)-pyrophosphate synthetases 1) binds phosphate at its regulator site, which is important for stabilizing the nearby assembly interface (which is completely disordered in the free protomers). Microtubules originate from centrosomes, facilitated by  $\gamma$ -tubulin ring complexes that act as nucleation and anchoring sites (1). Actin filaments, on the other hand, nucleate in response to various ligands such as Arp2/3 complex, which binds to existing actin filaments, promoting the formation of new branches (2).

Does the complex behavior of natural fibers reflect extremely fine evolutionary optimization, or can it arise in simpler systems? De novo protein design provides an approach to address this question. Computational design methods have been used to generate static fibers from ordered protein monomers, and fibers whose assembly and disassembly are pH dependent (3). We reasoned that a designed fiber system in which the monomeric subunit binds an exogenous ligand over an extensive interface independent from the interfaces involved in fiber formation could exhibit ligand-dependent assembly, and provide a simple model system for studying the dynamics of protein filament assembly independent of evolutionary optimization.

We set out to design self-assembling protein filaments from a designed protein monomer (RPB\_PLP3\_R6) (4) with an extended binding interface for a peptide ligand composed of multiple repeats of the sequence pro-leu-pro (PLP). We chose this system as a starting point because ligand binding affinity can be readily tuned: peptides with 3, 4, 5, and 6 tandem PLP repeats bind with increasing affinity to the monomer, which is composed of six repeating units which each binds one PLP in the peptide. We docked the crystal structure of RPB\_PLP3\_R6 into protein filament models using the computational method described in Shen *et al.* (5), ensuring the peptide binding site was not at any of the filament-generating interfaces (Fig. 1). We generated 91569 helical filament backbones and selected 69 designs for experimental testing based on predicted energy of fiber formation and other metrics (see Methods). The designs were fused to green fluorescent protein (GFP), expressed in *Escherichia coli* under the control of a T7 promoter, and purified by immobilized metal affinity chromatography (IMAC). The peptide ligand (PLP)<sub>6</sub> was added to concentrated IMAC eluates and fiber formation was examined by fluorescent microscopy. Most of the designs expressed solubly, and while none formed fibers in the absence of the peptide, one design, HP35, formed fibers upon the addition of the peptide. Negative stain EM examination confirmed the formation of micron-length filaments with the expected morphology (fig S1).

We froze the HP35-(PLP)6 fibers, generated an initial cryoEM structure using *ab initio* 3D reconstruction in cisTEM (6) and performed further helical refinement in Relion (7). The structures were solved to 4.0 Å resolution with the protein and the majority of the peptide resolved, with 2.2 Å r.m.s.d. over three chains containing all unique interfaces (Fig. 2). The backbone and side chain conformations at the subunit interfaces are very close to those in the design model. As the peptide is not directly involved in the fiber interface, the peptide dependence on fiber formation must reflect the allosteric effects of the peptide on the conformation of the monomer; the peptide could lock the monomer into an assembly-competent state, or disrupt weak monomer-monomer interactions incompatible with fiber assembly involving the extended peptide binding site.

We next examined the dependence of fiber formation on the affinity and extent of the monomer-peptide interaction. HP35 fiber formation could be induced with (PLP)4, (PLP)5, and (PLP)6, but not the shorter (PLP)3 (Fig. 3 A and B, Fig S2). The rate of fiber formation increased with peptide length; the half times for 4, 5, and 6 PLP repeat fibers were 1250, 680, and 250 minutes respectively.

We used a photobleaching experiment to investigate the exchangeability of peptides on fibers. For this experiment, we used a more polar version of the fiber, HP35\_B3, with a smaller binding pocket that binds to and forms fibers in the presence of the shorter (PLP)3 and (PLP)4 peptides (Fig. 3B bottom, Fig S3). After forming fiber with cy5 labeled (PLP)4, segments were bleached by laser illumination; after 20 minutes there was no recovery suggesting little or no exchange with unbleached peptide. To monitor peptide incorporation into the fiber over longer times, we added cy5 labeled (PLP)4 to unlabeled HP35\_B3 to form the fiber and then added FITC labeled (PLP)4 peptide to monitor incorporation of the new peptide. We found the intensity of FITC increased over hours to days (Fig. 3C and D), indicating that peptides can add to the fiber once formed.

We studied fiber extension independent of nucleation using pre-assembled GFP-labeled fiber as seeds and then incubated with mKO2-labeled peptide and monomer. We observed robust seeding of growth, with 45% of green fibers having a newly formed red tip (Fig. 4A and B). Unexpectedly, in almost all cases the growth was just from one end: amongst the 740 fibers we analyzed, 340 had one red tip and only 2 had two red tips. Given that 23% of fiber ends (two ends per fiber) have a red tip, we would predict (if the fibers do not have polarity) that the number of fibers with two tips should be 5.3% ( $0.23 \times 0.23$ ), but the observed value is clearly much lower than this (Fig. 4C). This strongly suggests that the fibers have polarity, and grow from only one end, similar to several natural fibers (8).

To determine which end the fibers preferentially grow from, we converted the monomeric subunits into fiber capping units by mutating the interfaces that drive fiber formation on one side of the monomer or the other (Fig. S4). We define one of these as the “N-cap” as it contains mutations closer to the monomer N terminus; the other we refer to as the “C-cap”. We first tested the caps

by growing new HP35 fiber without seed in the presence of N-cap or C-cap; we found that both inhibit fiber formation (Fig. S5). To determine which end the seeded filaments grow from, we generated green fiber seeds as described above, incubated them with either N-cap, C-cap, or no cap, and then added far-red cy5 labeled new protein and peptide to monitor growth (Fig. 4D). In the absence of cap, 8 red tips were found on 27 fibers; in the presence of N-caps, 0 red tips were found on 19 green fibers, and in the presence of C-cap, 7 red tips were found on 14 fibers, (Fig. 4E-H). We interpret this blocking of fiber extension by the N-caps to indicate that growth from the fiber N-terminal end is strongly favored over growth from the C-terminal fiber end.

While protein fibers in biology exhibit nucleation-dependent formation, for example, microtubules are nucleated by the  $\gamma$ -tubulin ring complex, and actin by the Arp2/3 complex, such behavior to our knowledge has not been observed in designed protein fiber systems. A multivalent assembly of monomeric fiber subunits (with a subset of the fiber-forming interactions removed by mutation) was found to capture already existing fibers, but did not promote fiber growth at sub-critical concentrations (5). We explored whether assemblies of peptide ligand geometrically matched to the end of the HP35 fiber could nucleate assembly by fusing 4 copies of PLP5 connected by flexible GS linkers to a C3 trimeric oligomer compatible with the fiber structure; we reasoned that this could increase the local concentration of fiber forming-competent monomer and promote nucleation (Fig. 5A). Addition of small amounts of nucleator to HP35 protein and PLP6 peptide substantially increased the rate of fiber formation (Fig. 5B). Green fluorescent labeled nucleator was detectable at the end of the far-red labeled fiber (Fig. 5C). Fiber extension remains dependent on the peptide ligand: addition of nucleator to HP35 alone did not lead to fiber growth (Fig. 5D). This absolute dependence of fiber assembly on the nucleator opens up possibilities for control of assembly at specific sites in cells and for amplification of signals in diagnostics assays (a very small amount of nucleator can generate a large assembly that can be read out using fluorescence or other means).

## Discussion

The assembly of the HP35 fiber is considerably more intricate than previously designed constitutively forming fibers and approaches the complexity of protein filaments of nature. Assembly is allosterically controlled, directional, and sensitive to the addition of nucleators. In this section, we review these properties and discuss their origins.

Assembly of the HP35 fiber is absolutely dependent on the addition of the effector peptide. This control is allosteric, as the peptide is distant from the subunit-subunit interfaces that generate the fiber. This allosteric regulation likely reflects the large hydrophobic groove in the subunit that forms the binding pocket for the effector; in the absence of an effector, this groove may close on itself or drive the formation of transient assemblies, both of which could block fiber assembly. In a manner somewhat similar to the binding of steroid hormones to steroid hormone receptors (9),

the binding of effector could promote assembly both by driving completion of folding and locking the subunit in a fiber assembly competent state and by disfavoring oligomeric species.

The observation of polarized growth of the HP35 fiber is notable as it requires no external energy input. The energy for adding a new subunit in both directions must be equivalent, and hence the different growing speeds at the two ends must arise from kinetic effects. In contrast to the previously designed filament where the four interfaces are separated (3, 5), one interfacial helix of HP35 contacts the other two adjacent subunits (Fig 2E) completing an overall triangle-shaped interface. In contrast, the addition of a subunit on the other end leads to the formation of two disconnected half interfaces (Fig. S6). The completion of the triangle interface may be more concerted than the formation of the two half-interfaces, and be kinetically favored. Alternatively, the effector-bound HP35 subunit may have a conformation slightly different from the fiber-integrated form, with a better fit for binding to the N side than the C side.

The nucleation of the HP35 fiber by oligomers presenting large numbers of subunits provides a means to direct fiber formation to specific locations. The nucleator likely functions by increasing the local peptide concentration, creating a large number of peptide-bound HP35 locally that are competent for fiber formation. This genetically encoded nucleator could enable the directing of filament growth to specific locations both in cells and in higher-order materials, and form the basis for signal amplification in detection applications.

More generally, our results suggest that the complex assembly dynamics of many protein filament systems in biology may not have required extensive evolutionary optimization. More concretely, our results show that a considerable increase in the richness of assembly behaviors accompanies the transition from constitutive single-component fibers as in our previous work (5) to two-component effector-dependent filament such as the one studied here, likely due to the greater conformational heterogeneity of two-component systems.

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## Methods

### Computational Design Strategy

The helical docking and design method (5) was applied to RPB\_PLP3\_R6 to generate models of helical filament designs. Design trajectories were filtered using the following criteria: a Rosetta Energy Unit difference greater than -15.0 between the polymeric (bound) and monomeric (unbound) states, an interface surface area over 700 Å<sup>2</sup>, Rosetta shape complementarity above 0.62, and fewer than 5 unsatisfied polar residues. Designs meeting these criteria underwent manual refinement, with non-contributory mutations reverted to their original state. The top-scoring design for each configuration was then included in a final protein set for experimental validation.

To tune and facilitate the exchangeability of peptides on the original HP35 fiber, we designed and screened a series of HP35 polar variants, in the presence of shorter PLP repeating peptides, especially (PLP)3 and (PLP)4. In brief, the asymmetry unit of HP35-(PLP)6 fiber model was extracted and served as the input for redesign. The PLP repeating peptide was shortened from six repeats to three or four repeats. The newly exposed surface residues along the peptide binding interface on the HP35 monomer were identified and redesigned by Rosetta FastDesign, in the context of the original fiber filament symmetry. The design process was guided by the Rosetta score function “REF2015”. For the successful design HP35\_B3, the interface surface residues “36, 37, 40, 41, 44, 84, 87, 88, 91, 200, 247, 248, 251, 292, 299” were chosen for redesign.

### Protein Expression and Purification

Synthetic genes for 69 designs were optimized for expression in *Escherichia coli* and synthesized by IDT. These genes were cloned into the pET29b+ based vectors with either N- or C-terminal GFP fusion, inserted between the NdeI and XhoI restriction sites. The constructs were transformed into BL21\* (DE3) *E. coli* competent cells and grown in 50 ml of Terrific Broth medium containing 200 mg L<sup>-1</sup> kanamycin. Protein expression, driven by a T7 promoter, proceeded for 24 hours at 37°C using Studier autoinduction(10), after which cultures were harvested by centrifugation. Cell pellets were resuspended in Tris-buffered saline (TBS) and lysed with Bugbuster detergent. The soluble fraction, clarified by centrifugation, was purified using Ni<sup>2+</sup>-immobilized metal affinity chromatography (IMAC) on Ni-NTA Superflow resin. The resin-bound lysate was washed with 10 column volumes of 40 mM imidazole and 500 mM NaCl, then eluted with 400 mM imidazole and 75 mM NaCl. Both soluble and insoluble fractions were analyzed by SDS-PAGE. Samples showing protein bands at the expected molecular weight were selected for electron microscopy screening. For further characterization, selected designs were scaled up to 0.5 L, expressed under the same conditions for 24 hours at 37°C, and harvested by centrifugation. Cell pellets were resuspended in TBS, lysed via microfluidization, and purified using the same method.

### Negative Stain Electron Microscopy

The soluble fractions were concentrated in a buffer (25 mM Tris, 75 mM NaCl, pH 8) for electron microscopy screening. A 6 µL droplet (1 µL sample immediately diluted with 5 µL buffer) was placed onto negatively glow-discharged, carbon-coated 200-mesh copper grids, washed with Milli-Q water, and stained

with 0.75% uranyl formate (pH 4.0) as previously described (11). The screening was performed using a 120 kV Talos L120C transmission electron microscope (ThermoFisher). Images were captured using a bottom-mounted Teitz CMOS 4k camera system and processed for enhanced contrast with Fiji software(12)

## Cryo-Electron Microscopy

CryoEM samples were prepared by applying protein to CFLAT 2/2 holey-carbon grids, blotting away liquid, and plunging into liquid ethane using a Vitrobot (ThermoFisher). Movies were acquired on a Glacios microscope (ThermoFisher) equipped with a K-2 Summit Direct Detect camera (Gatan Inc.) operating in counting mode, with a pixel size of 1.16 Å/pixel, 50 frames, and a total dose of 65 electrons/Å<sup>2</sup>. Automated data collection was performed using Leginon (13). The data processing workflow is summarized in Fig. S7. Movies were aligned using `group1_data = output_df[output_df['construct'].isin(group1)]`  
`group2_data = output_df[output_df['construct'].isin(group2)]`  
`group1_data = output_df[output_df['construct'].isin(group1)]`  
`group2_data = output_df[output_df['construct'].isin(group2)]`  
`group1_data = output_df[output_df['construct'].isin(group1)]`  
`group2_data = output_df[output_df['construct'].isin(group2)]`g the Relion (7) implementation of MotionCor2 (14), and CTF parameters were estimated using Gctf (15). Helices were picked manually from a subset of images and subjected to 2D classification to generate templates for subsequent automated picking in Relion. Particles from automated picking were subjected to further 2D classification using Relion. *Ab initio* reconstruction in cisTEM (6), using a subset of particles from automated picking, was used to generate an initial 3D map from which initial helical symmetry parameters were also estimated. Helical auto-refinement was then performed in Relion using the *ab initio* map as a starting reference, with local searches of helical rise and rotation parameters enabled. Bayesian polishing as well as beamtilt, anisotropic magnification, and defocus refinement were performed in Relion. Density modification was performed using ResolveCryoEM in Phenix (16, 17). Atomic models were refined into cryoEM maps using ISOLDE(18), followed by real-space refinement in Phenix, with rotamer and Ramachandran restraints disabled and with reference restraints imposed by the input starting model. Cryo-EM data collection, refinement, and validation statistics are summarized in Table S1.

## Fluorescent Microscopy

Sample preparation:

1uL of fiber solution was diluted in 1mL of TBS buffer (25 mM Tris, 75 mM NaCl, pH 8), and 50 uL of diluted sample was added to the well for imaging.

Image acquisition

IN Cell Analyzer 2500HS

Fibers were attached directly to the glass surface of a 384-well plate (Greiner Bio-one, cat# 781892, <https://www.gbo.com/>) and allowed to stick for 5 minutes before washing to remove the background signal. Images were automatically sampled using an IN Cell Analyzer 2500HS and a 60X 0.95 NA Plan Apo CFI objective (Nikon, 12605). Fibers were illuminated with a seven-color Solid State Illuminator (SSI) for fluorescence excitation. Fluorescent signals were acquired sequentially using the following filter setup: Green (excitation 473/28 nm, emission 511/23 nm), Red (excitation 575/30 nm, emission 623/42 nm), and Far-red (excitation 631/28 nm, emission 684/24 nm). Imaging was controlled using the IN Cell



Analyzer 2500 HS software version 7.4 and the light was collected on a sCMOS camera without binning. Between 15-25 fields of view spaced evenly throughout the well were imaged per condition.

## OMX

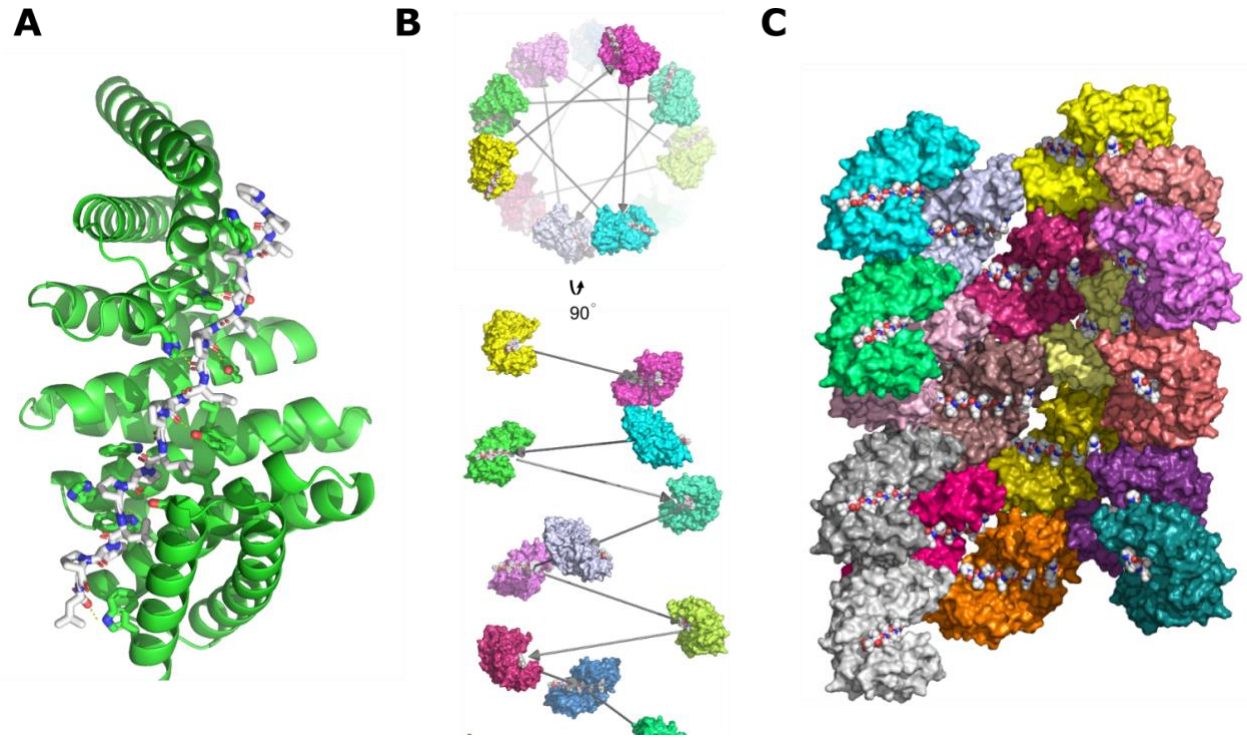
Fibers were attached directly to the glass surface of either an Ibidi  $\mu$ -Slide 0.5 or an Ibidi 18-well glass-bottom coverslide (cat# 806076, 81817, respectively. Ibidi, <https://ibidi.com/>). Two-color images of fibers were acquired with a commercial OMX-SR system (Leica-microsystems) equipped with an Olympus Apo N 60 $\times$  1.49 NA TIRF oil immersion lens. Topica diode lasers with wavelengths of 405nm, 488nm, and 640nm were used for excitation. Emission was collected on three separate PCO.edge sCMOS cameras. 256 $\times$ 256 images (pixel size 6.5  $\mu$ m) were captured with no binning. The acquisition was controlled with AcquireSR Acquisition control software. Images from different color channels were registered in SoftWoRx using parameters generated from a gold grid registration slide (GE Healthcare).

## Image analysis

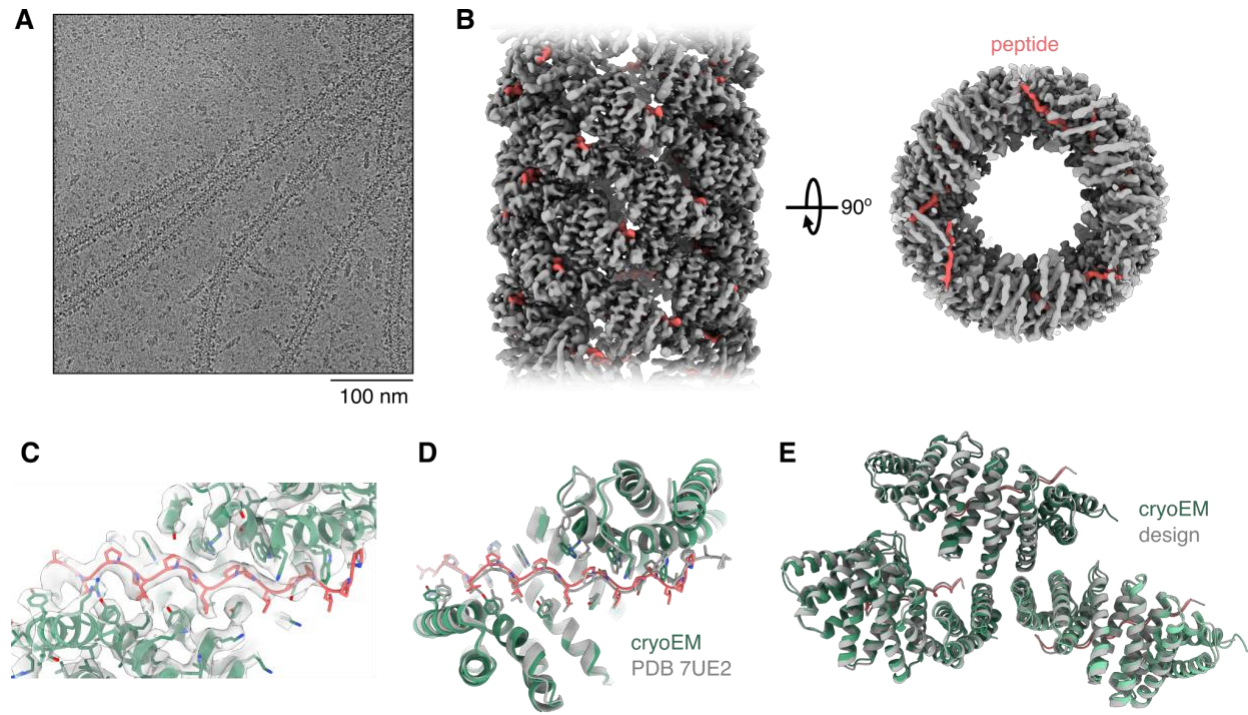
To segment the fiber images, a subset of 35 images were set aside for training a random forest classifier in a custom Jupyter notebook. The training images were manually annotated to define true positive fiber pixels in each image. The manual annotations were then used to train a classifier using the RandomForrestClassifier module from the Scikit-learn library (19). Our model defined the max tree depth as 10 and the number of estimators to be 50. Accuracy scores  $> 0.99$  were measured in our training set and the results of this custom model were qualitatively similar to a previously generated model using the Weka trainable segmentation module in Fiji (12, 20). Following model training a second analysis notebook was used to segment each image. After model-generated segmentation, we processed the output by removing small objects (less than 64 pixels) and performed a binary closing operation to prevent over-segmentation. The resulting objects were measured for size and intensity.

## Directional growth experiment

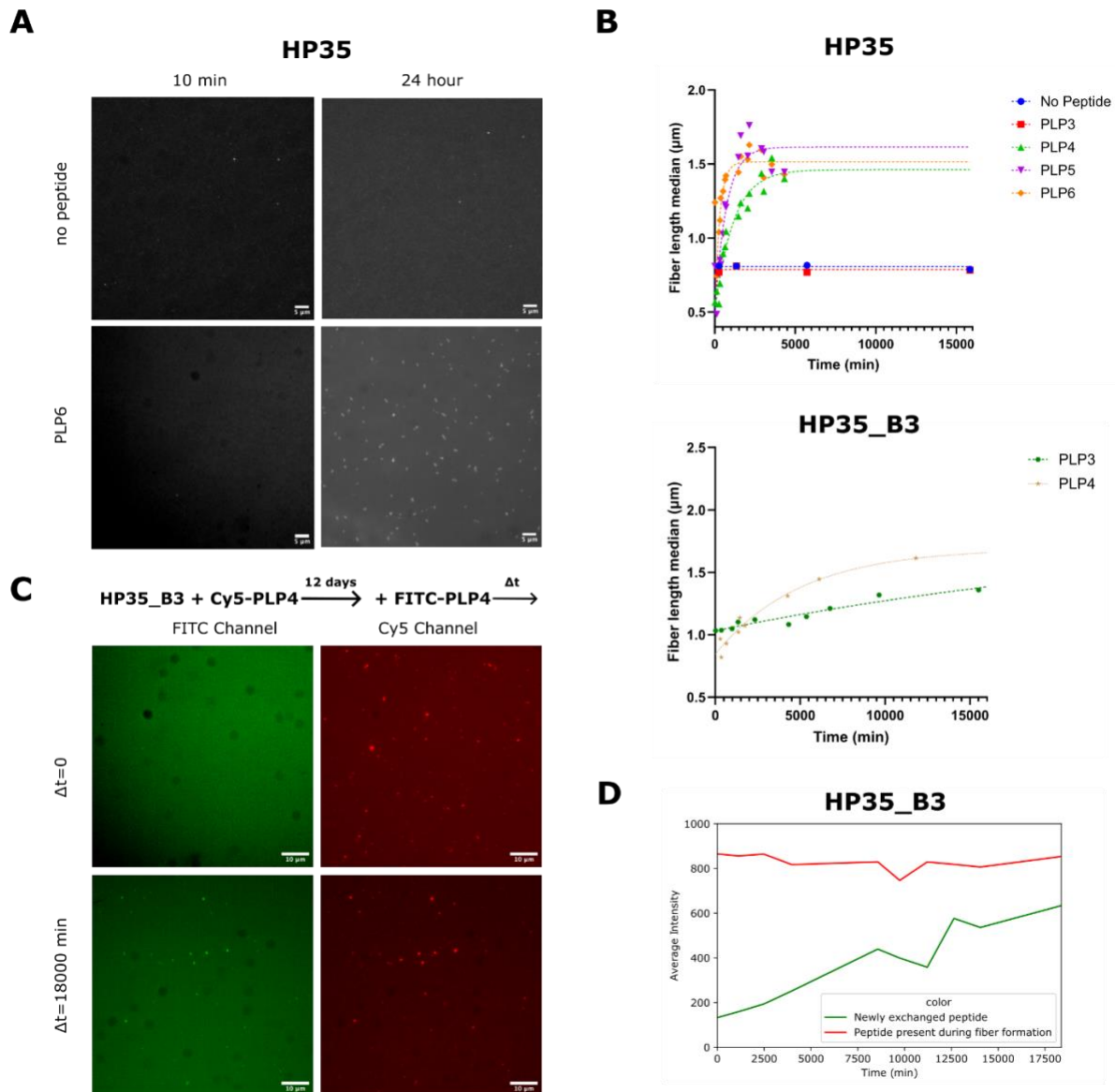
HP35-GFP (30 $\mu$ M) and PLP6-GFP (30 $\mu$ M) were incubated at room temperature for 48 hours in 25mM Tris, 75mM NaCl, 1mM TCEP. After 48 hours, HP35-cys (30 $\mu$ M) and PLP6-mKO2 (30 $\mu$ M) were separately mixed in 25mM Tris, 75mM NaCl, 1mM TCEP. The pre-assembled HP35-GFP/PLP6-GFP fibres were added at a 1:5 v/v ratio. This mixture was incubated for 1 hour, before 1:20 dilution with 25mM Tris, 75mM NaCl, 1mM TCEP, 0.4% methlycellulose. 20 $\mu$ l was added to an Ibidi flow cell mounted on clean room-grade coverslips (custom, 25  $\times$  75 mm<sup>2</sup>, Nexterion), and passivated with PLL-PEG (0.1 mg ml<sup>-1</sup> in 20 mM Hepes, pH 7.6; 5 min). Fibers were allowed to deposit on the glass for 15 minutes, before washing with and TIRF imaging in 25mM Tris, 75mM NaCl, 1mM TCEP, 0.4% methlycellulose.



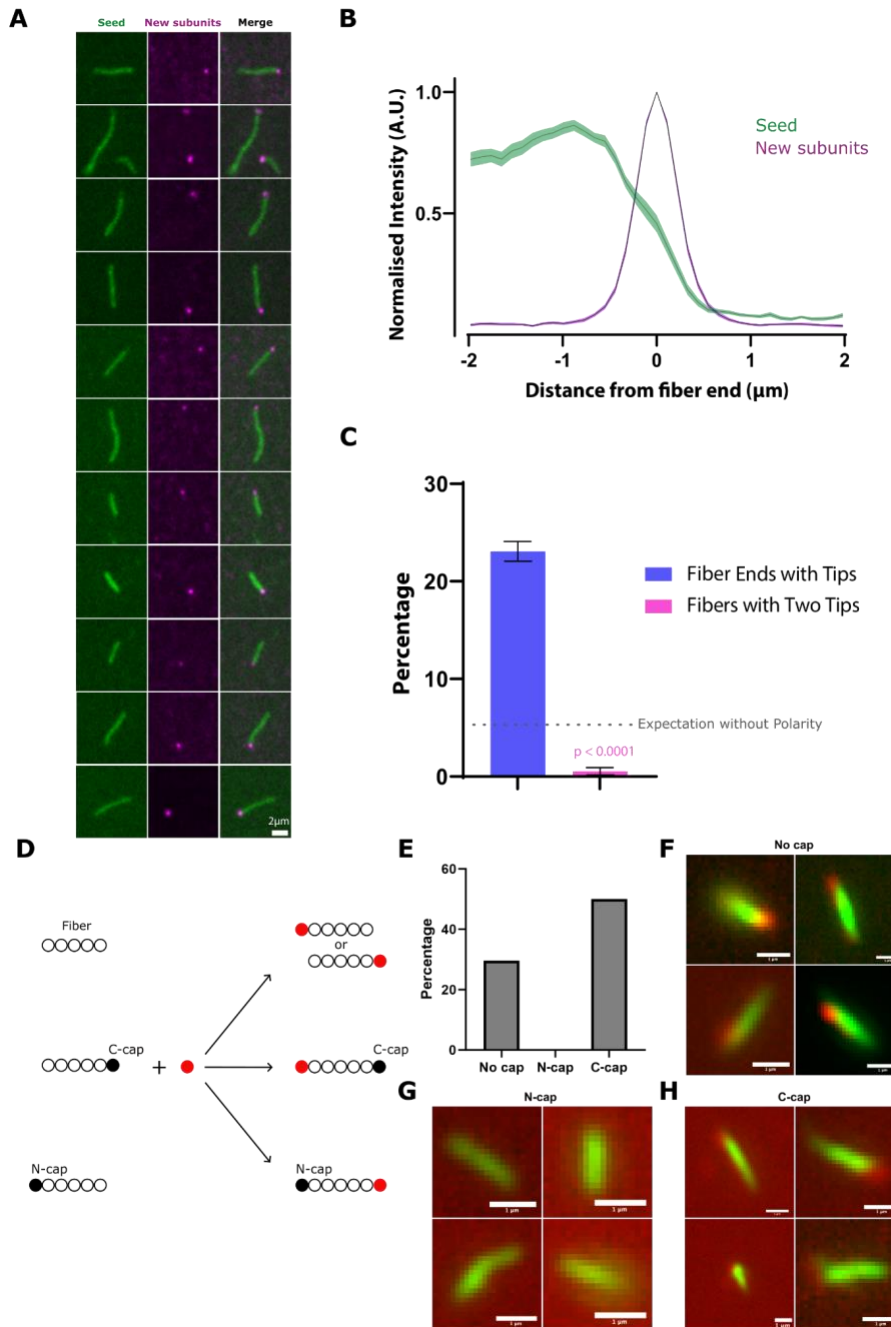
**Fig. 1 Design strategy for generating peptide-dependent protein filaments.** (A) Designed peptide binding monomeric starting point (4) for fiber design. (B) Computational sampling of alternative helical packing arrangements of monomer shown in A, with each monomer in a different color. The grey arrows represent the rigid body transform that generates the helical assembly. (C) Compact packing arrangement of fibers subunits in lowest energy assemblies following sequence design. Monomers are colored as in B.



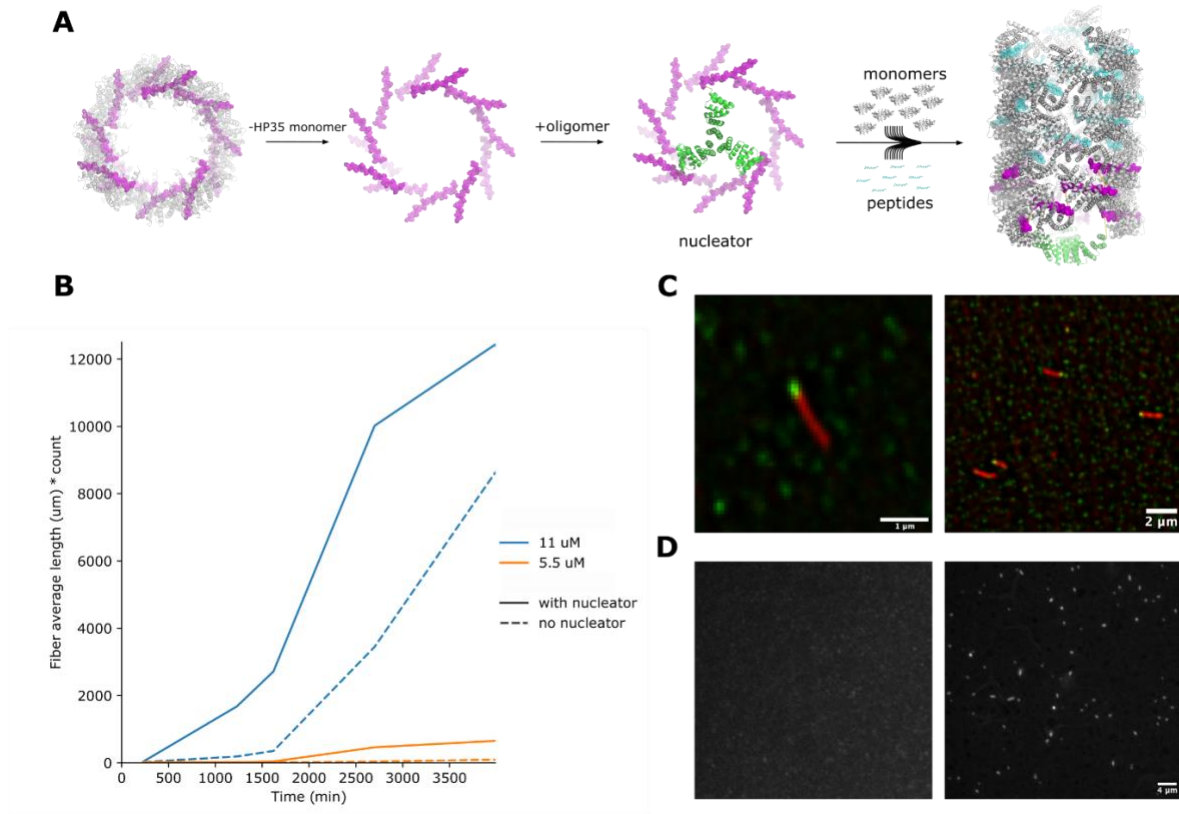
**Fig. 2 CryoEM structure of HP35.** (A) Representative cryoEM micrograph of HP35 filaments. (B) CryoEM map of HP35. Peptides are colored pink. (C) Peptide binding site in the HP35 cryoEM structure. (D) Comparison of a monomer from the HP35 filament structure (green and pink) with the original scaffold RPB\_PLP3\_R6 (PDB 7UE2) (grey). (E) Comparison of the HP35 computational design (grey) and cryoEM structure (green and pink) over three monomers comprising the filament interfaces.



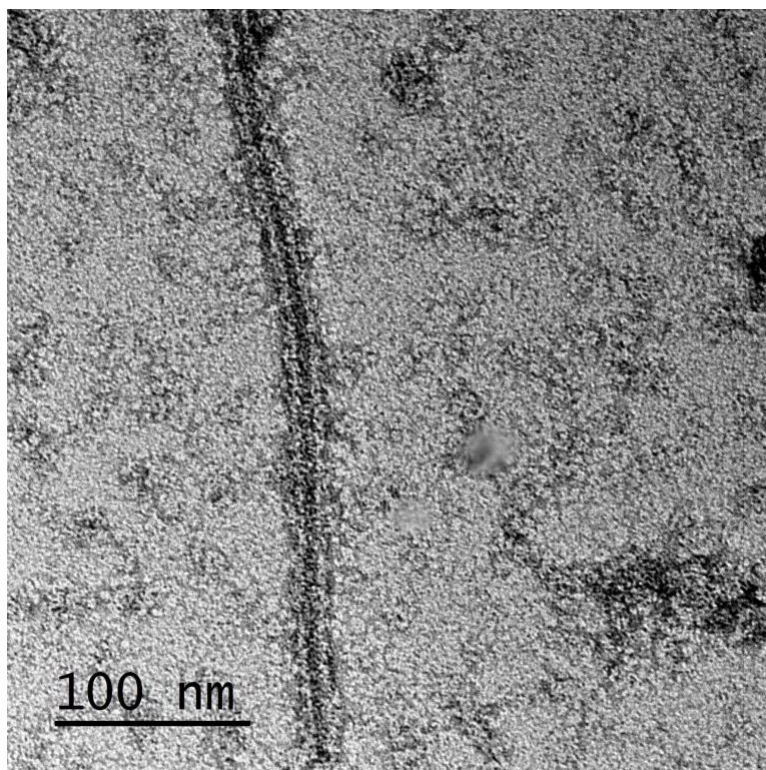
**Fig. 3 Filament growth dynamics.** (A) Representative fluorescent microscopy images of HP35 mixed with different lengths of peptide (no peptide vs PLP6) over time and (B) quantification of median fiber lengths over time and with the nonlinear fit for HP35 no peptide vs PLP3-6 GFP (top) and HP35\_B3 with PLP3-4 GFP (bottom). (C) Representative images of HP35\_B3 fiber formed by mixing with PLP4<sub>cy5</sub> (top panels) and then PLP4<sub>FITC</sub> was added (bottom panels). (D) Quantification of PLP4<sub>FITC</sub> (green) intensity over time compared to PLP4<sub>cy5</sub> (farred).



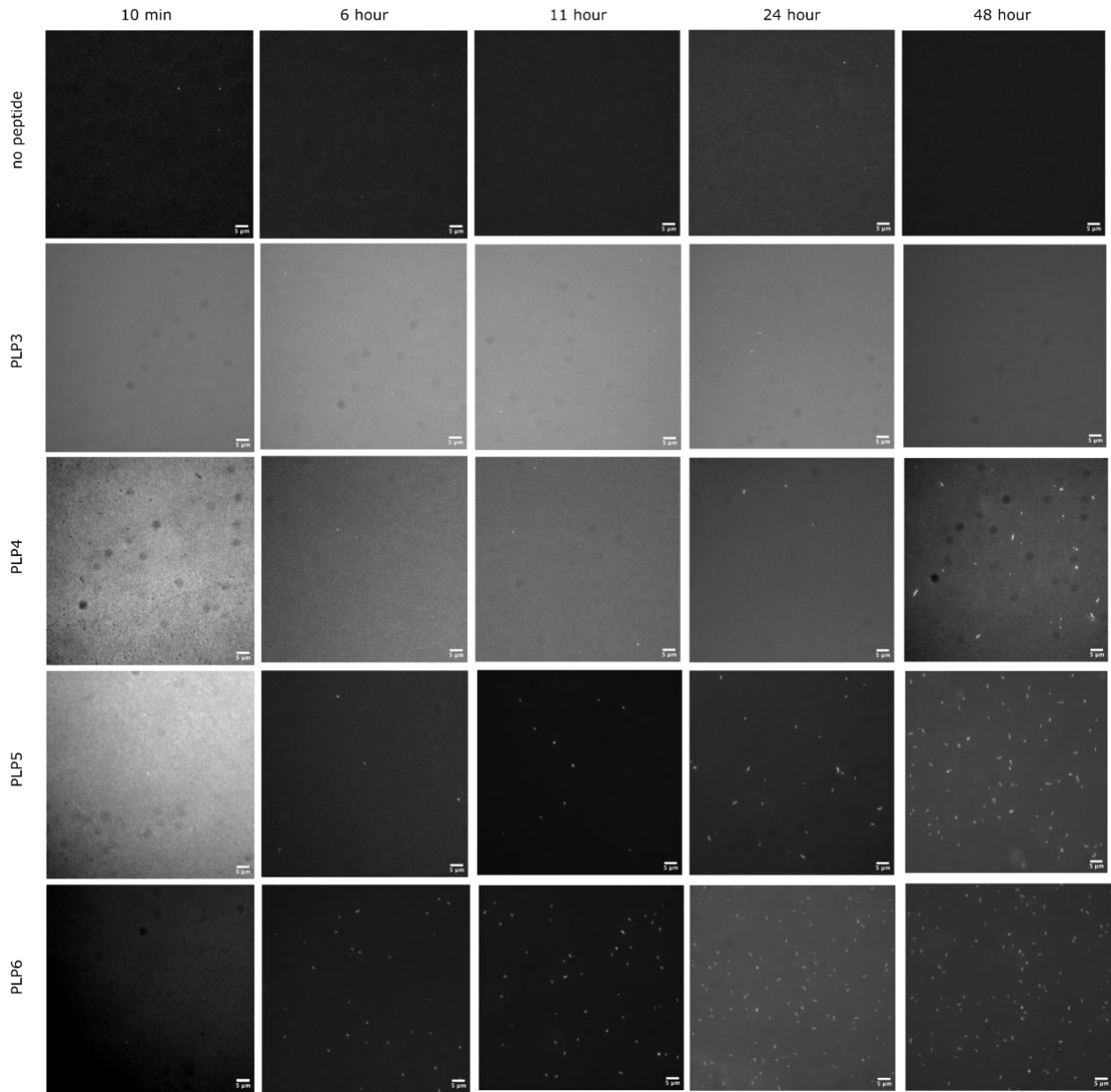
**Fig. 4 Polarized growth of filaments.** (A) Fibers were grown using GFP-labeled subunits (HP35 GFP + PLP6-GFP). Additional HP35 subunits and PLP6 peptides labeled with red fluorescent protein mKO2 were then added. The TIRFM images show that the red additions are almost always on just one end of the green fiber seeds. (B) Intensity analysis of fiber growth at the end.  $n=42$  fibers. (C) Statistical analysis of fiber ends with tips.  $N = 52$  fields of view (FOV),  $n = 8-21$  fibers per FOV. (D) Schematic of control experiment of green fiber seed and red new fiber with caps (top) and the design of caps (bottom). (E). The percentage of red tip growth quantified from fluorescent images of fiber growth (F) without caps and (G) with N-cap or (H) with C-cap. Scales bars are one micrometer long in F-H



**Fig. 5 Filament growth with designed nucleator.** (A) Construction of nucleator with designed oligomers displaying multiple peptides locally. Flexible GS linkers are shown in yellow dashed lines. (B) Quantification of fiber growth over time at different concentrations with (solid line) and without (dash line) nucleator from fluorescent microscope images. (C) TIRFM images of fibers (red) grown from nucleators (green). (D) HP35 mixed with nucleator without (left) and with (right) PLP6 peptide.

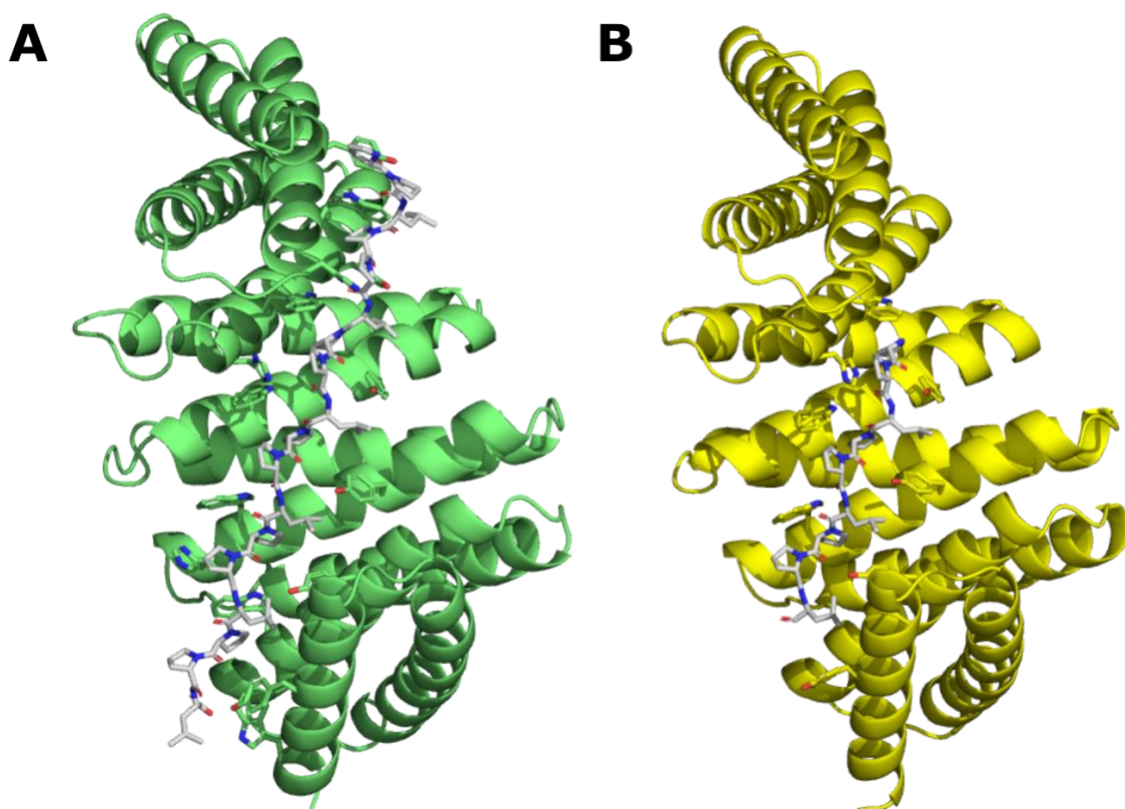


**Fig S1. Negative stain EM image of HP35 fiber**

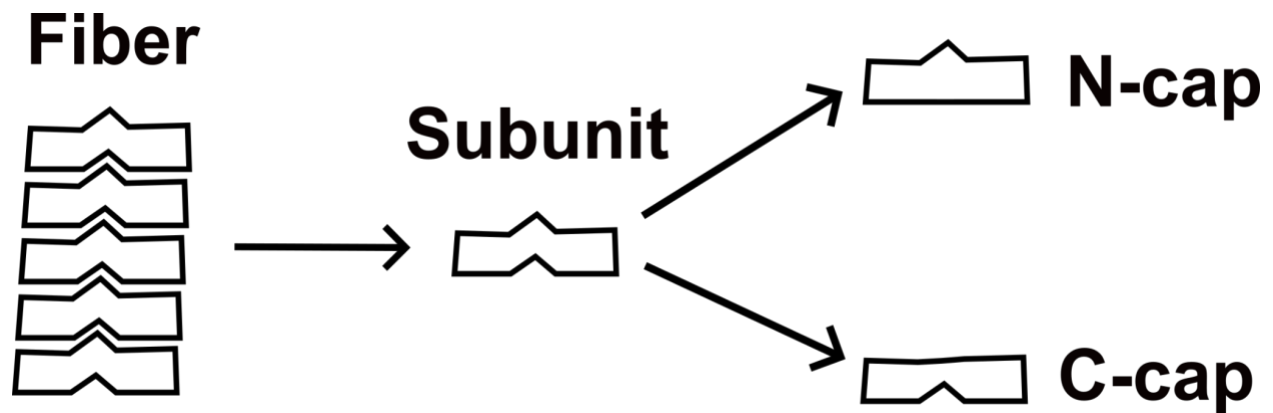


**Fig S2. Filament growth dynamic.** Representative fluorescent microscopy images of HP35 mixed with different lengths of peptide PLP3-5 over time

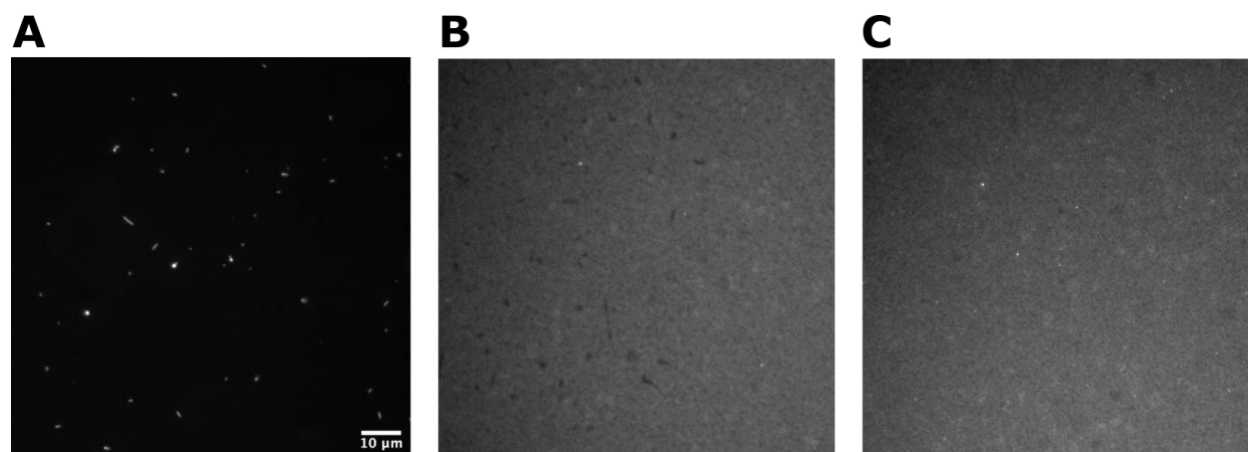




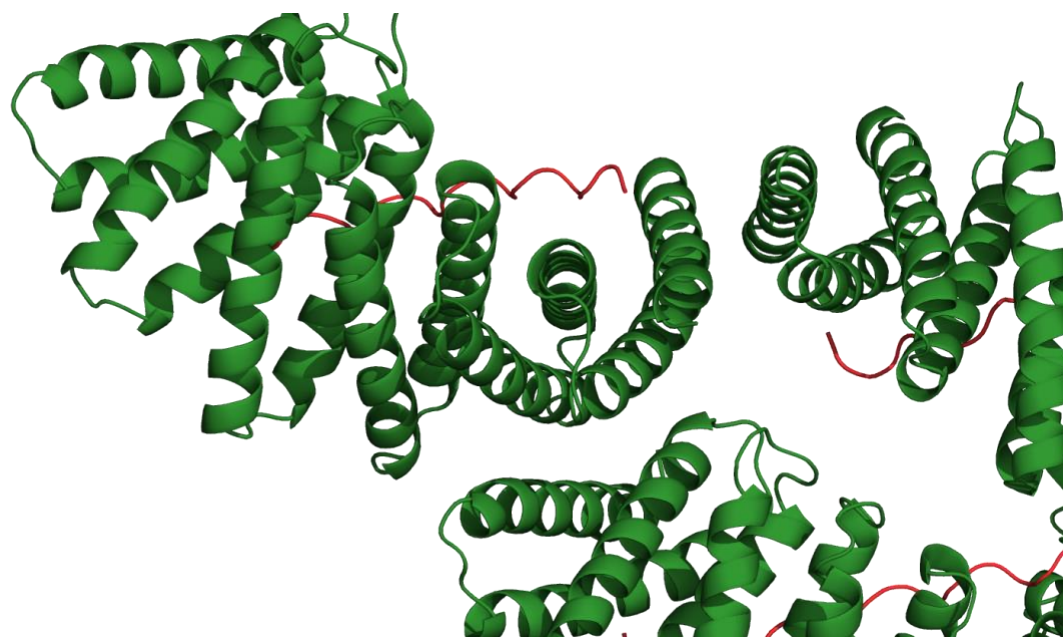
**Fig S3. HP35\_B3 design model. (A) Design model of HP35 and (B) Design model of HP35\_B3**



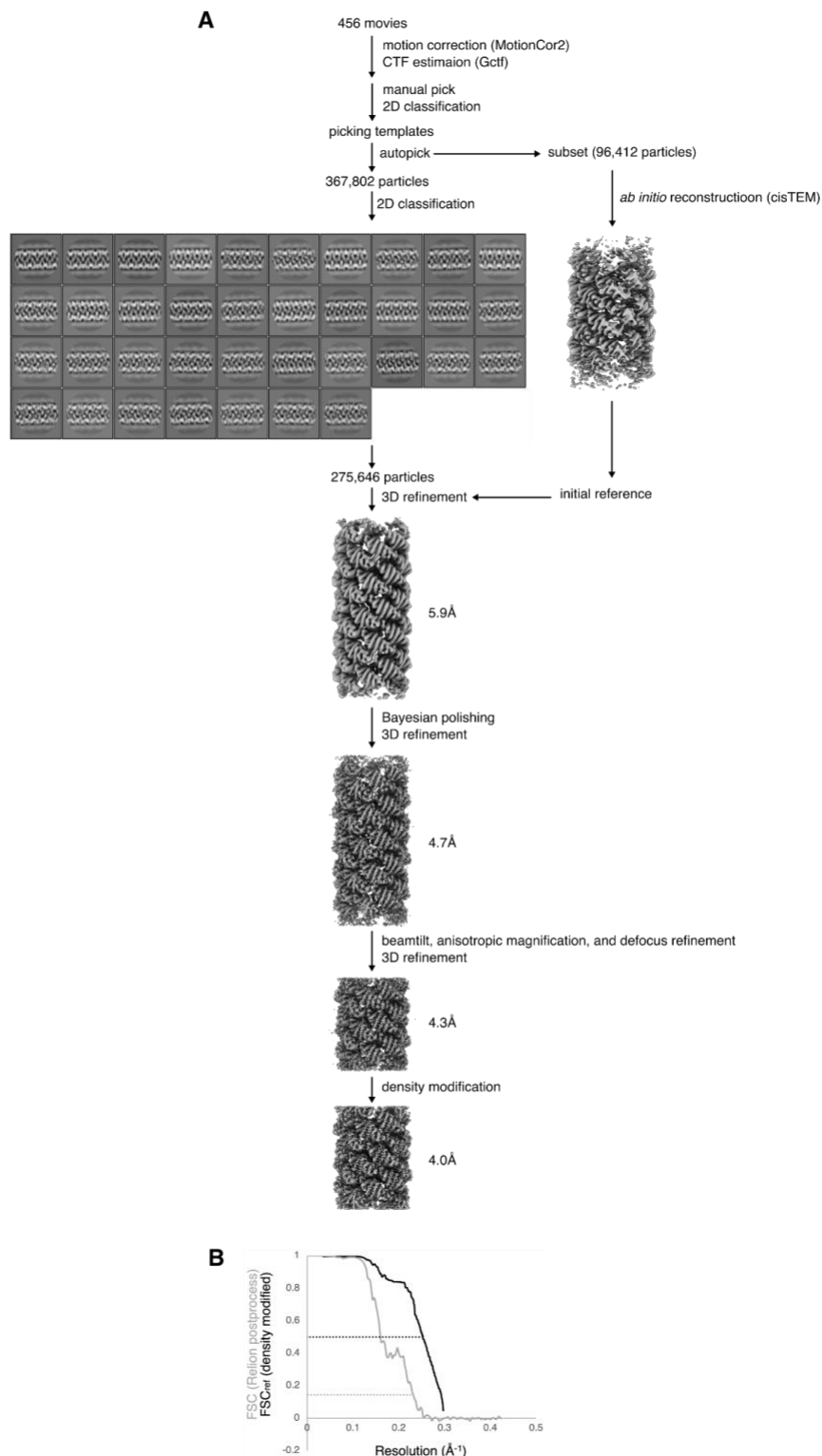
**Fig S4. Fiber cap design.** Design of N-cap and C-cap by knocking out different halves of the design interfaces



**Fig. S5 Fiber growth in the presence of caps.** HP35 mixed with PLP6 peptide in the absence of caps (A) and the presence of N-cap (B) or C-cap (C).



**Fig. S6. Formation of two disconnected half interfaces with the addition of a subunit on the other end**



**Fig. S7 CryoEM data processing.** (A) Flowchart of cryoEM data processing. (B) Half-map FSC curve from Relion postprocessing (grey) and FSC<sub>ref</sub> curve after density modification (black) and corresponding resolution estimates.

**Table S1: Cryo-EM data collection, refinement and validation statistics**

Structure	HP35 filament
PDB code	8VIM
EMDB code	43266
Magnification	36000
Voltage (kV)	200
Electron fluence (e <sup>-</sup> /Å <sup>2</sup> )	65
Defocus range (μm)	0.6-2.2
Pixel size (data collection) (Å)	1.16
Pixel size (reconstruction) (Å)	1.16
Point group symmetry	C1
Helical rise (Å)	6.2
Helical rotation (degrees)	132
Particle images (no.)	275,646
Resolution (0.143 FSC) (Å)	4.3
Resolution, density modified (0.5 FSC <sub>ref</sub> ) (Å)	4
R.m.s. deviation bond lengths (Å)	0.42
R.m.s. deviation bond angles (°)	0.66
MolProbity score	1.16
Clashscore	2
C-beta deviations	0
Rotamer outliers (%)	0
Ramachandran favored (%)	97
Ramachandran allowed (%)	2
Ramachandran outliers (%)	0

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