

1 **Resolving Filament Level Mechanics in Collagen Networks using Activity**
2 **Microscopy**

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12
13 **Abstract**

14 Collagen is the most abundant protein in humans and the primary component of the
15 extracellular matrix, a meshwork of biopolymer networks, which provides structure and
16 integrity to tissues. Its mechanical properties profoundly influence the fate of cells. The
17 cell-matrix interaction, however, is not well understood due to a lack of experimental
18 techniques to study the mechanical interplay between cells and their local environment.
19 Here we introduce Activity Microscopy, a new way to visualize local network mechanics
20 with single filament resolution. Using collagen I networks *in vitro*, we localize fibril
21 positions in two-dimensional slices through the network with nanometer precision and
22 quantify the fibrils' transverse thermal fluctuations with megahertz bandwidth. Using a
23 fibril's thermal fluctuations as an indicator for its tension, we find a heterogeneous stress
24 distribution, where “cold” fibrils with small thermal fluctuations surround regions of highly
25 fluctuating “hot” fibrils. We seed HeLa cells into collagen networks and quantify the
26 anisotropy in the propagation of their forces.

27

28 **Introduction**

29 Filamentous biopolymer networks fulfill a plethora of mechanical functions both inside
30 and outside of cells. Intracellular networks impart motility and mechanical strength to
31 cells¹, while the networks of the extracellular matrix (ECM) provide integrity to tissues
32 and entire organs². A particular biopolymer network can fulfill many distinct functions,
33 often achieved by the same basic building blocks arranged in different architectures. For
34 example, collagen, the most prominent component of the extracellular matrix, can be
35 arranged into networks of fibers with wildly diverse densities and connectedness, from the
36 loose and soft elastic networks that form the interstitial matrix of the skin to densely packed
37 mineralized fibers in bones³. Many physiological processes are regulated by the stiffness
38 of the ECM, such as cellular migration, differentiation or proliferation⁴⁻⁶. The stiffness also
39 plays a role in the progression of skin cancer, as cancerous cells remodel the collagen
40 network of the extracellular matrix to reach blood vessels⁷.

41

42 Filamentous biopolymer networks are used by engineers as scaffolds to build artificial
43 tissues that mimic true physiological mechanical properties. However, such approaches
44 remain challenging without a better understanding of the complex interplay between
45 individual filament properties, network architecture and mechanical function^{8,9}. Since the
46 typical pore size of these networks is on the same order as the size of cells embedded in
47 the network, cells interact mainly with the individual filaments that surround them, rather
48 than with the global network. To migrate through the extracellular matrix, for example,
49 cells must either squeeze through pores in the network or remodel fibers. Systematic
50 progress in understanding such local cell-matrix interaction is hindered by a lack of
51 techniques that can simultaneously resolve the local network architecture and the
52 interaction forces between cells and filaments.

53

54 Currently, the three-dimensional architecture of networks and their interaction with cells
55 can be resolved by confocal microscopy, either in fluorescence or reflection mode, or by
56 light sheet microscopy¹⁰. These imaging modes do not, however, directly measure the
57 mechanics of the interaction. Forces generated by cells can be indirectly quantified from
58 network deformation data by traction-force microscopy (TFM), provided that the

59 mechanical properties of the filaments and their connectedness are known. Recently,
60 Steinwachs *et al.*¹¹ computed for the first time¹² the forces that breast carcinoma cells apply
61 to biopolymer networks designed to mimic the cells' physiological surroundings. Forces
62 were calculated from network deformations around the cell using a finite-element approach
63 based on a constitutive equation that captures the complexity of the surrounding network.
64 A drawback of this method is that the measurement of network deformations requires
65 knowledge of the force free state, which is achieved by disassembling the force generating
66 filamentous actin network within the cells. The method then assumes that all deformations
67 are elastic, which neglects plastic deformations that have been observed to occur in the
68 ECM^{13,14}. There currently exists no method capable of continuously quantifying cell-
69 matrix interactions with single filament resolution and without the assumption of elastic
70 deformations.

71

72 Here we introduce Activity Microscopy, a method for measuring the precise location and
73 the lateral thermal fluctuations of filaments in fibrous biopolymer networks with
74 subnanometer precision and megahertz bandwidth. The magnitude of the lateral
75 fluctuations is a function of filament tension, bending stiffness, network architecture, and,
76 possibly, fluctuating active forces. Therefore, lateral fluctuations are a direct measure for
77 cell-matrix interactions. For an *in vitro* collagen I network, we localize fibrils with
78 nanometer precision and measure their fluctuations along their contour. We find that
79 tension bearing fibrils with small fluctuations surround pockets of weakly loaded fibrils
80 with larger fluctuations. To demonstrate the sensitivity of Activity Microscopy to changes
81 in fibril stiffness, we observe the reduction of fibril fluctuations after cross-linking with
82 glutaraldehyde. Finally, we seed collagen matrices with HeLa cells and measure the
83 reduction in magnitude and the increase in asymmetry of fluctuations near cells.

84 Results

85

86 Principle of Activity Microscopy

87 Activity microscopy aims to visualize the contribution of every filament to a network's
88 macroscopic mechanical response while simultaneously providing precise information
89 about the network architecture. The filament bending stiffness, stretching and compression
90 forces, and network connectivity affect transverse filament fluctuations, which can either
91 be thermally driven or caused by active elements in the network. We recently demonstrated
92 that thermally driven filament fluctuations can be quantified even in “athermal” collagen I
93 networks¹⁵ for which transverse fluctuations are not expected to contribute to the
94 mechanical response¹⁶. Here, we develop this method further and introduce two-
95 dimensional Activity Microscopy. We image areas that include many pore diameters and
96 thus provide insight into how individual filaments contribute to the network's overall
97 mechanical response.

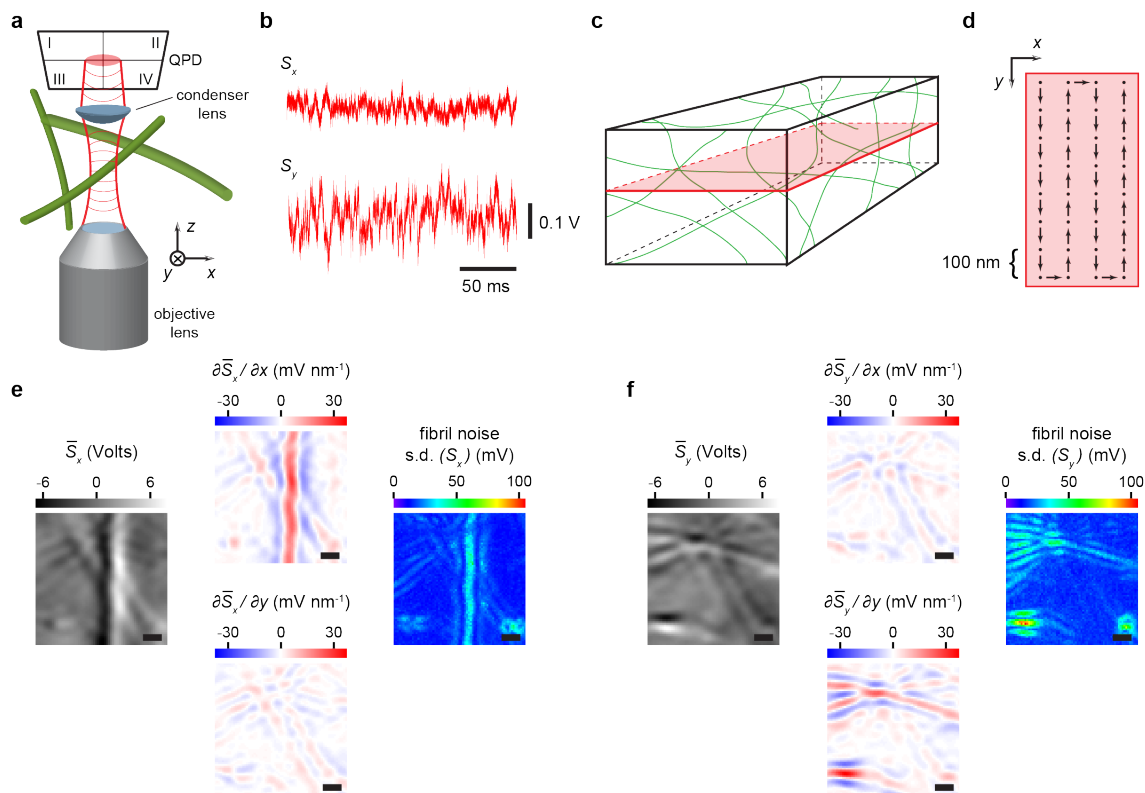
98 For Activity Microscopy imaging, a near infrared laser beam is focused into a sample
99 chamber filled with a filament network. The forward scattered and unscattered laser light
100 is then collected and guided onto a quadrant photo diode (QPD) (**Fig. 1a**). This detection
101 scheme is commonly used in optical trapping experiments^{17,18}. The QPD outputs two
102 signals, which are calculated from the voltage differences between the quadrants, $S_x =$
103 $(S_I + S_{III}) - (S_{II} + S_{IV})$ and $S_y = (S_I + S_{II}) - (S_{III} + S_{IV})$ (**Fig. 1a,b**). In optical
104 trapping experiments, S_x and S_y can be directly related to the x - and y -position of a
105 trapped particle. However, in the case of filaments, the signals are not independent but
106 depend on the orientation of the filament relative to the QPD (**Supplementary Fig. 1**).

107 For two-dimensional imaging, a plane of interest is selected (**Fig. 1c**) and raster scanned
108 by translating the sample in steps (**Fig. 1d**), which are chosen to be smaller than the
109 diffraction limited spot diameter. **Figure 1e** and **f** show the average detector signals, the
110 calculated detector sensitivities and the raw noise signal for the example of an *in vitro*
111 collagen I network. A total area of $7 \mu\text{m} \times 7 \mu\text{m}$ was scanned with a step width of
112 100 nm . At each point, a time series of 100 ms was recorded at 100 kS s^{-1} . The left panel
113 in **Figure 1e** shows the averaged detector signal \bar{S}_x along the x -axis from which the
114 detector sensitivities $\partial\bar{S}_x/\partial x$ and $\partial\bar{S}_x/\partial y$ (mid panels) are calculated. Clearly, S_x is

115 mainly sensitive to fibrils that are oriented vertically. The highest sensitivity is measured
 116 when the laser beam is centered on the fibril (broad vertical line). This is expected from
 117 the detector response to a fibril, which has its maximal slope at the midpoint of the fibril
 118 (**Supplementary Fig. 1**). Thus, the location of maximal sensitivity can in turn be used to
 119 determine the precise position of the fibril axis. The fibril's transverse position fluctuations
 120 can be calculated from the standard deviation s.d. of the times series of S_x and S_y . Since
 121 the highest detector sensitivity leads to the largest raw position noise for a given filament
 122 fluctuation amplitude, the maxima in the raw noise image (right panel) also indicate the
 123 positions along the fibril axis.

124

125 **Figure 1**



Activity Microscopy. (a) A near infrared laser beam is focused into the sample filled with a collagen I *in vitro* network (green: fibrils). The forward scattered light from a fibril is collected by a condenser lens and guided onto a quadrant-photo diode (QPD) where it interferes with the unscattered portion of the laser beam. A time series of lateral voltage signals $S_x = (S_I + S_{III}) - (S_{II} + S_{IV})$ and $S_y = (S_I + S_{II}) - (S_{III} + S_{IV})$, recorded at one position are shown in b. (c, d) For imaging, a plane of interest in the sample is selected and

scanned line by line while recording time series at each position (black dots). The mean detector signal \bar{S}_x is shown in the left panel of **e**. To find the fibrils' axes, the spatial derivatives $\partial\bar{S}_x/\partial x$ and $\partial\bar{S}_x/\partial y$ are calculated, termed detector sensitivity (middle panel). The raw transverse fibril noise is calculated from the time series of S_x and displayed as the standard deviation (s.d.) (right panel). **f** shows the averaged detector signal, derivatives and transverse fibril noise for the y -signal, which has its maximal sensitivity for fibrils oriented along the horizontal axis. Scale bars in **e** and **f** are $1\ \mu\text{m}$. The collagen network was polymerized according to protocol I (see Methods). Distance between scanning lines and dots in **d** is $100\ \text{nm}$. Time series at each location in **e** and **f** is $100\ \text{ms}$ long.

126

127 The fibril noise images in **Figure 1e** and **f** show “ghost filaments” that run next to the main
128 fibril axis. They result from the characteristic detector response of a fibril. Besides the
129 maximal detector sensitivity when the fibril is directly in focus, the detector also has
130 significant sensitivities to the left and right of the center (**Supplementary Fig. 1c**, dashed
131 lines). These regions of the detector response with negative slope are clearly visible in the
132 detector sensitivity panels as blue lines, i.e. negative sensitivities, that follow the fibril on
133 either side. Since negative sensitivities can be easily identified, this detection artifact can
134 be corrected for by removing signals originating from locations in the sample with negative
135 sensitivities (**Supplementary Fig. 2**).

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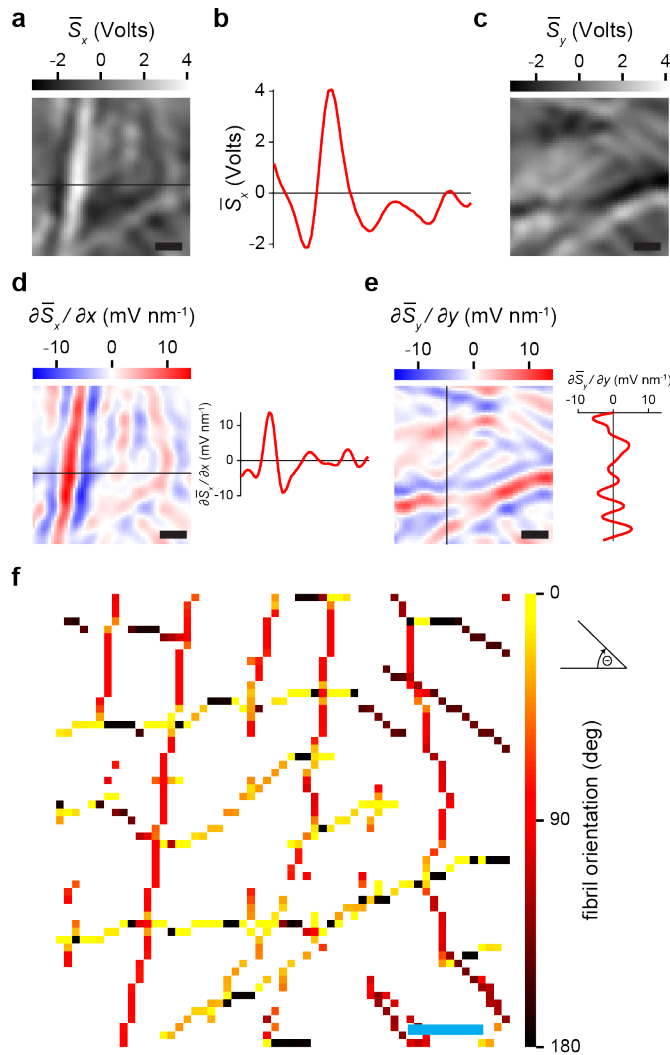
137 **Finding Filament Axes and Orientations**

138 Point-by-point Activity Microscopy imaging is time consuming because at each point in
139 the region of interest a time series must be recorded for a sufficiently long time to
140 characterize the local fluctuations of the filaments. Given an approximate pore size of
141 $2 - 3\ \mu\text{m}$ for the collagen concentration used here ($2.4\ \text{mg}\ \text{ml}^{-1}$) (ref. 19), most of the
142 scanning time is spent inside pores. Additionally, fibril fluctuation amplitudes can only
143 change along the contour of a filament, but not transversally to the contour. Thus, data
144 from the same position along the contour of the filament, but at different positions
145 transversal to its axis, yield the same fluctuation amplitude as long as the filament remains
146 within the linear range of the detector (**Supplementary Fig. 3**). For characterizing the
147 fluctuation of each fibril in the network, it is therefore sufficient to measure fibril
148 fluctuations only at points along the fibril axis. To implement such an imaging algorithm,

149 the locations and orientations of all fibrils in a network must be known. We achieve this
 150 by rapidly pre-scanning the network (“fast scan”), only recording time series sufficiently
 151 long to determine the average detector signals \bar{S}_x and \bar{S}_y .

152

153 **Figure 2**



Localization of fibrils. Fibrils are located using the maximum sensitivity values either along the x - or y - direction depending on their orientation relative to the detector coordinate axes. A line profile of the mean signal \bar{S}_x (black line in **a**) reveals the detector response to an individual fibril (**b**). (**c**) The mean signal \bar{S}_y from the same scan as in **a**. The detector sensitivities $\partial\bar{S}_x/\partial x$ (**d**) and $\partial\bar{S}_y/\partial y$ (**e**) show maxima at the fibrils' locations (see also representative horizontal and vertical line profiles, respectively). $\partial\bar{S}_x/\partial x$ is used to identify fibrils oriented with an angle $\leq 45^\circ$ w.r.t. the vertical direction, while $\partial\bar{S}_y/\partial y$ serves to find fibrils oriented with an angle $< 45^\circ$ w.r.t. the horizontal direction. The locations of the fibrils are plotted in **f**. The color indicates the fibril orientation w.r.t. the horizontal direction (see Methods). Scale bars are $1 \mu m$. The collagen network was polymerized according to protocol I (see Methods).

154

155 **Figure 2a** and **c** show the results for a time series with a length of $5 ms$ (**Supplementary**
 156 **Note 1**). There is no obvious degradation in the \bar{S}_x and \bar{S}_y images in comparison to the
 157 images shown in **Figure 1e** and **f**. The line profiles of the average detector signals are still

158 smooth (**Fig. 2b**) and the calculated detector sensitivities are essentially noise free (**Fig.**
159 **2d,e**). Since the positions of maximal detector sensitivities colocalize with the fibrils'
160 locations, we can use these images to localize the fibrils. To achieve this, horizontal and
161 vertical line profiles of the \bar{S}_x and \bar{S}_y signals, respectively, are extracted and the locations
162 of the maxima determined. In this way, the fibrils' axes are found with nanometer precision
163 (**Supplementary Note 1**). **Figure 2f** shows the locations of the fibrils identified using this
164 algorithm on a grid with a pixel size of $100\text{ nm} \times 100\text{ nm}$. Ghost filaments are excluded
165 because they correspond to minima in the detector sensitivity. The orientation of filaments
166 relative to the detector orientation is then determined (see Methods) and displayed color-
167 coded for each filament position. This algorithm reliably finds fibril axes locations and
168 orientation in a $5\text{ }\mu\text{m} \times 5\text{ }\mu\text{m}$ image in approximately two minutes.

169

170 **Long-range imaging**

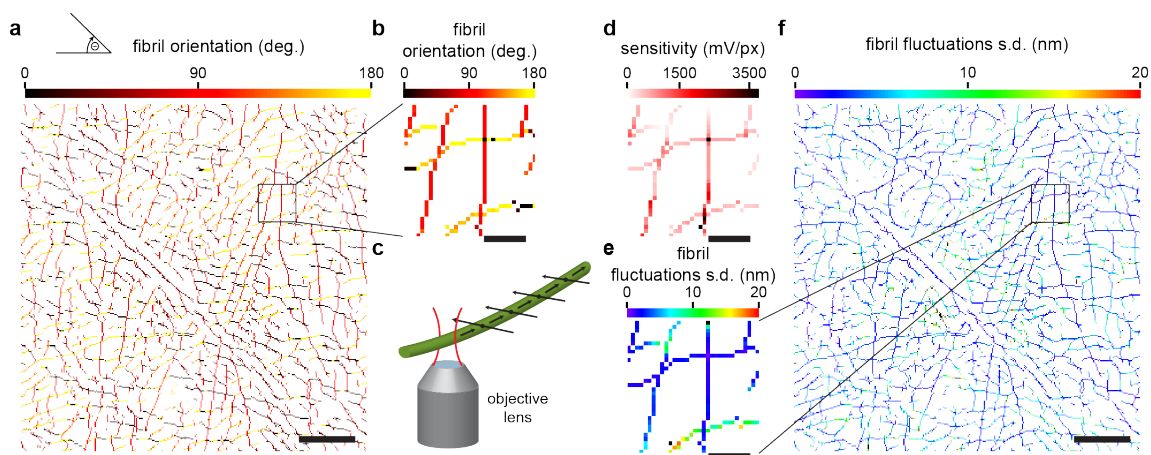
171 For visualizing how individual filaments contribute to the overall mechanical response of
172 a network, it is necessary to image areas that are large relative to the pore diameter. This is
173 challenging because slow drift of the sample or instrument will lead to increasing
174 discrepancies between the fibril's positions found by fast scanning, and the locations at
175 which the fluctuations are recorded. To solve this problem, we subdivide the area of interest
176 into smaller tiles of $5\text{ }\mu\text{m} \times 5\text{ }\mu\text{m}$ and perform Activity Microscopy imaging tile by tile,
177 thus avoiding significant relative drifts. As shown in **Figure 3a**, fibrils can be traced this
178 way over long distances without discontinuities. In each tile, a sequence of four steps is
179 performed. The fast scan with dwell time of 5 ms per pixel is recorded and the fibrils' axes
180 and their orientations are determined (**Fig. 3b**). The list of fibril positions is then used to
181 record fluctuation data for 200 ms only directly on the fibrils' axes. To obtain an accurate
182 detector calibration at the time that fluctuation data are recorded, the detector sensitivity is
183 measured at each pixel of interest by scanning the fibril perpendicularly to its axis through
184 the laser beam (**Fig. 3c,d**). The scan is used to calibrate the raw voltage signals which then
185 provide an accurate measure for the transverse fibril fluctuation at a given position on the
186 filament axis (**Fig. 3e,f**). At every position we chose either the x - or y -channel, depending
187 on which one has the greater sensitivity. Both channels will pick up the same fluctuation
188 amplitude except for fibrils that are perfectly aligned with one of the detector axes. This

189 selection of channels minimizes the uncertainty in fluctuation measurements
190 (Supplementary Fig. 4).

191 **Figure 3f** shows a $30\ \mu\text{m} \times 30\ \mu\text{m}$ Activity Microscopy large scale image of a collagen
192 network, in which the amplitude of the transverse filament fluctuations is color-coded for
193 the range $0 - 20\ \text{nm}$. The distribution of fluctuation amplitudes in the network is
194 heterogeneous: a few fibrils with strongly suppressed fluctuations cross the entire network
195 and in-between are pockets of strongly fluctuating fibrils.

196

197 **Figure 3**



198

Quantitative imaging of fibril fluctuations. The map of fibril locations and orientations (a, b) are used to record fibril fluctuations only at selected points corresponding to fibrils' axes. In a first step, the detector signal is recalibrated by scanning the laser beam perpendicularly to a fibril axis (c). d shows the refined detector sensitivities for each fibril position. A 200 ms time series is then recorded, and the raw voltage signals are converted into calibrated transverse fibril fluctuations using the measured detector sensitivity. e and f show the final Activity Microscopy image. Each fibril position is shown color-coded with the magnitude of its transverse fluctuation. Scale bars are $5\ \mu\text{m}$ in a and f, $1\ \mu\text{m}$ in b, d, and e. The collagen network was polymerized according to protocol I (see Methods).

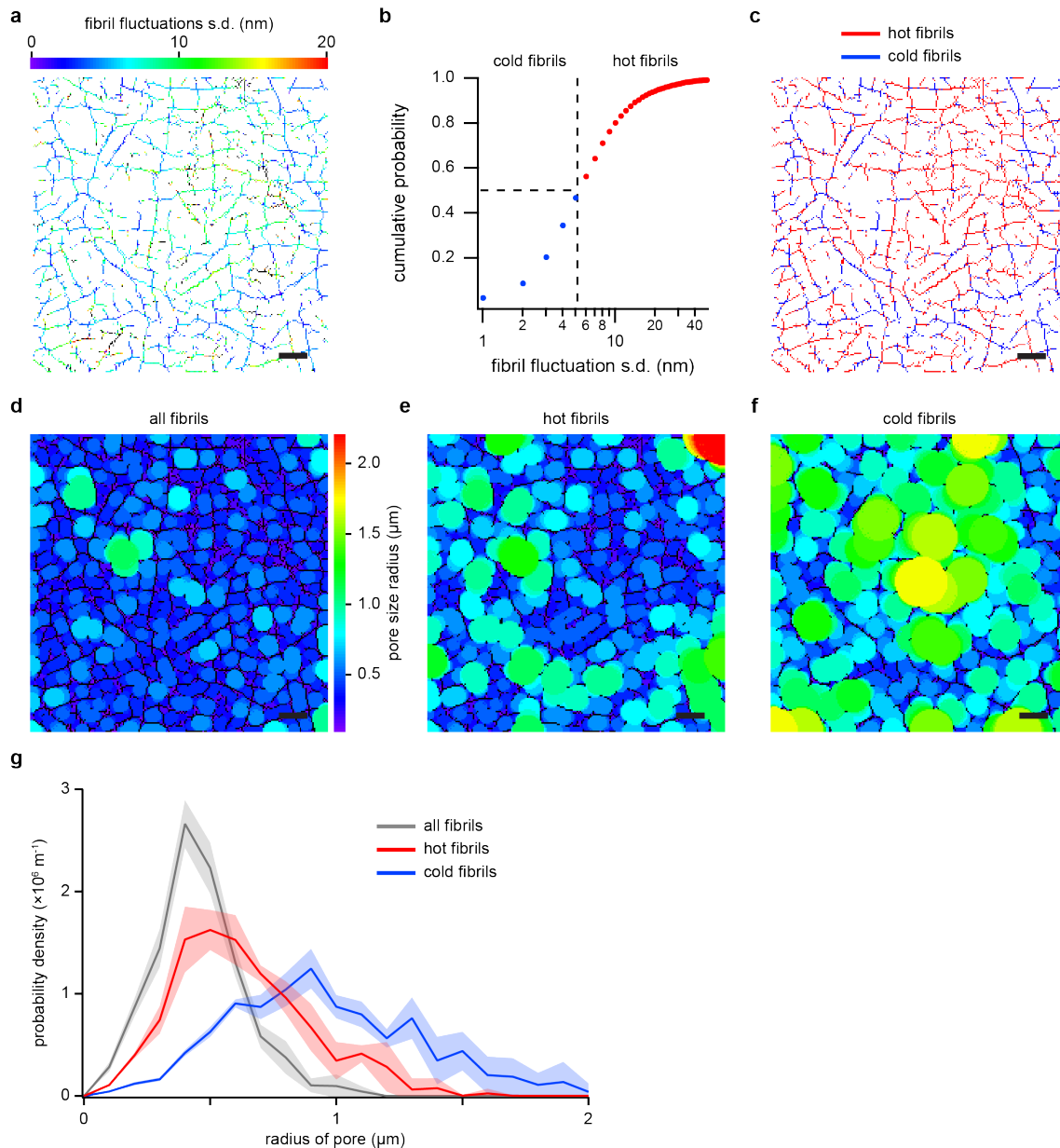
199

200 **Pore size distribution**

201 Activity Microscopy images of collagen networks suggest a heterogeneous distribution of
202 fibril fluctuations. A few fibrils with strongly suppressed fluctuations divide the field of
203 view, while the regions between them are filled with fibrils with large fluctuations. To

204 quantify this observation, we calculate the average pore size distribution for “hot” (strongly
205 fluctuating) and “cold” (weakly fluctuating) fibrils.
206 To calculate the pore size distribution, we adapted a robust method for calculating pore
207 sizes from confocal images of collagen networks developed by Mickel *et al.*¹⁹, which does
208 not rely on specific assumptions or network models. Fibrils are represented by medial axes
209 instead of their diffraction limited image. In our case, fibrils are represented by one-pixel
210 wide lines corresponding to a width of 100 nm. For each pixel of the fluid phase, the radius
211 of the largest disk that can be fitted into the pore while still including the pixel, is
212 determined. For pixels within the center of a pore, the maximal disk size is given by the
213 contact of the disk with the medial axes that form the pore. Pixels confined by fibrils in
214 corners will have a smaller radius. The pore size distribution for a network can be
215 summarized by a histogram of radii for all pixels of the fluid phase. The goal in our case is
216 to show that cold and hot filaments have different pore sizes. To achieve this, we divide all
217 fibrils (**Fig. 4a**) into an equal number of cold (blue) and hot (red) filaments based on the
218 cumulative probability of fibril fluctuations (**Fig. 4b**). An equal number of cold and hot
219 fibrils is achieved at a fluctuation threshold of ~ 5 nm. A binary image representing the
220 cold and hot fibrils shows again the large-scale pattern of cold filaments surrounding
221 pockets of hot filaments (**Fig. 4c**). To visualize the pore size distribution within an Activity
222 Microscopy image, we plot the maximal pore size, color-coded for every pixel in the fluid
223 phase either for all (**Fig. 4d**), only hot (**Fig. 4e**), or only cold fibrils (**Fig. 4f**). The overall
224 characteristics of the three cases is summarized in the probability density distribution of
225 pore radii (**Fig. 4g**). Except for a few pores, the pore radius for most pores is narrowly
226 distributed around 435 nm. Hot filaments alone have a slightly higher peak pore radius of
227 567 nm (**Supplementary Note 2**). In contrast, cold filaments have clearly a much larger
228 average pore radius indicated by the change in color and size of pores (**Fig. 4f**) and the
229 distribution of pore radii with a peak of 897 nm (**Fig. 4g**). This confirms quantitatively the
230 observation that cold filaments divide the network into pockets of hot filaments. Assuming
231 that cold filaments are a result of tension along their axes, one could interpret the data as a
232 few filaments carrying tension over long distances, while leaving pockets of relatively low-
233 tension fibrils between them. To our knowledge, such a heterogeneous stress distribution
234 has never been experimentally shown before.

235 **Figure 4**



Pore size distribution for strongly fluctuating (hot) versus weakly fluctuating (cold) fibrils. (a) Activity Microscopy image ($20 \mu\text{m} \times 20 \mu\text{m}$) of a network of native *in vitro* collagen I fibrils. The cumulative probability of fibril fluctuations for the image shown in a is used to divide fibrils equally into hot (red) and cold (blue) (b). The dashed lines indicate the median used to split the fibrils into hot and cold as also displayed in c. This image is used to calculate pore sizes between the network's fibrils (d) using a method described by Mickel *et al.*¹⁹ (see Methods). The subset of hot fibrils forms smaller pores than the subset of cold fibrils (e, f). (g) The difference in pore sizes of networks of hot and cold fibrils can

be quantified by the probability densities of their pore size radii. Cold fibrils form pores $\approx 1.6 \times$ the size of the pores formed by hot fibrils (shaded region represents s.d. of four independent Activity Microscopy images of two different samples). Scale bars are $2 \mu\text{m}$. Collagen was polymerized according to protocol II (see Methods).

236

237 **Collagen crosslinking**

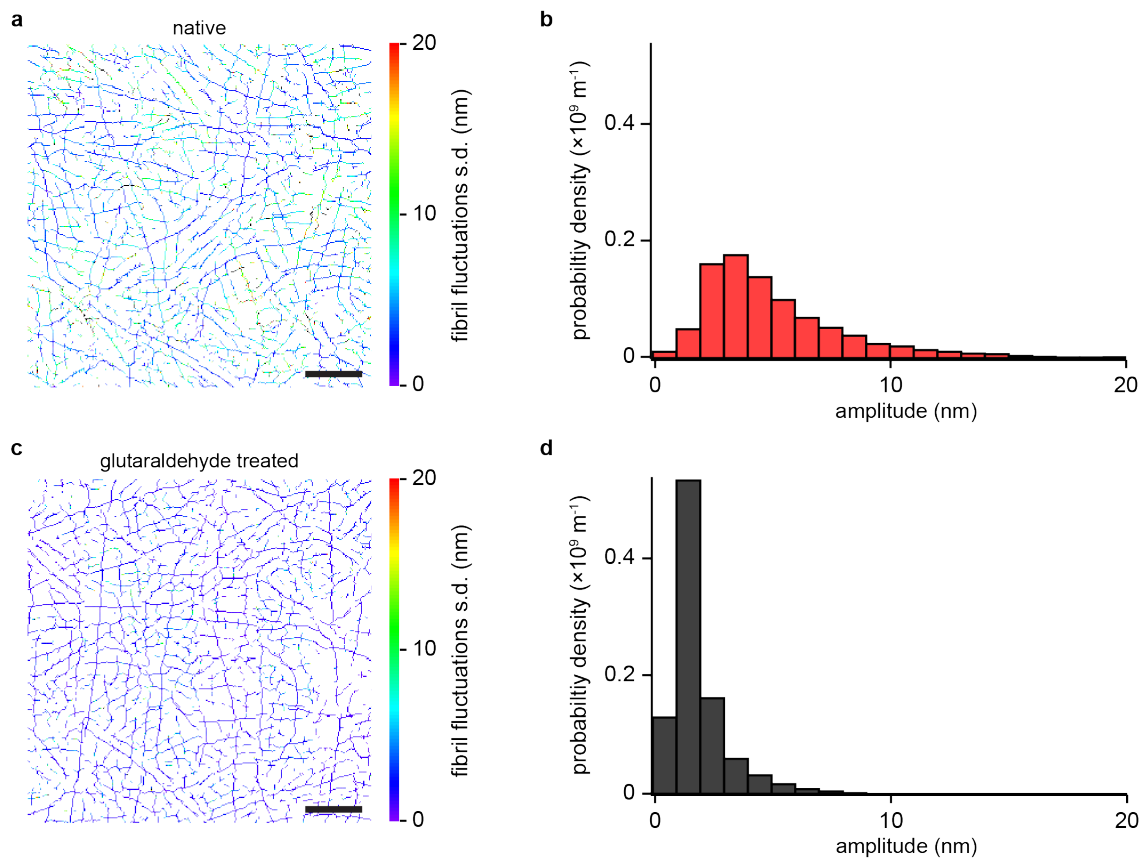
238 To demonstrate that Activity Microscopy imaging is sensitive to changes in the mechanics
239 of collagen fibrils, we investigated networks crosslinked with glutaraldehyde, a molecular
240 crosslinking agent commonly used for cell and tissue fixation. Glutaraldehyde crosslinks
241 collagen fibrils internally by coupling individual collagen molecules and connecting the
242 constituent proteins' surface lysines. It does not, however, change aspects of the network's
243 structure, such as fiber width and length, or pore size²⁰. Therefore, mainly a change in the
244 stiffness of the fibrils and a corresponding reduction in transverse fluctuations are expected.

245 **Figure 5a** and **c** show $30 \mu\text{m} \times 30 \mu\text{m}$ Activity Microscopy images of a native collagen
246 network and a crosslinked network, respectively. Both networks were prepared in the same
247 way except for the added crosslinking step (see Methods). The suppression of transverse
248 fibril fluctuation through internal crosslinking of fibrils is immediately visible as a color
249 shift corresponding to smaller fluctuations. To quantify the change in fibril fluctuations,
250 we calculated the probability density distribution from the fluctuation amplitude for all
251 fibrils shown in the Activity Microscopy images. **Figure 5b** shows such a distribution for
252 the native network. The distribution is asymmetric with a maximum around $3 - 4 \text{ nm}$ and
253 a long tail up to 20 nm . After crosslinking, the distribution shows a strong shift towards
254 smaller amplitudes with a maximum around 1 nm and a long tail up to 10 nm (**Fig. 5d**).
255 These data demonstrate the sensitivity of Activity Microscopy to subtle changes in filament
256 mechanics.

257

258

259 **Figure 5**



Reduction in collagen fibril fluctuations due to chemical crosslinking. (a) Activity Microscopy image of an *in vitro* network of native collagen I fibrils. (b) Probability density of fibril fluctuation amplitudes computed from all pixels shown in a. The peak of the distribution is located at $4.7 \pm 0.7 \text{ nm}$ (s.d. of five independent Activity Microscopy images of two different samples). (c) Activity Microscopy image of a collagen I network cross-linked with 4 % glutaraldehyde. (d) Probability density for the crosslinked network in c. Crosslinking leads to a reduction in fluctuation amplitudes with a new peak location around $1.6 \pm 0.4 \text{ nm}$ (s.d. of 4 independent Activity Microscopy images of 2 different samples). The bin width in b and d is 1 nm. Scale bars are 5 μm . The collagen networks were polymerized according to protocol I (see Methods).

260

261 **HeLa cells in a collagen I network: active forces**

262 Cells growing and migrating in collagen matrices apply forces to individual fibrils that
263 result in network deformations and remodeling⁴. To demonstrate that Activity Microscopy
264 is able to quantify and visualize these changes, we seeded HeLa cells into collagen I

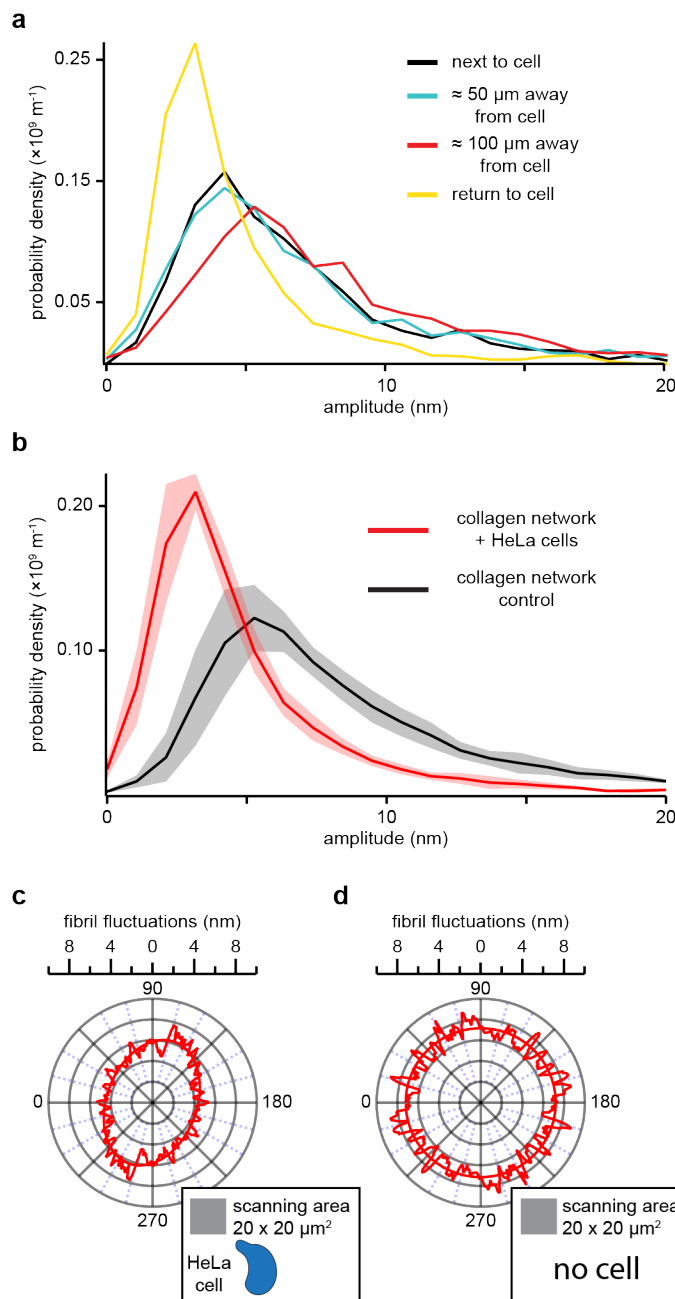
265 matrices and measured fibril fluctuations at varying distance from the cells. HeLa cells are
266 known to proliferate but not migrate in collagen^{21,22}, yet actively apply forces to the
267 network. We expect fluctuations of fibrils under tension to be reduced, and the tension to
268 fall off with the distance from the cell. **Figure 6a** shows the probability densities of
269 fluctuations amplitudes measured in $10\ \mu\text{m} \times 10\ \mu\text{m}$ areas next to (black), $50\ \mu\text{m}$ (blue)
270 and $100\ \mu\text{m}$ (red) away from a cell. While there is no significant difference between the
271 distributions measured close to the cell (peak at $3.8\ \text{nm}$) and $50\ \mu\text{m}$ away (peak at
272 $3.9\ \text{nm}$), the distribution shifts towards larger amplitudes at $100\ \mu\text{m}$ away (peak at
273 $4.9\ \text{nm}$). Each measurement took $\approx 30\ \text{min}$ to complete which is sufficiently long to
274 expect changes in the cell's state. We therefore recorded another distribution next to the
275 cell (yellow). In contrast to the first recording, fluctuations are now strongly suppressed
276 with a peak amplitude of $2.6\ \text{nm}$ and a corresponding overall contraction of the
277 distribution to smaller amplitudes. To confirm our observation, we repeated the
278 experiments several times and increased the scan area to $20\ \mu\text{m} \times 20\ \mu\text{m}$ to increase the
279 quality of the probability density distribution. For reference, we measured the distribution
280 for cell free matrices prepared under the same conditions. The average of all measurements
281 for matrices with cells (red) and without cells (black) are shown in **Figure 6b**. The
282 reduction in fluctuation amplitude relative to the cell free matrices confirms that the cells
283 apply tension to the network. This tension seems to fall off within $100\ \mu\text{m}$ as the
284 distribution measured $100\ \mu\text{m}$ away from the cell approaches the average distribution
285 measured in cell free matrices. Anisotropy in tension is an additional signature expected
286 for cells applying tension to the surrounding matrix. Cells pulling on the fibrils in their
287 surroundings will deform the network. Fibrils radiating out will carry away and distribute
288 the tension while concentric fibrils will experience reduced tension. To show this
289 anisotropy, we analyzed the fluctuation amplitude dependent on their orientation to the
290 detector coordinate axes. **Figure 6c** shows a polar graph for fluctuations recorded around
291 a point $20\ \mu\text{m}$ away from a cell. Fibrils radiating away from the cell show an $\approx 37\ \%$
292 reduction in fluctuation amplitude relative to fibrils oriented in perpendicular direction.
293 The orientation of the ellipse agrees with the location of the cell relative to the scan area
294 (inset). This can be independently confirmed by light microscopy imaging which allows
295 us to determine the position of the cell relative to the scanned area. As a control, we applied

296 the same analysis algorithm to data recorded on cell free collagen networks. In this case,
 297 as expected, fluctuation amplitudes are distributed isotropically (**Fig. 6d**).

298 In summary, these data demonstrate that the sensitivity range of Activity Microscopy
 299 imaging is well within the range of changes caused by cells to the matrix.

300

301 **Figure 6**



Collagen I network response to embedded HeLa cells. (a)

Probability density distributions of fibril fluctuation amplitudes at varying distances from a HeLa cell in an *in vitro* collagen I network. Each distribution was computed from a $10 \mu\text{m} \times 10 \mu\text{m}$ Activity

Microscopy image. Each scan took $\approx 30 \text{ min}$ to complete. (b) Average probability density distribution (red) calculated from four different $20 \mu\text{m} \times 20 \mu\text{m}$ scans next to HeLa

cells ($< 5 \mu\text{m}$). The shaded region indicates the s.d. For reference, an average probability distribution (black) calculated from four scans in a cell free network is shown. The presence of cells shifts the distribution towards smaller

fluctuations, from 5.1 nm for cell free networks to 2.7 nm for networks with embedded HeLa cells. Fibrils analyzed for **b** can be sorted by their angle relative to the detector coordinate axes and the dependence

of their fluctuations on this angle can be displayed in a polar graph (c). The fluctuation

of their fluctuations on this angle can be displayed in a polar graph (c). The fluctuation

amplitudes reveal that radially oriented fibrils (w.r.t. the cell) display $\approx 37\%$ weaker fluctuations on average than tangentially oriented fibrils (see inset). No significant anisotropy is observed in cell free networks (**d**). Here, the data from one representative cell free network from **b** were used. All probability densities distributions have a bin width of 1 nm . The polar graphs show smoothed raw data (sliding box count = 4) and an elliptical fit. The collagen networks were polymerized according to protocol II (see Methods).

303 Discussion

304 Networks made of stiff biopolymers such as collagen I are usually modeled as
305 athermal^{14,23,24}, i.e. the thermal undulations of their constituent filaments are assumed to
306 play no significant role in their mechanical response. Here, we showed that thermally
307 driven transverse filament fluctuations can nevertheless be measured, despite their
308 ångström to nanometer amplitudes. Finding the location of fibrils previous to fluctuation
309 measurements allowed us to reduce two-dimensional imaging to effective one-dimensional
310 fluctuation measurements along a fibril's contour. In this way, we were able to probe large
311 areas of a network that include many pores, thus bridging the gap between single filament
312 and overall network behavior. The observation of different pore sizes for cold and hot
313 fibrils is one prominent result of this new ability. Further, we calculated probability density
314 distributions of filament fluctuations representing the state of a network and its changes.
315 We demonstrated that these changes can originate from increased stiffness of the individual
316 fibrils in case of crosslinking or from tension applied to the fibrils in case of cells. Thus,
317 Activity Microscopy imaging covers the relevant range of fluctuation amplitudes to
318 characterize networks and their interaction with embedded HeLa cells. In both cases, the
319 thickness of the sample ranged from 0.5 – 1.0 *mm*, sufficiently large to treat the matrix as
320 a three-dimensional network.

321
322 Recently, Steinwachs *et al.*¹¹ estimated the total force applied by a breast cancer cell to a
323 collagen matrix to be on the order of 50 *nN*. As the pore size of the network is small
324 relative to the cell's volume, the total force is expected to be distributed over many fibrils,
325 with a maximum force per fibril in the nanonewton range or smaller. To estimate whether
326 transverse fibril fluctuations are still within the range of Activity Microscopy, we
327 calculated the transverse filament fluctuations based on a theory of semiflexible filaments
328 under tension for fibril lengths of 10 – 40 μm and forces up to 10 *nN* (refs. 16,25)
329 (**Supplementary Fig. 5, Supplementary Note 3**). Assuming a detection limit of 1 *nm*, all
330 fluctuation amplitudes are within the range of Activity Microscopy. Only for the shortest
331 collagen fibrils of length 10 μm and an applied tension of 10 *nN*, the detection threshold
332 of 1 *nm* is reached. Very small forces do not significantly change the fluctuation
333 amplitudes (see plateaus in **Supplementary Fig. 5**). The minimal detectable tension

334 depends on the fibril length. For example, for collagen fibrils of length $10\ \mu\text{m}$, only
335 tensions larger than $100\ \text{pN}$ significantly change fluctuations amplitudes. For $40\ \mu\text{m}$
336 fibrils, this threshold is much lower, around tens of piconewtons. These estimates show
337 that Activity Microscopy covers the relevant force range for studying cell-matrix
338 interaction also for cells that are expected to show strong interactions with the surrounding
339 matrix, such as migrating cancer cells²⁶.

340

341 Imaging speed is another factor required to follow cell-matrix interaction of motile cells.
342 For our study, we chose HeLa cells that are known to interact with collagen matrix, but are
343 unable to migrate^{21,22}, leaving sufficient time to image the selected areas of the network
344 around them. Migrating cancer cells, however, move with speeds on the order of
345 micrometers per minute²⁷, a time scale on which one would like to monitor their interaction
346 with all fibrils of the matrix. This poses a challenge for Activity Microscopy that can be
347 met with current technology. Pre-scans of volumes around a cell, required for finding all
348 fibrils in a network, can be sped up to less than a minute by using a fast scanning mode of
349 the scanning stage instead of the step by step scanning mode used in this work. This will
350 reduce the precision in localizing the fibrils, but as we pointed out above, the precise
351 knowledge of fibril position is not required for the measurement of fibril fluctuations as
352 long as the fibril remains in the linear range of the detector (**Supplementary Fig. 3**). The
353 characteristic length scale of the network is set by the pore size. The imaging speed can
354 therefore be increased by measuring a fibril's fluctuations only at single points spaced by
355 the network pore size. Assuming a pore diameter of $2.5\ \mu\text{m}$ (ref. 19), this will reduce the
356 time required to measure fibril fluctuations by a factor of 25, compared to our current
357 stepsize of $100\ \text{nm}$. These two improvements in imaging speed might be sufficient for
358 studying the interaction between motile cells and their surrounding matrix.

359

360 Activity Microscopy also has broad applications in materials research, where networks are
361 static and hence much easier to image. Collagen networks are extremely stable and
362 therefore large volumes can be imaged. We expect to gain new insight into how single
363 filament mechanics and network architecture lead to macroscopic mechanical properties.
364 This allows for a systematic study of network preparation conditions and a rational design

365 of networks with desired properties. The response to point and shear forces and other
366 network manipulations such as enzymatic activity or chemical treatment by cross-linkers
367 can also be studied.
368

369 **Materials and Methods**

370

371 **High Bandwidth, High Precision Position Detection**

372 Measurements were performed on a custom-built photonic force microscope (PFM) as
373 described in detail by Bartsch *et al.*^{15,28}. In brief, the beam of a 1064 nm laser (Mephisto,
374 500 mW, Coherent, CA, USA) was expanded and focused through a water immersion
375 objective lens (UPlanSApo, 60 × W, Olympus, Tokyo, Japan) into the sample. The sample
376 chamber was mounted on a three-dimensional nano-positioning stage (Nano-
377 View/M375HS, Mad City Labs, WI, USA), which allowed for the sample to be moved
378 relative to the stationary optical trap. Forward scattered light from a collagen fibril, together
379 with unscattered light of the laser beam was collected by a condenser lens and projected
380 onto a quadrant photodiode (G6849, Hamamatsu Corporation, NJ, USA), where the two
381 waves interfered. The differential signal from the QPD were amplified by custom-built low
382 noise differential amplifiers (SA500, Oeffner MSR, Plankstadt, Germany). This detection
383 scheme has a megahertz electronic bandwidth.

384

385 **In vitro Collagen I Networks – Polymerization Protocol I**

386 Collagen networks were prepared and polymerized *in vitro* as described in Bartsch *et al.*¹⁵.
387 In brief, acid-soluble rat-tail tendon collagen (Collagen I, rat tail, 354236, Corning®, NY,
388 USA) and bovine-dermis collagen (Collagen I, bovine, 354231, Corning®, NY, USA) were
389 mixed at relative concentrations of 1:2. The mixture was then diluted to a total collagen
390 concentration of 2.4 mg ml⁻¹ by adding equal parts of 10 × DMEM (D2429, Sigma
391 Aldrich, MO, USA) and 0.27 M NaHCO₃. To induce gel polymerization, the pH of the
392 solution was raised to pH 10 using 1 M NaOH. All components were kept on ice during
393 mixing. The mixture was then quickly pipetted into a preassembled sample chamber
394 consisting of a glass coverslip attached by vacuum grease to a metal sample chamber
395 (**Supplementary Fig. 7**) and left to polymerize for ~ 1 h at 37° C in a humidity-controlled
396 incubator with 5% CO₂ atmosphere. Polymerizing the network inside the sample chamber
397 ensures its attachment to the coverslip, which is a prerequisite for a mechanically stable
398 assay. Networks were between 500 μm and 1 mm thick. After polymerization, the gel was

399 gently rinsed with 1 ml of 1 × Phosphate-buffered saline (PBS). Care was taken to never
400 let the network dry out.

401 After polymerization of the collagen network, the sample chamber was closed by
402 attachment of a top coverslip (**Supplementary Fig. 7**) and the chamber was filled with
403 1 × PBS before being completely sealed off with vacuum grease.

404

405 **Collagen Network Crosslinking**

406 After polymerization of the collagen network as described above, approximately 100 µl of
407 4 % *v/v* glutaraldehyde (GA) in deionized water, was pipetted on top of the collagen and
408 the sample was placed back into the incubator for ~ 2 h. After incubation, the collagen was
409 thoroughly rinsed with 1 × PBS and the sample chamber was assembled and mounted on
410 the PFM.

411

412 **Cell Culture**

413 HeLa cells were kindly gifted to us by Prof. Aaron Baker (Biomedical Engineering
414 Department, The University of Texas at Austin). HeLa cells were adapted to and cultured
415 in CO₂ independent medium (Gibco™, 18045088), which allows for cell culture under
416 atmospheric conditions. The medium was supplemented with 10 % bovine calf serum
417 (GE, HyClone, SH30072), 4 mM L-glutamine, 100 IU ml⁻¹ penicillin and 100 µg ml⁻¹
418 streptomycin at 37° C. Cells were passaged every three days. To detach cells from the
419 culture flask, the disassociation agent TrypLE™ Express Enzyme 1 × (ThermoFisher
420 Scientific, 12605036) was used.

421

422 **Cell Seeding in Collagen Matrix – Polymerization Protocol II**

423 To produce collagen networks with seeded HeLa cells, the collagen network protocol as
424 described above had to be altered in order to incorporate the cell culture medium with a
425 buffer system independent of CO₂ control as the sample is subject to atmospheric
426 conditions while mounted on the PFM. Similarly to the above, acid-soluble rat-tail tendon
427 collagen (Collagen I, rat tail, 354236, Corning®, NY, USA) and bovine-dermis collagen
428 (Collagen I, bovine, 354231, Corning®, NY, USA) were mixed at relative concentrations
429 of 1:2 on ice. The mixture was then diluted with 10 % (*vol/vol*) 10 × PBS and the pH

430 neutralized to $\text{pH} = 7.2 - 7.4$ with 1 M NaOH and kept on ice. HeLa cells in culture at
431 $80 - 90 \%$ confluency were detached from the culture flask using the disassociation agent
432 TrypLE™ Express Enzyme $1 \times$ (ThermoFisher Scientific, 12605036) and spun down at
433 $125 \times g$ for 10 min . The cells were then resuspended in serum-free CO_2 independent
434 medium (Gibco™, 18045088) and supplemented with 4 mM L-glutamine at a
435 concentration between $(40 - 80) \times 10^4 \text{ cells ml}^{-1}$. Finally, the medium containing the
436 cells was added to the neutralized collagen solution on ice at a relative concentration of
437 20% (*vol/vol*) to yield a final cell concentration between $(7 - 15) \times 10^4 \text{ cells ml}^{-1}$,
438 and collagen concentration of 2.4 mg ml^{-1} . The solution was carefully mixed, pipetted
439 into the preassembled sample chamber and placed in a humidity controlled 37° C incubator
440 for 1 h . The network was then topped with $\sim 100 \mu\text{l}$ of serum-free CO_2 independent cell
441 media and allowed to incubate overnight.

442

443 **Calculating the angle of fibril axis**

444 Having determined all locations belonging to the fibril axis, we assign an orientation angle
445 value to each individual location, termed filament location in the following. Choosing one
446 filament location after another as the center, we sweep a full circle of given radius r (here:
447 $r = 4 \text{ pixels}$) in discrete angle steps of $\pi/100$. For every step we calculate an alignment
448 score corresponding to the number of filament locations that are aligned with the straight-
449 line segment connecting $-r$ and r . The alignment score can take a value between 0 and $2r$
450 pixels. We then find the angle corresponding to the maximum by Gaussian fitting, which
451 corresponds to the local fibril angle at this location. If a filament location is not connected
452 to any others, therefore alone standing, it is discarded.

453

454 **Calculation of Collagen Fibril Fluctuation, Background Correction and Thresholding**

455 While performing Activity Microscopy measurements a small part of the signal can be
456 attributed to background noise, which arises primarily from four different sources:
457 electronic noise in the amplifier and stage control, laser power fluctuations, mechanical
458 instabilities in the PFM setup, as well as background contributions by parts of the network
459 out of focus. Those noise sources could be characterized individually, but it is simpler to
460 estimate the sum background signal by positioning the laser focus inside of network pores

461 in a collagen sample, far away from fibrils. We determined that on average the s.d. of the
462 background signal is $\bar{\sigma}_{background,x} = 12.56 \pm 0.39 \text{ mV}$ and $\bar{\sigma}_{background,y} = 12.33 \pm$
463 0.92 mV . Assuming all signals to be independent random variables we can subtract the
464 variance of the background signal $\bar{\sigma}_{background,x}^2$ from the variance of the signal measured
465 in volts $\sigma_{S_x}^2(x, y)$ to get a fibril's true motion:

$$466 \quad u_x(x, y) = \frac{\sqrt{\sigma_{S_x}^2(x, y) - \bar{\sigma}_{background,x}^2}}{\partial \bar{S}_x / \partial x(x, y)}, \quad (3)$$

467 where $\partial \bar{S}_x / \partial x(x, y)$ is the local detector sensitivity. An analogous expression is valid for
468 the y -signal. In addition, we chose the smallest acceptable detector sensitivity as
469 20 mV pixel^{-1} for pixels with side lengths of 100 nm . Any locations with local sensitivity
470 values below our threshold were deleted. Around all measured scanning areas a two pixel-
471 wide frame was deleted as the spatial derivative for finding fibril locations is ill defined at
472 the edges. For larger scans that required the successive scanning of multiple subsections, a
473 $1 \mu\text{m}$ overlap between the subsections prevented gaps in the data after deleting pixels at
474 the edges.

475

476 **Pore Size Distribution for Hot versus Cold Fibrils**

477 An Activity Microscopy image displaying fibril fluctuations is first divided into equally
478 sized sets of “hot” and “cold” fibrils by finding the median of the fluctuation amplitude
479 distribution. We then perform a pore size analysis partially adapted from Mickel *et al.*¹⁹,
480 on three Activity Microscopy images, albeit in two dimensions. In brief, we first convert
481 the image into a simple binary dataset, where all pixels that belong to a fibril are assigned
482 an arbitrarily high value, and all pixel belonging to a network pore are assigned the value
483 zero. We then determine for every pore pixel the largest disk to fit into a given network
484 pore without intersecting any pixels belonging to a fibril. Each pore pixel in the image is
485 then assigned the radius of the largest disk that incorporated said pixel as its value.

486

487 **Data Analysis, Visualization, and Code Availability**

488 The data were acquired and analyzed using custom software written in Labview (National
489 Instruments, TX, USA) and Igor Pro (Wavemetrics, OR, USA) and is available from the
490 corresponding author upon request.

491

492 **Data Availability**

493 The data that support the findings of this study are available from the corresponding author
494 upon reasonable request.

495

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501

502 **Author Contributions**

503 E.N.L., T.F.B., and E.-L.F. conceived the experiments. E.N.L. performed the
504 experiments, developed analysis software and analyzed data. T.F.B. and E.N.L. wrote
505 software to control the instrument. T.F.B. wrote the software to calculate the angle of a
506 fibril's axis. E.N.L., T.F.B, and E.-L.F. wrote the manuscript.

507

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