We study the micromechanics of collagen-I gel with the goal of bridging the gap between theory and experiment in the study of biopolymer networks. Three-dimensional images of fluorescently labeled collagen are obtained by confocal microscopy, and the network geometry is extracted using a 3D network skeletonization algorithm. Each fiber is modeled as an elastic beam that resists stretching and bending, and each crosslink is modeled as a torsional spring. The stress–strain curves of networks at three different densities are compared with rheology measurements. The model shows good agreement with experiment, confirming that strain stiffening of collagen can be explained entirely by geometric realignment of the network, as opposed to entropic stiffening of individual fibers. The model also suggests that at small strains, crosslink deformation is the main contributor to network stiffness, whereas at large strains, fiber stretching dominates. As this modeling effort uses networks with realistic geometries, this analysis can ultimately serve as a tool for understanding how the mechanics of fibers and crosslinks at the microscopic level produce the macroscopic properties of the network. © 2010 Wiley Periodicals, Inc. Complexity 16: 22–28, 2011

Key Words: biopolymer; collagen; strain stiffening

1. INTRODUCTION

Collagen is the most abundant animal protein [1] and its mechanics have been studied in great detail [2]. It takes on many morphologies, including skin, tendons, ligaments, individual fibers, and gels. Of particular interest is the mechanics of collagen-I gels, shown in Figure 1(a). These gels provide a relatively simple structure that can be noninvasively observed by confocal microscopy [3, 4] and used as a scaffold for growing artificial tissues [5], and as a 3D environment for studying cell motility [6] and tumor invasion [7, 8]. A critical first step in understanding these systems is to develop a model for the collagen gel alone. In this article, we give a successful theoretical model of the micromechanics of realistic networks.
Collagen-I gels belong to a class of materials known as biopolymers. Other examples include actin, found in the cytoskeleton, and fibrin, a component of blood clots. Common mechanical features of biopolymers are negative normal stresses under shear [9], significant contraction and alignment as the gel is axially stretched [10, 11], and strain stiffening by 2–3 orders of magnitude at large strains, as shown in Figure 1(f). The cause of this strain stiffening is not well understood. Storm et al. [12] attributed strain stiffening in all biopolymer networks exclusively to the pulling out of entropic modes of individual filaments. Their calculation required the assumption that deformations are affine. On the other hand, Heussinger et al. [13, 14] showed how one could deconstruct the network deformation into a set of nonaffine floppy modes. They concluded that accounting for the nonaffinity was necessary in describing the elastic properties of the network. They hypothesize that the non-linearity comes from a transition from a bending regime to a stretching regime. Once and coworkers [15, 16] have proposed the alternative hypothesis that strain stiffening is due to the rearrangement of the fibers as the network is strained.

Resolving this debate has been difficult as almost all theoretical analysis has been on artificially generated networks in 2D. Whether such artificially generated networks can be used to accurately describe real biopolymer gels has not yet been confirmed. In particular, key results of the above theories rest on assumptions on how the crosslink spacing scales with density. Although a direct relationship has been obtained in two dimensions for random stick networks [17, 18], in three dimensions, the scaling relationship will also depend on how the gel polymerizes. Although the nature of collagen polymerization is currently subject of active research [19], it is still not well understood. Moreover, the few examples of quantitative comparisons to experiment in the literature [12, 20, 21] are not able to quantitatively fit the full stress–strain response of the gel at varying densities using a single set of parameters. The goal of this research is to develop a model for collagen gel based on three-dimensional images of fluorescently labeled collagen gels at different densities. The images are obtained by confocal microscopy [Figure 1(a)], and the network geometry is extracted using a custom Fibre Extraction (FIRE) algorithm [Figure 1(b)] [3]. The gel is modeled as a random network of crosslinked fibers, as described below, and the stress–strain response is compared with that measured by an AR-G2 rheometer. Good agreement between model and experiment is obtained by fitting a single parameter, that is, the crosslink stiffness.

2. EXPERIMENT

The experiments were defined in more detail previously [3]. Briefly, we labeled bovine collagen type I with the fluorescent molecule TAMRA. Collagen solution at final concentrations of 0.5, 1.0, and 1.5 mg/mL were prepared, and a Leica SP5 resonant confocal microscope with a 63X 1.2-NA water immersion objective was used to image the collagen samples. Four networks at each density were extracted using the FIRE skeletonization algorithm [3], as shown in Figure 1(b). After applying FIRE, the collagen density was re-estimated in two different ways: by counting the number of bright pixels in the image and by the product of the average fiber cross-sectional area ($A = 2800 \text{ nm}^2$ with an assumed radius of 30 nm [3, 22]) and total fiber length in the extracted networks. The re-estimated densities of the three networks were 0.5, 1.2, and 1.4 mg/mL. The final step before simulation is to remove dangling portions of the network that do not contribute to its stiffness [Figure 1(c)]. The shear response of the gel was measured using an AR-G2 rheometer with a 4°, 40-mm cone-plate.
geometry with a 109-μm gap. Approximately 1.2 mL of collagen solution was pipetted onto the 37°C preheated bottom plate of the rheometer, and the cone was lowered onto the sample. A solvent trap was used to prevent the sample from drying during the measurement.

3. MODEL

In the model for the collagen gel, each fiber is treated as an elastic beam that resists stretching and bending and each crosslink is treated as a torsional spring, making it more stiff than a freely rotating pin joint but less stiff than a welded joint of fixed angle. The stretching modulus of an individual fiber is given by \( K_x \), where \( E \) is the Young’s modulus and \( A \) is the cross-sectional area. The Young’s modulus of a fiber in aqueous conditions has been estimated to be between 30 and 800 MPa [23–25], and we use a modulus of 50 MPa, which fits the data well and is also close to the value chosen by Stylianopoulos and Barocas [21].

It has been shown that a single fiber will stiffen by a factor of 2–4 when strained [25]. We choose here to use a constant \( f \) both to reduce the number of parameters in the model and to see if geometric reorientation of the network is enough to explain strain stiffening. Stylianopoulos and Barocas [21] also explored the bilinear and exponential constitutive relations for the individual fibers and observed only minor effects on the macroscopic network behavior. The radius of each fiber is given by \( r \), and the cross-sectional area is known about the nature of the naturally formed collagen crosslinks. We find that we can fit all the data by setting the torsional spring stiffness to \( K_x = 32 \text{ pN-m} \cdot \mu \text{m} \), where \( I = r^4/4 \) [26]. No crosslinking agent has been added to the gel, and very little is known about the nature of the naturally formed collagen crosslinks. We find that we can fit all the data by setting the torsional spring stiffness to \( K_x = 32 \text{ pN-m} \cdot \mu \text{m} \). To compare the previously reported tensile modulus of a sample. In shear, we do not allow the boundary nodes to move freely. We compare the shear results to cone-plate rheometer experiments, where the shear faces are observed to be the rheometer. Here, the distance between \( F_1 \) and \( F_3 \) is 109 μm, and the simulated network represents a small region near the center of a sample. In shear, we do not allow the boundary nodes on \( F_1 \) and \( F_3 \) to move freely. We compare the shear results to cone-plate rheometer experiments, where the shear faces are observed to be the rheometer. Here, the distance between \( F_1 \) and \( F_3 \) is 109 μm, and the simulated network represents a small region near the center of a sample. In shear, we do not allow the boundary nodes on \( F_1 \) and \( F_3 \) to move freely. 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Figure 2. The close agreement between model and experiment indicates that strain stiffening due to the geometric rearrangement of the collagen fibers is enough to explain the strain stiffening seen in experiments.

At large strains (~50% in tension and ~200% in shear), the stress–strain curve of the model becomes linear again, although with a much steeper slope. In Figure 3(b), we compare the large strain tensile behavior of the model to the experiments of Roeder et al. [28]. Although our model underestimates their experimental measurement by a factor of 2.5, we find this to be reasonable as the two experiments used different collagen protocols. In particular, different buffers were used. In Figure 2, we also explore the case where \( K_x = K_b = 0 \), such that we have only a network of springs connected at freely rotating pin joints. At low strains, the network can be deformed without exerting any stress, but at strains higher

the stress–strain response is linear, as expected, and at larger shear strains, the stress–strain response stiffens considerably. In Figure 3(a), we show that the small strain modulus scales by \( \sigma_{12} \sim \rho^{2.68} \epsilon_{12} \), where \( \rho \) is the collagen density. At this time, it is not possible to verify the power law scaling in the model as only densities of 0.5, 1.2, and 1.4 were observed. The fluorescent labeling of the network changes the polymerization properties of the network, causing it to clump at higher densities. We use this scaling relationship to collapse the curves in Figure 3(b). The small strain shear modulus is compared with the cone–plate rheometer experiments, and a scaling law of \( G' \sim \rho^{2.68} \) is observed for the experiment. The large strain tensile modulus is from the model and from the experiments of Roeder et al. [28]. Results differ by a factor of 2.5, which is reasonable as the two experimental protocols were different [28].
FIGURE 4

Average negative normal stresses for the 1.2 mg/ml gel in response to shear.

than 25%, we see that this simplification adequately describes the gel.

A topic of investigation explored by many is the validity of the assumption that these networks deform affinely [14, 15]. For brevity, we only state that the deformations are highly nonaffine at small strains where the majority of the energy is stored in bending rather than stretching. This is best seen in Figure 2. Affine deformations have the property that the majority of the energy stored in the network is in stretching the fibers, but here, fiber stretching makes no contribution to the network stiffness until strains are greater than 10%.

In response to shear stress, the simulated networks also exhibited negative normal stresses that grow quadratically with strain (Figure 4), as observed in the experiments of Janmey et al. [9]. We also examined the perpendicular contraction of the gel in response to axial strain in comparison with the experiments of Vader et al. [11] and Roeder et al. [4], as shown in Figure 5. The perpendicular strain was taken to be the perpendicular displacement divided by the original perpendicular distance from the origin averaged over all the nodes. Here, the model does not match the experiments well. In particular, both experiments exhibited considerably more perpendicular compression than seen in the simulation. One possible explanation is that the geometry of the experiments done here (sheared cone-plate rheometer) is considerably different from that in the other experiments (stretched disk and stretched 3D I-beam). Another possibility is that the FIRE network extraction algorithm overestimates the number of crosslinks of the network. It may also be that some of the crosslinks in the network slip, thus allowing the network to further compress in response to axial tension. To give the same strain-stiffening behavior, this would require stiffer crosslinks. Further exploration, however, was beyond the scope of this article. Despite the lack of quantitative agreement, we note that nonlinearly increasing perpendicular strain is qualitatively similar to that observed in Vader et al. [11].

5. SUMMARY AND DISCUSSION

In summary, we have presented a microstructural model of a 3D biopolymer gel using a network geometry that is based on the true network architecture. It differs from previous work in that we use realistic network architectures that have been extracted using the FIRE algorithm. We specifically focus on the mechanics of collagen-I networks, but we emphasize that this modeling approach is generalizable to other biopolymer networks. The model has three parameters: \(E, r, K_x\). The fiber radius and tensile modulus can be measured experimentally, and the model uses realistic parameters. The crosslink torsional spring constant must be fit to the data and we used \(K_x \sim 20k_BT/l\). Fitting this single parameter gives the right strain-stiffening behavior for networks at three different densities at strains that vary from 0.5 to 50%. This result lends support to the hypothesis put forward by Oruc et al. [15] that strain stiffening in polymer networks and particularly collagen-I gels is governed by rearrangement of the gel. Here, we note that the strain-stiffening mechanisms for collagen are likely different than for other gels. Because of its long persistence length (1 cm) fibers, the effects of entropic stiffening in these gels is known to be minimal. Therefore, the entropic stiffening hypothesis [31] is not applicable here. However, for other biopolymer gels such as actin, where the...
underestimated the actual large strain modulus by a factor of two. This difference could potentially be explained by a stiffening of the individual fibers by a factor of two at large strains, but confirmation would require additional experiments measuring the stress–strain properties of individual fibers in these collagen gels.

We note that this model has been designed to capture the intermediate time scale behavior of the network (minutes to hours), where the network behaves as an elastic solid. The viscous modulus is known to be a factor of 10 less than the elastic modulus within the range of 0.1–100 Hz [36, 37], but at even faster time scales, the viscosity of the fluid will play a critical role. At slower time scales, the assumption that the crosslinks remain relatively fixed [36] breaks down as plastic deformation and creep occur. This simplified model provides a starting point in the development of a more complete model of collagen gel. Ultimately, a more sophisticated approach, such as that taken by Rodney et al. [38] will be necessary to capture the full dynamic behavior of the gel, where crosslinks are allowed to slip and break.

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