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Mini Review 1

Mechanical and Systems Biology of Cancer 2

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ABSTRACT

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50	1.4 Conclusions and Outlook

Mechanics and biochemical signaling are both often deregulated in cancer, leading to cancer cell phenotypes that 19 exhibit increased invasiveness, proliferation, and survival. The dynamics and interactions of cytoskeletal compo- 20 nents control basic mechanical properties, such as cell tension, stiffness, and engagement with the extracellular 21 environment, which can lead to extracellular matrix remodeling. Intracellular mechanics can alter signaling and 22 transcription factors, impacting cell decision making. Additionally, signaling from soluble and mechanical factors 23 in the extracellular environment, such as substrate stiffness and ligand density, can modulate cytoskeletal 24 dynamics. Computational models closely integrated with experimental support, incorporating cancer-specific 25 parameters, can provide quantitative assessments and serve as predictive tools toward dissecting the feedback 26 between signaling and mechanics and across multiple scales and domains in tumor progression. © 2018 Spill et al., Published by Elsevier B.V. on behalf of the Research Network of Computational and Structural 28 Biotechnology. This is an open access article under the CC BY-NC-ND license

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1. Introduction 55

The mechanical microenvironment in cancer is vastly altered 56 compared to healthy tissue. Typically, the extracellular matrix (ECM) 57 is stiffened in the tumor microenvironment [1–3], but individual cancer 58 59 cells may actually be softer [4]. There is a bimodal distribution of nano-60 mechanical stiffness across advanced cancer tissues [5]. Moreover, more 61 complex mechanical and geometric characteristics, including the

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fibrous matrix structure, porosity, or viscoelastic parameters may be 62 changed in tumors [6, 7]. Similarly, solid and fluid stresses are greatly 63 altered in cancers [8]. It is well known that cancers exhibit increased 64 fluid pressures, in part due to remodeling of the vasculature and 65 lymphatics [9]. 66

The altered ECM stiffness and geometry of the tumor microenviron- 67 ment are sensed by tumor cells via mechanosensing structures, which 68 can activate intracellular signaling pathways that drive behaviors such 69 as unrestrained proliferation, increased survival, tissue invasion, 70 stemness, and drug resistance [10-12]. While cancer has been tradition-71 ally considered a genetic disease, alterations in ECM stiffness and geom-72 etry can force normal cells to adopt phenotypes characteristic of 73 transformed and/or metastatic cells in the absence of any genetic 74

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change [13, 14]. Theoretical work suggests that environmental cues,
coupled with various possible oncogenic alterations (e.g. overexpression of c-Src [15]), can drive cancer progression [16, 17]. Cancer
progression can be promoted by genetic changes that alter how cells
respond to ECM stiffness and geometry and that enable cancer cells to
remodel their environment in ways that promote disease.

81 To open new therapeutic avenues that seek to manipulate the 82 response of cancer cells to their environment as a way to treat 83 cancer, predictive mathematical models are required to describe 84 how cell fate decisions are due to interactions between tumor 85 cells and their ECM and how these interactions differ between normal and cancer cells. The problem is inherently multiscale in 86 nature and involves diverse components such as biochemical 87 88 reactions, cell-matrix and cell-cell interactions, and tissue-level 89 alterations. The field of mechanotransduction has long embraced 90 modelling tools in order to describe how cells respond to mechanical and geometric cues, and these models serve as key starting 91 92 points for more complex descriptions of how cancer cells interact 93 with their ECM. For example, models have been developed that provide insights into diverse aspects of mechanobiology including: 94 force-dependent molecular bonds [18-21], spatiotemporal 95 96 organization of intracellular molecules [22-24], impact of cell 97 shape [25–29], and the dynamics of the cytoskeleton [30–32]. Here we review some of these models and supporting experimen-98 tal findings with a look toward the future. We first review recent 99 work on cytoskeletal interactions that modulate intracellular 100 mechanics and the propagation of cytoskeletal forces inside and 101 102 outside the cell. Next we focus on the cell-matrix adhesion complexes that act as key signal transducers and mechanosensors. 103 Finally, we review key signaling networks implicated in 104 105 mechanotransduction.

106 1.1. Generation and Propagation of Intracellular Forces

The active actin cytoskeleton provides basic structure and force 107 generation capabilities. The key components include actin filaments, 108 actin crosslinking proteins (ACPs) such as alpha-actinin and filamin, 109 110 and myosin II motors that generate contractility. Inside the cell, a large network of these components undergoes dynamic and stochastic 111 interactions, spontaneously resulting in pattern formation – including 112 the actin cortex at the cell periphery, thick contractile bundles of actin 113 114 (stress fibers), and cell polarity (leading and trailing edges). Local interactions and kinetics can control overall, global functionality of the 115 cytoskeletal network. In particular, actin turnover rates can modulate 116 117 cytoskeletal network tension, and the interplay between actin turnover, 118 actin crosslinking, and myosin II walking activity can regulate the 119 morphological state of the network, from homogeneous morphologies to local clusters (Fig. 1a) [30]. Computational simulations can isolate 120 individual features and determine their roles in cytoskeletal network 121 behavior. For example, altering actin nucleation rates can modulate 122 the stress fluctuation magnitudes in the cytoskeleton, a phenotype 123 124 observed in intracellular microrheology experiments that modulate 125 epidermal growth factor (EGF) signaling (known to influence actin nucleation) in breast cancer cells [33]. Additionally, spatial and tempo-126 ral profiles are important in regulating cell behavior. These can be 127 precisely tuned in computational models. For example, cell geometry 128 129 and dimensionality influence the anisotropy and amplitude of intracellular stress fluctuations [34]. While overall cell tensions have an 130 intuitive role of enabling cells to apply forces onto their substrate (e.g. 131 the ECM) and migrate, intracellular stress fluctuations can facilitate 132 the redistribution of organelles and molecular components inside the 133 crowded cytoplasmic space [35]. Furthermore, malignant tumor cells 134 appear to exhibit larger intracellular displacement and stress fluctua-135 tions compared to benign counterparts, as shown by experiments 136 measuring intracellular stiffness and force fluctuations [35]. Cytoskele-137 138 tal mechanics and fluctuations are the result of the interactions between many cytoskeletal components, each undergoing dynamic processes 139 (turnover, walking, binding, unbinding, etc.). Computational network 140 models of the cytoskeleton, based on physical principles (reaction 141 kinetics, mechanics) and incorporating realistic, experimentally 142 tangible features, can help dissect the local, molecular-level contribu- 143 tions to experimentally observable mechanical cellular phenotypes. 144 High resolution experimental techniques, e.g. super resolution imaging 145 or atomic force microscopy, can help guide the development and 146 validation of models of fine and distinct cytoskeletal features [36]. 147 Furthermore, models coupling cytoskeletal forces to critical intracellular 148 and extracellular features, particularly the nucleus and the ECM, can 149 start to elucidate a more holistic picture of cell behavior. 150

Cytoskeletal forces can be transmitted to the cell nucleus via the 151 LINC (Linker of Nucleoskeleton and Cytoskeleton) complex [37]. 152 Substrate stiffness modulates cytoskeletal tension and thus nuclear 153 stress and shape, which interestingly also modulates the expression 154 levels of a key nucleoskeletal protein lamin A, nuclear stiffness, and 155 stem cell differentiation [38]. The mechanical properties of the nucleus 156 can also influence nuclear shape and dynamics during cell deformation 157 and invasion through confined spaces (e.g. ECM pores or endothelial 158 junctions). Large nuclear deformations can lead to rupture and DNA 159 damage, as observed in experimental studies of cancer cells invading 160 through highly confined constrictions [39, 40]. Computational models 161 coupling cellular forces to the nucleus can generate quantitative details 162 of nuclear deformation and mechanical remodeling during physiologi- 163 cal processes and draw insights toward differences in nuclear behavior 164 due to biochemical or structural alterations. For example, experiments 165 show that lamin A/C deficiency leads to more plastic remodeling of 166 the nucleus after larger strains, which can be captured in a continuum 167 model of the nucleus featuring a hyperelastic shell and a poroelasto- 168 plastic core (Fig. 1b) [41]. Furthermore, the role of different types of 169 lamins (A and B) in regulating nuclear shape and geometry can be 170 explored in continuum models through incorporating heterogeneous 171 material profiles. In particular, a preferred mesh size difference between 172 lamin A and lamin B appears to explain nuclear blebbing tendencies 173 [42]. 174

In many types of solid tumors, cancer cells are embedded in a dense 175 fibrillar matrix. Cytoskeletal forces are transmitted into the ECM via 176 cell-matrix adhesions, which can lead to ECM remodeling and propa- 177 gate mechanical signals to surrounding cells [43]. Stiffer substrates 178 tend to promote increased cell traction forces and lead to a more 179 invasive phenotype [44, 45]. Relaxation of tension in the substrate in 180 laser ablation experiments [46, 47] tends to revert cell invasiveness. 181 Moreover, ECM networks exhibit nonlinear strain stiffening [48], 182 suggesting potential mechanical feedback mechanisms. These phenom- 183 ena have been demonstrated through a number of experimental 184 studies. Complementarily, computational models can provide quantita- 185 tive, mechanistic insights toward underlying driving factors of invasive 186 behavior in 3D ECMs - particularly to a level of detail that may be 187 unfeasible for experiments to achieve or parse out. Computationally 188 intensive models can capture a high degree of local details observed in 189 high resolution experiments of cell-ECM interactions. In a recent 190 study, a model capturing an entire cell with dynamic protrusions inside 191 a surrounding ECM showed that dynamic filopodia can act as rigidity 192 sensors that facilitate durotaxis in HUVECs (Fig. 2a) [49]. While stiffness 193 sensing (and many other cell behaviors) is a phenomenon exhibited by 194 normal and cancer cells, cancer-related parameters can be tuned in 195 generalizable models to explore disease phenotypes. In particular, the 196 above model showed that the number and length of filopodia can 197 modulate invasive behavior, supporting prior studies that showed that 198 deregulation in filopodia-related functions and pathways are implicated 199 in cancer progression and metastasis [50]. In another model that incor- 200 porates dynamic local forces and force-sensitive ECM fiber-fiber 201 crosslinks, it is demonstrated that the coupling of mechanical forces 202 and fiber-fiber biochemical kinetics can result in ECM densification 203 near the cell boundary, consistent with experiments in tumor and 204

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