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Evaluating the influence of mechanical stress on anticancer treatments through a multiphase porous media model

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Highlights

- Low-proliferating cell populations contribute to drug resistance
- Compressive stresses act on tumors by inhibiting cell proliferation
- We analyze the combined effects of drug action and mechanical compression
- Mechanical compression of tumors may compromise drug efficacy

Abstract

Drug resistance is one of the leading causes of poor therapy outcomes in cancer. As several chemotherapeutics are designed to target rapidly dividing cells, the presence of a low-proliferating cell population contributes significantly to treatment resistance. Interestingly, recent studies have shown that compressive stresses acting on tumor spheroids are able to hinder cell proliferation, through a mechanism of growth inhibition. However, studies analyzing the influence of mechanical compression on therapeutic treatment efficacy have still to be performed. In this work, we start from an existing mathematical model for avascular tumors, including the description of mechanical compression. We introduce governing equations for transport and uptake of a chemotherapeutic agent, designed to target cell proliferation. Then, model equations are adapted for tumor spheroids and the combined effect of compressive stresses and drug action is investigated. Interestingly, we find that the variation in tumor spheroid volume, due to the presence of a drug targeting cell proliferation, considerably depends on the compressive stress level of the cell aggregate. Our results suggest that mechanical compression of tumors may compromise the efficacy of chemotherapeutic agents. In particular, a drug dose that is effective in reducing tumor volume for stress-free conditions may not perform equally well in a mechanically compressed environment.

31 **Keywords**

32 Cancer; Chemotherapy; Tumor Spheroids; Mathematical Model; Mechanical Compression

33

34 **1. Introduction**

35 A major hurdle to chemotherapy success is resistance of tumor cells to therapeutic agents. In
36 general, resistance may arise as an intrinsic cellular response or as a result of drug treatment
37 (Zahreddine and Borden, 2013). It is known that the presence of a low-proliferating cell population
38 is one of the leading factors contributing to drug resistance in solid tumors (Mueller-Klieser, 2000;
39 Trédan et al., 2007). In fact, several chemotherapeutic agents are effective against rapidly dividing
40 cells. Moreover, as certain normal tissues display high rates of cellular divisions (such as the gut
41 mucosal and bone marrow cells), there exists a toxicity limit determining the maximum
42 administrable drug dose (Dawidczyk et al., 2014).

43 Such resistance mechanisms, dependent on the proliferative activity of tumor cells, are generally
44 investigated *in vitro* through the use of three-dimensional cell aggregates, known as tumor
45 spheroids (Vinci et al., 2012). Contrary to conventional monolayer cultures, tumor spheroids display
46 heterogeneous cell populations, including quiescent and necrotic cells, together with resistant
47 phenomena to different chemotherapeutic drugs (Mikhail et al., 2013). Cell quiescence results both
48 from the lack of nutrients and growth factors within the tumor, and from adhesion interactions
49 between cells of the same type. Indeed, cells from healthy tissues display a mechanism of “contact
50 inhibition” that regulates proliferation in a crowded environment (Abercrombie and Ambrose,
51 1962). This mechanism allows the cells to stop proliferation as soon as certain cell densities are
52 reached at a given site. Tumor cells exhibit an analogous behavior, even though to a significant lesser
53 extent than their healthy counterpart, and with more relevance in three-dimensional cultures than
54 in monolayers (St Croix et al., 1998).

55 The biochemical pathways underlying contact inhibition are still an active area of research. They are
56 linked to adhesive interactions between neighboring cells, mediated by adhesion proteins such as
57 cadherins. Moreover, these mechanisms include a series of proteins involved in cell cycle regulation.
58 To this regard, the G1 checkpoint, also known as the restriction point (R), represents a fundamental
59 step in the cell cycle, controlling cell commitment to mitosis (Planas-Silva and Weinberg, 1997).
60 Regulation of this cell checkpoint depends on the retinoblastoma protein (pRb). In particular, the

61 hypo-phosphorylated form of pRb prevents progression from the G1 to the S phase of the cell cycle,
62 inhibiting cell duplication. On the other hand, phosphorylation of pRb leads to its inactivation
63 allowing the cell to undergo mitosis. Phosphorylation of pRb depends on cyclin-dependent kinases
64 (cdks), which in turn are subject to the action of cyclins (Dietrich et al., 1997). Finally, the activity of
65 the whole complex is further regulated by several inhibitor proteins, in particular the cyclin-
66 dependent kinase inhibitor p27 (Hengst et al., 1994; Polyak et al., 1994). Interestingly, an over-
67 expression of this protein has been observed following cell-cell contact in three-dimensional
68 cultures, as compared to monolayers (St Croix et al., 1998, 1996; Xing et al., 2005). The adhesive
69 interaction between cells inside tumor spheroids leads to upregulation of p27, which results in cell
70 arrest in a quiescent phase of the cycle. Recently, the expression of p27 has been investigated
71 through a series of experiments involving mechanical compression of three-dimensional cell
72 aggregates (Delarue et al., 2014). Results show that a controlled compressive stress on tumor
73 spheroids inhibits cell proliferation by an over-expression of p27, blocking the cancerous cells at the
74 restriction point of the cell cycle.

75 At the beginning of this introduction, we have observed that the presence of a non-proliferating
76 cellular fraction has important consequences on the therapeutic efficacy of different
77 chemotherapeutic agents. Notably, previous works have shown that a reduction in p27 expression
78 in tumor spheroids could lead to better outcomes in terms of drug performance (St Croix et al.,
79 1998, 1996; Xing et al., 2005). However, experiments quantifying the influence of mechanical stress
80 on drug efficacy have still to be performed. Note that, interestingly, the compressive stresses that
81 can be induced in tumor spheroids are of the same order of magnitude of those measured *in vivo*
82 (Butcher et al., 2009; Fernández-Sánchez et al., 2015; Stylianopoulos et al., 2012), in the range of a
83 few kPa.

84 Phenomena concerning the mechanisms of drug action, as well as the mechanical characterization
85 of the state of a tissue, are difficult to investigate from a pure biological and biochemical framework.
86 To this end, mathematical models provide a valuable tool for establishing which of the biophysical
87 features of the tumor and the stroma are responsible for the observed behaviors. In the last years,
88 several review papers discussing different approaches to cancer modeling have been published
89 (Altrock et al., 2015; Byrne, 2010; Preziosi and Tosin, 2009; Sciumè et al., 2013). Some models
90 describe the action of a therapeutic agent on tumor spheroids (see for example (Frieboes et al.,
91 2009; Goodman et al., 2008; Ward and King, 2003)), whereas others take into account *in vivo*

settings, as in (Hossain et al., 2012; Kim et al., 2013; Mpekris et al., 2015). There are also models addressing the effects of mechanical stress on tumor development, such as those in (Kim et al., 2011; Loessner et al., 2013; Stylianopoulos et al., 2012). However, to the authors' knowledge, there is a lack of mathematical models focusing on the interactions between anticancer agents and the mechanical environment surrounding the tumor.

The aim of this work is to develop a theoretical framework that is able to take into account these interactions, providing new insights into mechanics-mediated drug resistance. In the following, we specialize our study to tumor spheroids. We address the effects of a chemotherapeutic agent, supposed to target cell proliferation, on these cell aggregates. Then, we evaluate the influence of mechanical compression on treatment efficacy.

The remainder of this work is organized as follows. Section 2 describes the mathematical model; the governing equations are presented, together with the assumed constitutive relations and parameter values. In Section 3 we report the results of the model. We start from the effects of different drug concentrations on the spheroid growth curve. Then, we consider a range of mechanical pressures acting on the spheroid surface and investigate their interactions with the treatment. Finally, we test different mathematical expressions for the drug-induced cell death term. Section 4, at the end, presents some concluding remarks.

2. Mathematical model

2.1 Governing equations

We build upon the mathematical model in (Mascheroni et al., 2016) to describe the transport of chemotherapeutic agents within an avascular tumor. The tumor is modeled as a biphasic porous material, and the governing equations are derived from porous media theory. We denote by t the solid phase of the porous medium, constituted by tumor cells (TCs) and ECM. The interstitial fluid (IF) constitutes the fluid phase (ℓ), which permeates the pores of the cellular scaffold. In our description, TCs are divided into living (Lt) and necrotic (Nt) fractions. In addition, we assume that the IF carries a nutrient, namely oxygen (ox), and a drug (ch). We consider a saturated material, where the IF fills all the voids of the porous medium. This results in the saturation constraint:

$$\varepsilon^t + \varepsilon^\ell = 1 \quad (1)$$

121 where ε^α denotes the volume fraction of phase α ($\alpha = t, \ell$). The mass balance equations for the
 122 phases in the biphasic system are given by:

$$123 \quad \frac{\partial(\varepsilon^t \rho^t)}{\partial t} + \text{div}(\varepsilon^t \rho^t \mathbf{v}^t) = \overset{\ell \rightarrow t}{M}_g - \overset{t \rightarrow \ell}{M}_d \quad (2)$$

$$124 \quad \frac{\partial(\varepsilon^\ell \rho^\ell)}{\partial t} + \text{div}(\varepsilon^\ell \rho^\ell \mathbf{v}^\ell) = -\overset{\ell \rightarrow t}{M}_g + \overset{t \rightarrow \ell}{M}_d \quad (3)$$

125 where ρ^α is the true mass density and \mathbf{v}^α the velocity of the α phase ($\alpha = t, \ell$). Here $\overset{\ell \rightarrow t}{M}_g$ is the
 126 term responsible for mass exchange between IF and TCs, dependent on cell proliferation; $\overset{t \rightarrow \ell}{M}_d$
 127 represents instead mass exchange between TCs and IF resulting from cell death and their following
 128 degradation. Oxygen and drug are described as species dissolved into the IF, and their mass balance
 129 reads:

$$130 \quad \frac{\partial(\varepsilon^\ell \rho^\ell \omega^{ox})}{\partial t} + \text{div}(\varepsilon^\ell \rho^\ell \omega^{ox} \mathbf{v}^\ell) - \text{div}[\varepsilon^\ell \rho^\ell D^{ox} \text{grad}(\omega^{ox})] = -\overset{ox \rightarrow t}{M}_{ox} \quad (4)$$

$$131 \quad \frac{\partial(\varepsilon^\ell \rho^\ell \omega^{ch})}{\partial t} + \text{div}(\varepsilon^\ell \rho^\ell \omega^{ch} \mathbf{v}^\ell) - \text{div}[\varepsilon^\ell \rho^\ell D^{ch} \text{grad}(\omega^{ch})] = -\overset{ch \rightarrow t}{M}_{ch} \quad (5)$$

132 where ω^β denotes the mass fraction of species β and D^β is its diffusion coefficient ($\beta = ox, ch$).
 133 The terms $\overset{ox \rightarrow t}{M}_{ox}$ and $\overset{ch \rightarrow t}{M}_{ch}$ represent oxygen and drug uptake by TCs, respectively. We describe the
 134 evolution for living and necrotic TCs through the system:

$$135 \quad \frac{\partial(\varepsilon^t \rho^t \omega^{Lt})}{\partial t} + \text{div}(\varepsilon^t \rho^t \omega^{Lt} \mathbf{v}^t) = -\varepsilon^t r^{Nt} + \overset{\ell \rightarrow t}{M}_g \quad (6)$$

$$136 \quad \frac{\partial(\varepsilon^t \rho^t \omega^{Nt})}{\partial t} + \text{div}(\varepsilon^t \rho^t \omega^{Nt} \mathbf{v}^t) = \varepsilon^t r^{Nt} - \overset{t \rightarrow \ell}{M}_d \quad (7)$$

137 where we have denoted by ω^{Lt} and ω^{Nt} the mass fractions of living and necrotic cells, respectively.
 138 Here $\varepsilon^t r^{Nt}$ is an intra-phase mass exchange term, commonly denoted as reaction term, accounting
 139 for the transfer of TCs from living to necrotic. Note that, by summing (6) and (7) we obtain (2)
 140 assuming that:

$$\omega^{L_t} = 1 - \omega^{N_t} \quad (8)$$

142

143 Following porous media theory (Lewis and Schrefler, 1998; Pinder and Gray, 2008), the mechanical
 144 stress exerted on the solid phase is described through the effective stress tensor $\mathbf{t}_{\text{eff}}^t$ given by:

$$\mathbf{t}_{\text{eff}}^t = \mathbf{t}^t + \alpha_B p^\ell \mathbf{I} \quad (9)$$

146 where \mathbf{I} is the unit tensor, \mathbf{t}^t the total stress tensor, p^ℓ is the fluid pressure in the interstitial fluid
 147 and α_B is Biot's coefficient defined by:

$$\alpha_B = 1 - \frac{K}{K_T}, \quad (10)$$

149 with K bulk modulus of the unsaturated skeleton and K_T bulk modulus of the solid phase. Then,
 150 we can state the linear momentum balance law for the tissue as (Lewis and Schrefler, 1998):

$$\text{div} \mathbf{t}^t = \text{div} (\mathbf{t}_{\text{eff}}^t - \alpha_B p^\ell \mathbf{I}) = 0 \quad (11)$$

152 Note that in (9) the tensile components of the stress tensors \mathbf{t}^t and $\mathbf{t}_{\text{eff}}^t$ are assumed positive.

153

154 2.2 Constitutive relations

155 In (Mascheroni et al., 2016), constitutive relationships for the effective stress and the mass transfer
 156 terms have been formulated. In particular, we have assumed the following form for the effective
 157 stress:

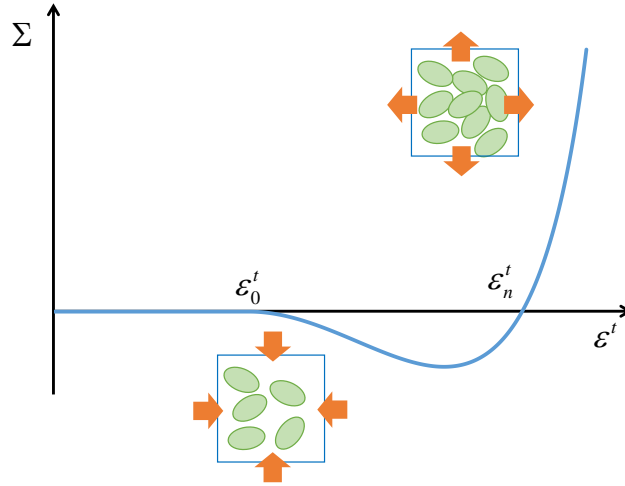
$$\mathbf{t}_{\text{eff}}^t = -\Sigma(\varepsilon^t) \mathbf{I} \quad (12)$$

159 with $\Sigma(\varepsilon^t)$ given by:

$$\Sigma(\varepsilon^t) = \begin{cases} \alpha (\varepsilon^t - \varepsilon_0^t)^2 \left[\frac{1 - \varepsilon_n^t}{(1 - \varepsilon^t)^\beta} - \frac{1}{(1 - \varepsilon^t)^{\beta-1}} \right], & \text{if } \varepsilon^t > \varepsilon_0^t \\ 0, & \text{otherwise} \end{cases} \quad (13)$$

161 This pseudo-potential law describes cells that do not interact if their volume fraction is below a given
 162 threshold (ε_0^t). Otherwise, they start to interact and develop attraction forces as long as their
 163 volume fraction is below a control value (ε_n^t). Finally, if TCs become too densely packed exhibiting a
 164 high volume fraction, they start to repel each other. This behavior is schematized in Figure 1. In the
 165 following we will denote by $\Sigma'(\varepsilon^t)$ the derivative of $\Sigma(\varepsilon^t)$ with respect to ε^t .

166



167

168 **Figure 1** Schematic for the stress function $\Sigma(\varepsilon^t)$, depicting the behavior of cells at different volume fractions.

169

170 The mass exchange terms in equation (2) represent TC growth and death, respectively. The first
 171 term describes cell proliferation and depends on the transfer of mass between the IF and the living
 172 fraction of the tumor. It takes the form:

173

$$174 \quad M_g^{\ell \rightarrow t} = \gamma_g^t \left\langle \frac{\omega^{ox} - \omega_{crit}^{ox}}{\omega_{env}^{ox} - \omega_{crit}^{ox}} \right\rangle_+ \left(1 - \delta_1 \frac{\langle \Sigma \rangle_+}{\langle \Sigma \rangle_+ + \delta_2} \right) \omega^{Lt} \varepsilon^t \quad (14)$$

175 here the coefficient γ_g^t accounts for oxygen uptake and for mass of IF that becomes tumor due to
 176 cell growth; ω_{crit}^{ox} is the critical mass fraction of oxygen, below which growth is inhibited, and ω_{env}^{ox} is
 177 the reference mass fraction of oxygen in the environment. The Macaulay brackets $\langle \cdot \rangle_+$ indicate the
 178 positive value of their argument: since the oxygen mass fraction ω^{ox} within the tumor can only be

179 equal to or smaller than ω_{env}^{ox} , the brackets will return a number between one ($\omega^{ox} = \omega_{env}^{ox}$) and zero
180 ($\omega^{ox} < \omega_{env}^{ox}$). The term in round squares describes growth inhibition by mechanical stress. The
181 constants δ_1 and δ_2 (with $\delta_1 < 1$) regulate the action of mechanical stress on cell proliferation and,
182 together with the term $\langle \Sigma \rangle_+$, model the inhibitory effect of compression on tumor cells proliferation
183 (Cheng et al., 2009; Helmlinger et al., 1997; Montel et al., 2012).

184 The rate of TC death in equation (2) is given by:

$$185 \quad \overset{t \rightarrow \ell}{M}_d = \overset{t \rightarrow \ell}{M}_{d,ly} + \overset{t \rightarrow \ell}{M}_{d,ch} \quad (15)$$

186 where the two contributions are related to cell lysis and drug action. In particular, the first term is
187 given by:

$$188 \quad \overset{t \rightarrow \ell}{M}_{d,ly} = \lambda_{\ell}^t \omega^{Nt} \varepsilon^t \quad (16)$$

189 where the coefficient λ_{ℓ}^t takes into account cellular degradation and mass conversion of necrotic
190 cells into IF. The second term takes the form:

$$191 \quad \overset{t \rightarrow \ell}{M}_{d,ch} = f_{ch} \lambda_{ch}^t \omega^{ch} \omega^{Lt} \varepsilon^t \quad (17)$$

192 Here λ_{ch}^t accounts for the rate of drug-induced cell death. The function f_{ch} is related to the
193 mechanism of action of the drug that is considered. Since we are interested in drugs that target TC
194 proliferation, we assume f_{ch} to depend on the growth term in (14):

$$195 \quad f_{ch}(\omega^{ox}, \Sigma) = \frac{M_g^{\ell \rightarrow t}}{\max(M_g^{\ell \rightarrow t})} = \left\langle \frac{\omega^{ox} - \omega_{crit}^{ox}}{\omega_{env}^{ox} - \omega_{crit}^{ox}} \right\rangle_+ \left(1 - \delta_1 \frac{\langle \Sigma \rangle_+}{\langle \Sigma \rangle_+ + \delta_2} \right) \quad (18)$$

196 where we highlight the dependence of f_{ch} on both the nutrient mass fraction ω^{ox} and the
197 mechanical stress Σ . In this way, the drug is most effective on the TCs that are well nourished and
198 not compressed. Note that, depending on the particular drug that is considered, different choices
199 for f_{ch} are possible (for example, in this framework it is possible to simulate drugs targeting hypoxia
200 or specific cellular species in the tumor).

201 The rate of necrosis of living tumor cells in equation (6) is described by:

$$\varepsilon^t \Gamma^{Nt} = \gamma_n^t \left\langle \frac{\omega_{crit}^{ox} - \omega^{ox}}{\omega_{env}^{ox} - \omega_{crit}^{ox}} \right\rangle_+ \omega^{Lt} \varepsilon^t \quad (19)$$

where the parameter γ_n^t regulates the rate of cell necrosis. The terms in the Macaulay brackets represent cell death by lack of nutrients.

During growth, TCs consume nutrients from the IF, a process that is described by the mass exchange term in equation (4):

$$\mathbf{M}_{ox}^{ox \rightarrow t} = \gamma_0^t \frac{\omega^{ox}}{\omega^{ox} + c^{ox}} \omega^{Lt} \varepsilon^t \quad (20)$$

This expression accounts for the dependence of oxygen consumption on its local level in the tumor. The coefficients γ_0^t and c^{ox} represent the order of magnitude of oxygen uptake and the oxygen mass fraction at which consumption is reduced by half, respectively.

Finally, the mass transfer term related to drug uptake in equation (5) takes the form:

$$\mathbf{M}_{ch}^{ch \rightarrow t} = \gamma_{ch}^t \omega^{ch} \omega^{Lt} \varepsilon^t \quad (21)$$

where we assumed the simplest kinetics for drug uptake (i.e. linear), with γ_{ch}^t accounting for the drug uptake rate by living TCs (Frieboes et al., 2009; Weinberg et al., 2007).

2.3 Model specialization to tumor spheroids

The equations of the model can be specialized to the case of tumor spheroids, following a similar procedure to that in (Mascheroni et al., 2016). The resulting system for the TC volume fraction, necrotic mass fraction, and oxygen and drug mass fractions can be summarized as:

$$\frac{\partial \varepsilon^t}{\partial t} - \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \varepsilon^t \frac{k}{\mu^\ell} \Sigma' \frac{\partial \varepsilon^t}{\partial r} \right) - \frac{1}{\rho} \left(\mathbf{M}_g^{\ell \rightarrow t} - \mathbf{M}_d^{t \rightarrow \ell} \right) = 0 \quad (22)$$

$$\frac{\partial (\omega^{Nt} \varepsilon^t)}{\partial t} - \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \varepsilon^t \omega^{Nt} \frac{k}{\mu^\ell} \Sigma' \frac{\partial \varepsilon^t}{\partial r} \right) - \frac{1}{\rho} \left(\varepsilon^t r^{Nt} - \mathbf{M}_{d,ly}^{t \rightarrow \ell} \right) = 0 \quad (23)$$

$$\frac{\partial [(1 - \varepsilon^t) \omega^{ox}]}{\partial t} + \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \varepsilon^t \omega^{ox} \frac{k}{\mu^\ell} \Sigma' \frac{\partial \varepsilon^t}{\partial r} \right) - \frac{1}{r^2} \frac{\partial}{\partial r} \left[r^2 (1 - \varepsilon^t) D^{ox} \frac{\partial \omega^{ox}}{\partial r} \right] + \frac{1}{\rho} \mathbf{M}_{ox}^{ox \rightarrow t} = 0 \quad (24)$$

$$\frac{\partial \left[(1 - \varepsilon^t) \omega^{ch} \right]}{\partial t} + \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 \varepsilon^t \omega^{ch} \frac{k}{\mu^\ell} \Sigma' \frac{\partial \varepsilon^t}{\partial r} \right) - \frac{1}{r^2} \frac{\partial}{\partial r} \left[r^2 (1 - \varepsilon^t) D^{ch} \frac{\partial \omega^{ch}}{\partial r} \right] + \frac{1}{\rho} \frac{M^{ch \rightarrow t}}{ch} = 0 \quad (25)$$

We have adopted spherical symmetry, and r is the radial coordinate over the spheroid radius. The parameters k and μ^ℓ are the intrinsic permeability of the cellular scaffold and the dynamic viscosity of IF, respectively. They arise by assuming Darcy's law for the relative velocity of the two phases (Mascheroni et al., 2016; Sciumè et al., 2013). Moreover, we take the phases to be incompressible and assign a common value for their densities, which we denote by the constant ρ . Note that this leads to $\alpha_B = 1$. Then, we model the growth of the spheroid as a free-boundary problem, where the interface constituted by TCs is a material surface for the TCs that moves with velocity v^t , given by:

$$\frac{dR}{dt} = v^t = - \frac{k}{\mu^\ell} \Sigma' \frac{\partial \varepsilon^t}{\partial r} \Big|_{r=R} \quad (26)$$

where R is the external radius of the spheroid. The closed form of the differential problem is then obtained by defining a proper set of boundary and initial conditions. In particular, regularity at the spheroid center requires:

$$\frac{\partial \varepsilon^t}{\partial r} = \frac{\partial \omega^{Nt}}{\partial r} = \frac{\partial \omega^{ox}}{\partial r} = \frac{\partial \omega^{ch}}{\partial r} = 0, \quad \text{in } r = 0, \quad (27)$$

while we enforce Dirichlet boundary conditions on the tumor external surface:

$$\varepsilon^t = \varepsilon_{ext}^t, \quad \omega^{Nt} = 0, \quad \omega^{ox} = \omega_{env}^{ox}, \quad \omega^{ch} = \omega_{env}^{ch}, \quad \text{in } r = R. \quad (28)$$

Finally, we assume the following initial conditions over the spheroid radius:

$$\varepsilon^t = \varepsilon_{ext}^t, \quad \omega^{Nt} = 0, \quad \omega^{ox} = \omega_{env}^{ox}, \quad \omega^{ch} = 0, \quad \text{on } 0 < r < R \text{ at } t = 0. \quad (29)$$

2.4 Model parameters

The parameters used in the model are listed in Table 1. Some of the values are taken from (Mascheroni et al., 2016), where the model results are compared to experimental data. In this work, we need to add the values for the parameters appearing in the equations governing drug transport and uptake. For these quantities we assume the values in (Frieboes et al., 2009), obtained for spheroids treated with Doxorubicin. Actually, the parameter governing drug-induced cell death,

247 λ_{ch}^t , depends on the particular therapeutic agent and cell line that are considered. Here it is selected
 248 to produce a reasonable response of the model when the spheroids are subjected to the given drug
 249 concentrations. Note that, as it will be shown in Section 3.3, model results will not be significantly
 250 affected by this choice.

251

252 **Table 1** Parameters used in the model.

Parameter	Value	Unit	Reference
ω_{env}^{ox}	7.7×10^{-6}	(-)	(Mueller-Klieser et al., 1986; Mueller-Klieser and Sutherland, 1982)
c^{ox}	1.48×10^{-7}	(-)	(Casciari et al., 1992a, 1992b)
γ_0^t	3.0×10^{-4}	$\text{kg}/(\text{m}^3 \cdot \text{s})$	(Casciari et al., 1992a, 1992b)
β	0.5	(-)	(Byrne and Preziosi, 2003)
ε_n^t	0.8	(-)	(Byrne and Preziosi, 2003)
ε_0^t	1/3	(-)	(Byrne and Preziosi, 2003)
k	1.8×10^{-15}	m^2	(Netti et al., 2000)
μ^ℓ	1.0×10^{-3}	$\text{Pa} \cdot \text{s}$	(Sciumè et al., 2013b)
D^{ox}	3.2×10^{-9}	m^2/s	(Sciumè et al., 2013b)
ρ	1.0×10^3	kg/m^3	(Sciumè et al., 2013b)
ω_{crit}^{ox}	2.0×10^{-6}	(-)	(Mascheroni et al., 2016)
γ_g^t	5.4×10^{-3}	$\text{kg}/(\text{m}^3 \cdot \text{s})$	(Mascheroni et al., 2016)
γ_n^t	1.5×10^{-1}	$\text{kg}/(\text{m}^3 \cdot \text{s})$	(Mascheroni et al., 2016)
λ_ℓ^t	1.15×10^{-2}	$\text{kg}/(\text{m}^3 \cdot \text{s})$	(Mascheroni et al., 2016)
α	1.0×10^5	Pa	(Mascheroni et al., 2016)
ω_{env}^{ch}	$8.696 \div 271.76 \times 10^{-9}$	(-)	(Frieboes et al., 2009)
D^{ch}	9.375×10^{-14}	m^2/s	(Frieboes et al., 2009)
γ_{ch}^t	1.157×10^{-2}	$\text{kg}/(\text{m}^3 \cdot \text{s})$	(Frieboes et al., 2009)
λ_{ch}^t	5.0×10^4	$\text{kg}/(\text{m}^3 \cdot \text{s})$	(-)

253

254

3. Results

3.1 Tumor spheroid growth in the presence of a drug

In this section we test the effects of a drug that targets cell proliferation in a three-dimensional cell aggregate. We consider first tumor spheroids that grow suspended in culture medium, subject to different drug concentrations. We assume drug concentration at spheroid boundary to start from zero and, following a ramp, to reach the final value ω_{env}^{ch} after 3h.

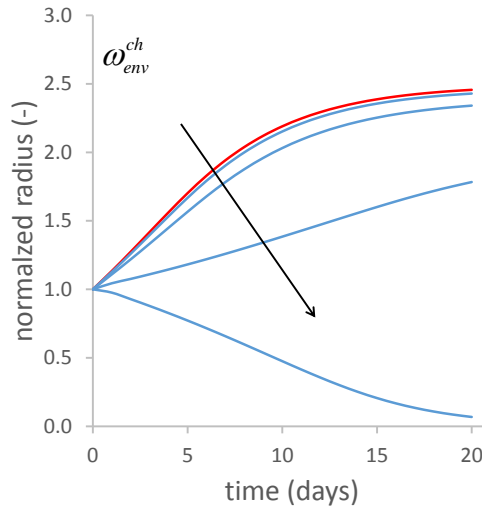


Figure 2 Effect of different drug concentrations on spheroid growth. The red line refers to a spheroid grown in the absence of drug. The other lines are for $\omega_{env}^{ch} = 0.086, 0.347, 1.391, 2.717 \times 10^{-7}$.

In Figure 2, we show the evolution of the spheroid radius over time for different drug mass fractions (i.e. $\omega_{env}^{ch} = 0.086, 0.347, 1.391, 2.717 \times 10^{-7}$). Here, the arrow points in the direction of increasing ω_{env}^{ch} . We consider the normalized value of the spheroid radius, namely the ratio between the present value of the radius and the initial radius of the spheroid (200 μm in this case). The red line represents a spheroid grown in the absence of drug. We can distinguish between the first stages of growth, displaying an exponential/linear behavior, followed by a phase of growth saturation where the radius tends to a steady value. Low concentrations of drug do not alter the shape of the growth curve, whereas for high levels of the chemotherapeutic agent the spheroid starts to shrink and, for the highest value of ω_{env}^{ch} , growth is almost completely inhibited. This behavior closely resembles the growth curves obtained for example in (Kim et al., 2010; Mikhail et al., 2013), where spheroids from various cell lines are subject to different drugs.

Figure 3 shows the drug mass fraction inside the spheroid for an intermediate value of ω_{env}^{ch} , at the end of the simulation. Note the steep gradient of drug appearing from the boundary towards the center of the cell aggregate. In this case, the therapeutic agent can exert its effect only over the outermost region of the spheroid. This phenomenon arises as a consequence of poor diffusion of the drug molecules inside the spheroid and due to drug uptake by proliferating TCs. Interestingly, similar results are obtained in the experimental literature (see for example (Gong et al., 2015; Wang et al., 2013)), analyzing the penetration of free drug into a spheroid.

Then, we look for the value of drug mass fraction that is able to provide a reduction of 50% in spheroid volume (usually identified with the label IC_{50} , for “half maximal inhibitory concentration” (Curtis et al., 2016)). We find a value of $\omega_{env}^{ch} = 1.185 \times 10^{-7}$, which we will denote from now on with IC_{50} . The growth curve relative to this drug mass fraction is shown in Figure 4.a, where we report the evolution of the normalized volume (i.e. the ratio between the spheroid volume and its initial volume) over time. The evolution of oxygen mass fraction over the spheroid radius is represented in Figure 4.b. Note the steep oxygen gradients at later times of the simulation, from the spheroid boundary towards its interior. The necrotic mass fraction of TCs is displayed in Figure 4.c. A necrotic population appears after a few days from the beginning of the simulation and gives rise to a necrotic core at later days. Both Figures 4.b and 4.c refer to a spheroid not treated with the drug, whereas the second row of Figures (4.d-f) pertains to a spheroid grown in the presence of a drug with a mass fraction equal to IC_{50} . The drug mass fraction over the spheroid radius is presented in Figure 4.d. Note that, after a few days from the beginning of the simulation, the therapeutic agent is mainly distributed over the spheroid periphery. Figure 4.e shows the oxygen mass fraction in the drug-treated spheroid. We can observe a behavior similar to the one in Figure 4.b, but this time over a smaller spheroid. Finally, the necrotic mass fraction in a spheroid subjected to the drug is shown in Figure 4.f. Compared to Figure 4.c, here the necrotic core is less extended and appears at later times in the simulation. This may be due to a smaller mass fraction of LTCs that can undergo necrosis, deriving from LTC killing by the chemotherapeutic agent.

$$\omega_{env}^{ch} = 3.478 \times 10^{-8}, \quad t = 20 \text{ d}$$

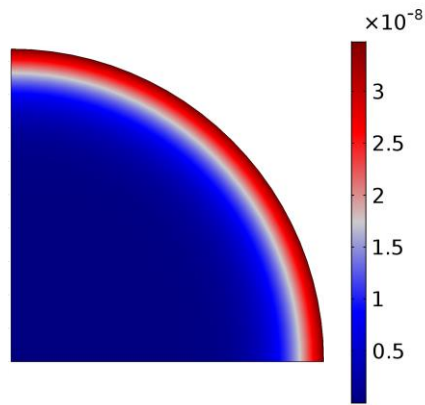


Figure 3 Drug mass fraction inside the spheroid at day 20 and for $\omega_{env}^{ch} = 3.478 \times 10^{-8}$.

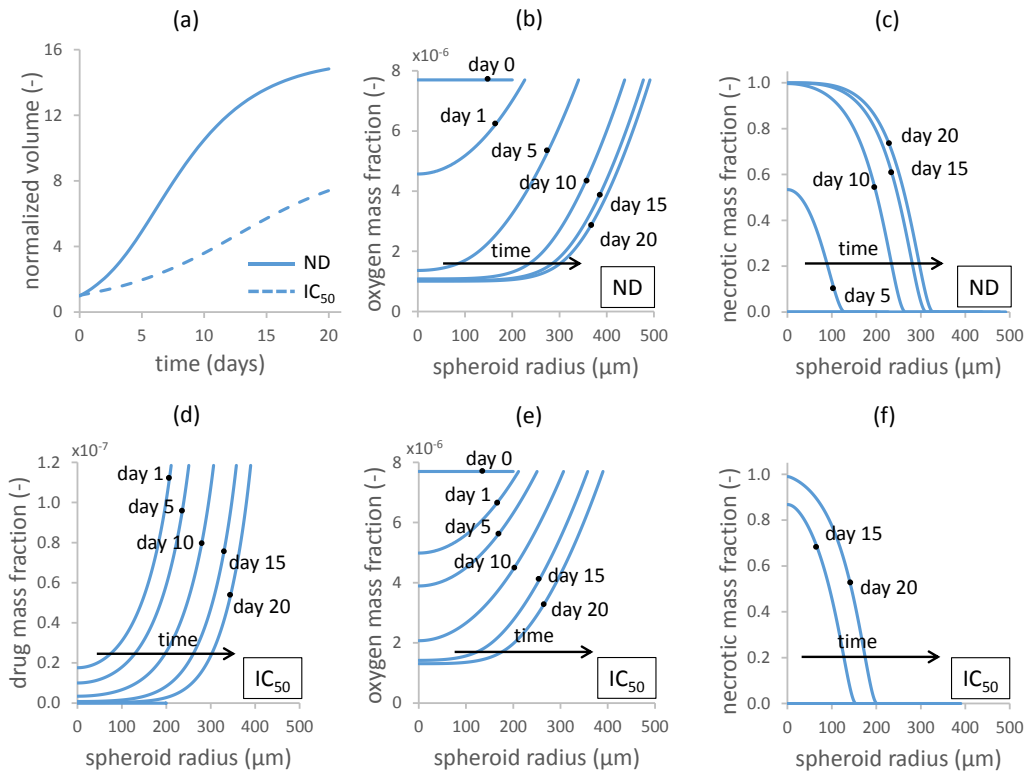


Figure 4 Comparison between a spheroid grown without an external drug (ND) and one treated with a drug mass fraction equal to IC_{50} .

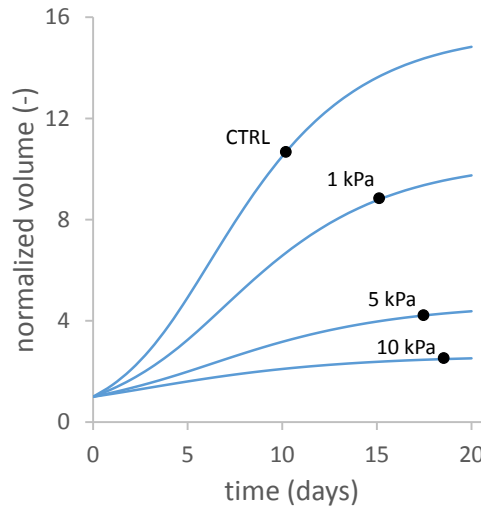


Figure 5 Normalized volumes of spheroids grown under different external mechanical pressures.

3.2 Effect of mechanical compression on drug efficacy

In (Mascheroni et al., 2016), we investigated the effects of an external mechanical pressure on the growth curves of tumor spheroids. Figure 5 report these previous findings, in terms of the evolution of the normalized volumes of spheroids subjected to different compressive stresses. We consider four compression levels, ranging from 1kPa to 10kPa. The growth of the most compressed spheroid shows a 7-fold reduction when compared to the control case (grown in the absence of an external stress). Note that the inhibitory effect of compressive stresses is included in the equations through the constitutive relation in (14). We make use of these results to test our newly introduced framework for drug transport and uptake in the spheroid. In particular, we apply the same external mass fraction of drug (IC_{50}) to each of the compression tests. Then, we check for variations in spheroid volumes with respect to the case with no drug added to the culture medium (Figure 6 and 7). Figure 6 compares the normalized volumes of spheroids undergoing different compressive stresses. We test spheroids in the absence (ND) or presence (IC_{50}) of a chemotherapeutic drug. Both the series, ND and IC_{50} , exhibit the same decreasing trend, although with a slower volume reduction for drug-treated spheroids. The variation between the two volumes for each compressive condition is shown in Figure 7. According to the definition of IC_{50} , the control case displays a 50% reduction in volume. Interestingly, the series exhibit a percentage variation decreasing with the extent of mechanical compression, as highlighted by the black arrow. The case undergoing maximum compression shows a reduction of about 30% in volume reduction. The observed behavior arises as a consequence of a lower proliferation index within the spheroid. In fact, mechanical stress inhibits

336 cell proliferation via equation (14) of the model, providing smaller values for the growth term as
337 compression increases. Since our discussion is based on drugs that target cell duplication, growth
338 inhibition is responsible for a cell population over which the therapeutic agent is less effective. Note
339 that this effect could be relevant for *in vivo* applications: a drug concentration that is known to be
340 effective in a particular regime (such as 3D cultures) could not provide the same results when the
341 tumor is subjected to mechanical compression. Moreover, since several drug screenings are
342 evaluated on monolayer cultures, such effects arising from a full three-dimensional setting may be
343 overlooked (Friedrich et al., 2009).

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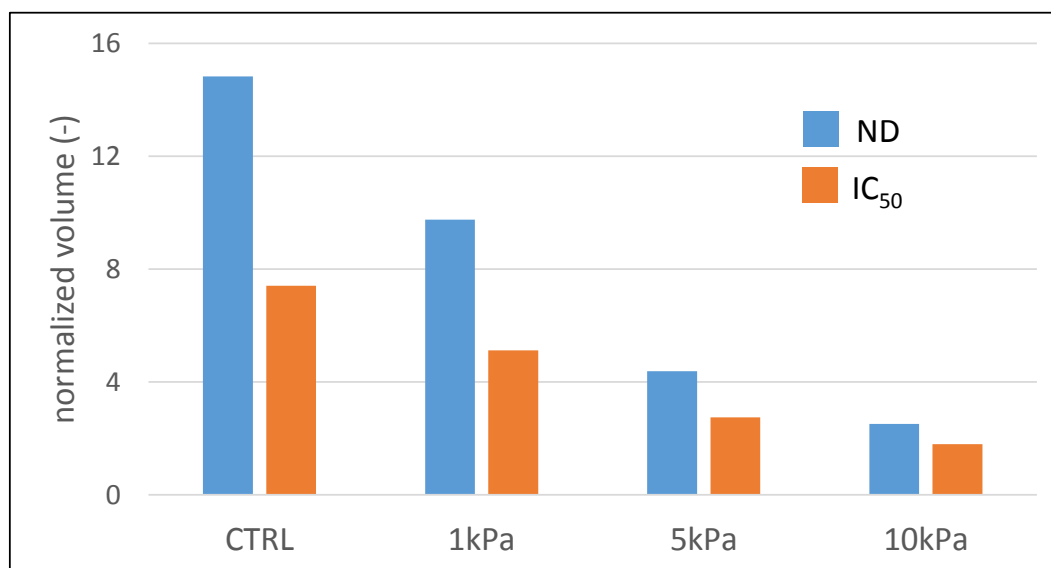


Figure 6 Comparison of the normalized volumes of spheroids subject to different mechanical stress, grown in the absence of drug (ND) or subject to a drug concentration of IC₅₀.

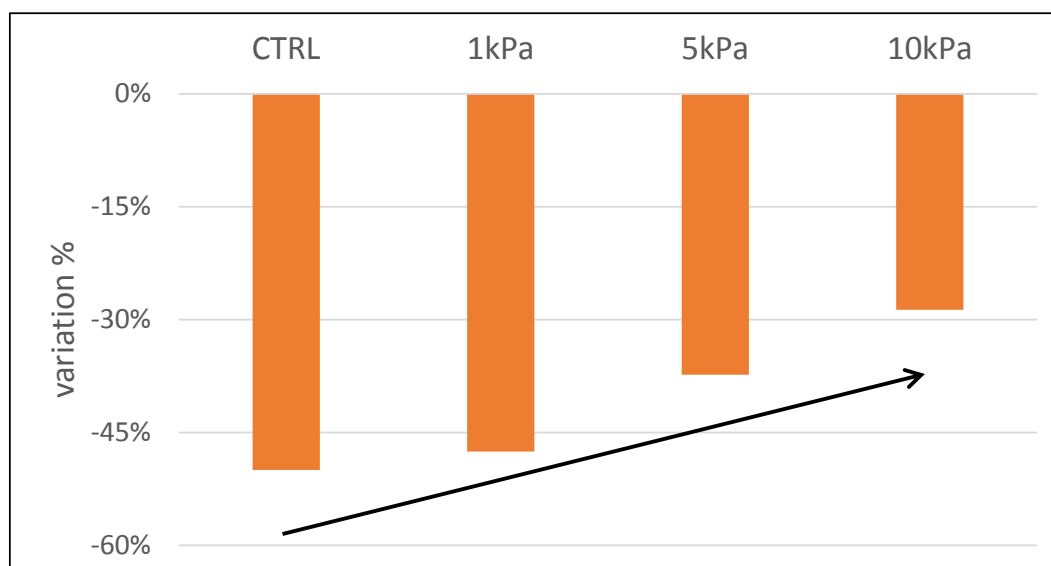


Figure 7 Variation in spheroid volume due to the action of a chemotherapeutic drug at different external mechanical pressures. As shown by the arrow, the efficacy of proliferation targeting drugs is less effective for higher tumor compressions, because of stress inhibition of growth.

358 3.3 Response of the model for different forms of the drug-induced death term

359 To confirm that model results are not biased by the particular choice of the term in (17), we test
 360 different mathematical expressions accounting for drug-induced cell death. The simplest
 361 hypothesis, assumed in (17), considers cell death to be proportional to the local amount of drug. In
 362 the following, we will refer to this case as the “linear” one. We introduce two additional
 363 relationships, given by:

$$364 \quad \overset{t \rightarrow \ell}{M}_{d, ch} = f_{ch} \frac{m_1 \omega^{ch}}{\omega^{ch} + m_2} \omega^{L_t} \varepsilon^t \quad (30)$$

$$365 \quad \overset{t \rightarrow \ell}{M}_{d, ch} = f_{ch} p_1 \left(\omega^{ch} \right)^{p_2} \omega^{L_t} \varepsilon^t \quad (31)$$

366 In (30), we assume a dependence of the Michaelis-Menten type; in (31) the assumed relationship
 367 takes the form of a power law. Note that, as the functional dependence on the local drug
 368 concentration changes, these relations give rise to new values for the inhibitory concentration IC_{50} .
 369 We report the new IC_{50} and the values for the parameters that characterize the above expressions
 370 in Table 2. Once the new forms for the drug-induced cell death term are implemented into the
 371 model, we perform the same numerical tests of the previous section to analyze the coupled effect
 372 of drug action and mechanical compression. In Figure 8, we report the variation in terms of spheroid
 373 volume induced by the drug for different compressive stresses. Like in the previous Figure, the first
 374 series of data serves as a control and indicates a variation of 50% with respect to the drug-free
 375 condition. The other series are related to the different compression regimes and compare the model
 376 response for the different mathematical relationships assumed for the death term. It is possible to
 377 observe that the variation in volume reduction is similar to the linear case, analyzed in the previous
 378 section. The effect of mechanical compression on drug efficacy described previously does not seem
 379 therefore to be originated from the particular mathematical form adopted for the death term.

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Table 2 Parameter values for the relations assumed in the cell death term.

Relation	Parameter	Value	Unit	IC ₅₀
Linear	λ_{ch}^t	5.0×10^4	kg / (m ³ · s)	1.185×10^{-7}
Michaelis-Menten	m_1	1.5×10^{-2}	kg / (m ³ · s)	5.345×10^{-8}
	m_2	1.0×10^{-7}	(-)	
Power law	p_1	2.5×10^{11}	kg / (m ³ · s)	1.862×10^{-7}
	p_2	2	(-)	

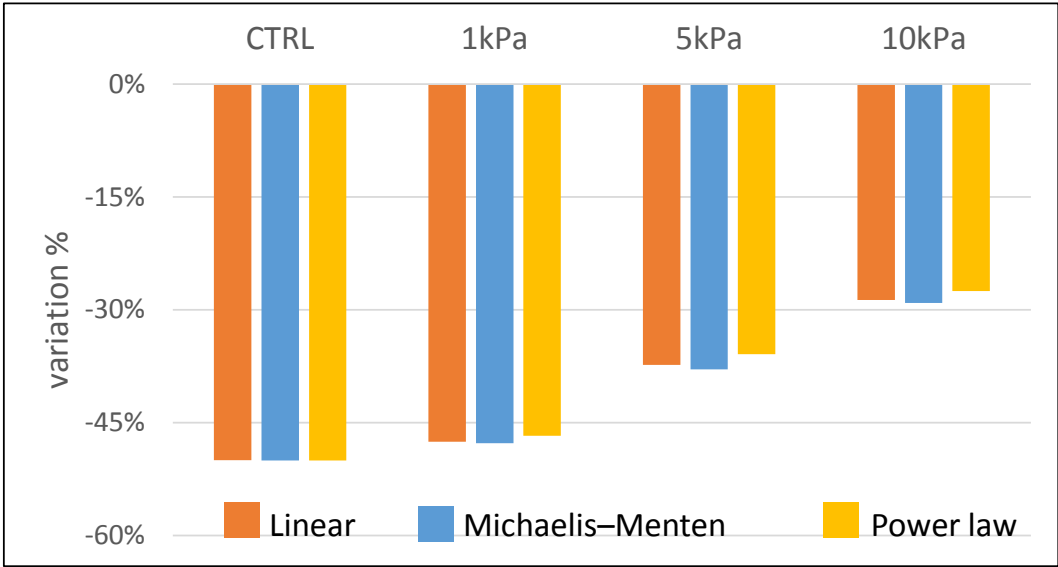


Figure 8 Effect of different mathematical relations on spheroid volume variation.

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394 4. Conclusion

395 In this work, we introduce equations for drug transport and uptake by TCs in our previous
396 mathematical model for avascular tumor growth. Then, we adapt the equations for the tumor
397 spheroid case and test the effects of a proliferation targeting drug on spheroid growth curves. We
398 observe a qualitative agreement between model results and experimental literature (Gong et al.,
399 2015; Kim et al., 2010; Mikhail et al., 2013; Wang et al., 2013). Then, we simulate tumor spheroids
400 undergoing mechanical compressive stresses of different amplitudes and consider their volume
401 reduction due to the presence of a therapeutic agent. Interestingly, we notice a decreased growth
402 inhibition efficacy of the drug in terms of the final volumes reached by the spheroids, arising because
403 of compressive stresses. Finally, we test three different mathematical expressions for the cell death
404 term induced by the therapeutic agent. The resulting predictions are similar for all the tested
405 relations, suggesting that the particular form of the adopted constitutive relation does not influence
406 model response. Taken together, these results suggest that mechanical compression of tumor
407 spheroids may compromise the efficacy of a chemotherapeutic agent targeting cell proliferation.

408 As several simplifying assumptions are considered in the work, the model is certainly open to further
409 improvements. In particular, here we model only one nutrient species, i.e. oxygen, diffusing in the
410 interstitial fluid and regulating TC proliferation. Even though the presence of other chemicals is
411 implicitly contained within the mass exchange term in (14), future inclusion of additional nutrients,
412 growth and necrosis factors could provide a more detailed description of the tumor system
413 (Chauhan and Jain, 2013; Jain et al., 2014). Moreover, since the particular physicochemical
414 environment in which the tumor is embedded affects significantly the outcomes of therapies (see
415 for example (Luk et al., 1990), (Seebacher et al., 2015)), proper consideration of these factors would
416 result in a better description of drug dynamics. Notably, some experiments couple therapeutic
417 agents to nanoparticle formulations, enabling a larger penetration into the tumor (Kim et al., 2010;
418 Wang et al., 2013). Note that this latter kind of results can be integrated in the current model, once
419 suitable mechanisms for nanoparticle delivery are hypothesized. Another point requiring some
420 attention is the proper choice of constitutive relations. As it happens frequently in literature, most
421 of these laws are derived from phenomenological arguments. More experimental work is needed
422 to link the mathematical form assigned to the various terms to the underlying biology. This kind of
423 reasoning should be applied to the constitutive relations accounting for the drug uptake and the
424 following effects on TCs, as well as the mechanical description of the tumor ensemble. For the latter,

here we consider a simple law, linking the stress in the tissue to the local volume fraction of tumor cells. This assumption provides a great simplification of the equations and is shown to give a good description of experimental observations (Mascheroni et al., 2016). However, it neglects several phenomena related to the mechanical behavior of a biological tissue. For example, viscoplastic effects existing at smaller timescales than those of cell proliferation are not taken into account (Forgacs et al., 1998; Giverso and Preziosi, 2012). Also, breaking and formation of cellular bonds during tumor development should be included to give a more complete description (Ambrosi et al., 2012; Preziosi et al., 2010). Finally, we highlight the need for experiments addressing the interactions between therapeutic agents and tumor mechanical environment. These experiments will serve to calibrate the parameters in the equations and to test model results. Part of future experimental work should also be devoted to the biochemical understanding of the growth inhibition process following mechanical stress. Although some work is already present in the literature (Cheng et al., 2009; Delarue et al., 2014; Loessner et al., 2013), several details remain to be elucidated. New investigations analyzing the interactions between the tumor and its bio-mechanical environment should allow for a better understanding of disease progression, with the final goal of aiding the design of effective therapeutic treatments.

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442 **Acknowledgements**

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445 **References**

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