

# MATHEMATICAL ANALYSIS OF A SIMPLE MODEL FOR THE GROWTH OF NECROTIC TUMORS IN PRESENCE OF INHIBITORS

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**Abstract:** We study a mathematical model for the growth of tumors with two free boundaries: the inner boundary delaying the necrotic zone and the outer boundary delaying the tumor. We consider the presence of inhibitors and establish the existence and uniqueness of the solution for the model under suitable conditions on the inhibitors interaction and the tumor growth.

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**Key Words:** growth of tumors, free boundary problems, partial differential equations, uniqueness of solution

## 1. Introduction

The growth of a tumor is a complicated phenomenon which involves many different aspects, from the sub-cellular scale (gene mutation or secretion of substances) to the body scale (*metastasis*). In the behavior of the tumor cells there appear biological aspects such as *necrosis*

(death of cells caused by insufficient level of nutrients), *apoptosis* (natural cell death, it is an intrinsic property of the cell), *mitosis* (birth of cells by cell division), diffusion of nutrients and inhibitors and *vascularization* (contribution of nutrients through vessels). We study a simple mathematical model for this process. Previous similar models were considered by Greenspan [10], Byrne and Chaplain [4], Friedman and Reitich [9] and Cui and Friedman [5], [6]. The tumor comprises a central necrotic core, where the cells die as a result of necrosis, when the concentration of nutrients  $\hat{\sigma}$  (oxygen, glucose, etc.) falls below a critical level  $\sigma_n$ . Then there is an early disintegration of the cells into simpler chemical compounds (mainly water). These substances form a *necrotic* core in the center of the tumor. This necrotic core is covered by a layer, where apoptosis and mitosis occur. In the study of the internal mechanisms of the tumor growth two unknown free boundaries appear: the outer boundary denoted by  $R(t)$  (limiting the tumor) and the inner free boundary denoted by  $\rho(t)$  (separating the necrotic core of the remaining part).

We consider the presence of *Growth Inhibitor Factors* (GIFs) as chalone in the same spirit as the pioneering papers by Greenspan [10], [11]. As in any tissue, the cell proliferation is controlled by chemical substances (GIFs) secreted by the cells, which reduce the mitotic activity. Two different kinds of inhibitors appear, depending on the phase of the cell cycle stage at which inhibition has been shown. The inhibitor can act before DNA synthesis (as epidermal chalon in melanoma or granulocyte chalon in leukemia) or before mitosis (see Attallah [2]). The properties of these chemical inhibitors have been studied in several works (see e.g. Inversen [12], [13]).

The effectiveness of an anticancer drug delivered to the tumor can be compared to therapy designed to administer the drug by diffusion from neighboring tissue.

According to principle of conservation of mass, the tumor mass is proportional to its volume  $\frac{4}{3}\pi R^3(t)$ , assuming the density of the cell mass is constant. The balance between the birth and death rate of cells is given as a function of the concentration of nutrients and inhibitors. Let  $\hat{S}$  be this balance, then after normalizing we obtain the law

$$\frac{d}{dt}\left(\frac{4}{3}\pi R^3(t)\right) = \int_{\{\tilde{x} < R(t)\}} \hat{S}(\hat{\sigma}(\tilde{x}, t), \hat{\beta}(\tilde{x}, t)) d\tilde{x}. \quad (1.1)$$

Depending on the author, the function  $\hat{S}$  can be written in different

ways. Greenspan [10] studied the problem in the presence of an inhibitor, and the possibility that this affects mitosis, when the concentration of the inhibitor is greater than a critical level  $\tilde{\beta}$ . He proposed  $\widehat{S}(\widehat{\sigma}, \widehat{\beta}) = sH(\widehat{\sigma} - \tilde{\sigma})H(\tilde{\beta} - \widehat{\beta})$ , where  $H(\cdot)$  denotes the maximal monotone graph of  $\mathbb{R}^2$  associate with the Heaviside function, i.e.  $H(k) = 0$  if  $k < 0$ ,  $H(k) = 1$  if  $k > 0$  and  $H(0) = [0, 1]$ . Byrne and Chaplain [4] study the growth when the inhibitor affects the cell proliferation and propose  $\widehat{S}(\widehat{\sigma}, \widehat{\beta}) = s(\widehat{\sigma} - \tilde{\sigma})(\tilde{\beta} - \widehat{\beta})$  (for a positive constant  $s$ ). In the absence of inhibitors or in case that the inhibitor does not affect mitosis, they choose  $\widehat{S}(\widehat{\sigma}, \widehat{\beta}) = s\widehat{\sigma}(\widehat{\sigma} - \tilde{\sigma})$ . Friedman and Reitich [9] and Cui and Friedman [5] study the asymptotic behavior of the radius,  $R(t)$ , with the cell proliferation rate free of the action of inhibitors. They assume that  $\widehat{S} = s(\sigma - \tilde{\sigma})$ , where  $s\sigma$  is the cell birth-rate and the death-rate is given by  $s\tilde{\sigma}$ .

The transfer of nutrients to the tumor through the vasculature occurs below a certain level  $\sigma_B$ , and it is done with a rate  $r_1$ . During the development of the tumor, the immune system secretes inhibitors as a immune response to the foreign body. The structure of inhibitor absorption is similar to the transference of nutrients (for a constant  $r_2$ ). If we assume that the nutrient consumption rate is proportional to the concentrations of nutrients, the nutrient consumption rate is given by  $\lambda\widehat{\sigma}$ . Both processes, consumption and transference, occur simultaneously in the exterior of the necrotic core, where cells are inhibited by  $\widehat{\beta}$ . We assume that the host tissue is homogenous and that the diffusion coefficient,  $d_1$ , is constant. The reaction between nutrients and inhibitors can be globally modelled by introducing the Heaviside maximal monotone graph (as function of  $\widehat{\sigma}$ ) and some continuous functions  $g_i(\widehat{\sigma}, \widehat{\beta})$ . Then  $\widehat{\sigma}$  satisfies

$$\frac{\partial \widehat{\sigma}}{\partial t} - d_1 \Delta \widehat{\sigma} \in r_1((\sigma_B - \widehat{\sigma}) - \lambda\widehat{\sigma} - \widehat{\beta})H(\widehat{\sigma} - \sigma_n) + \widehat{g}_1(\widehat{\sigma}, \widehat{\beta}). \quad (1.2)$$

We also assume a constant diffusion coefficient for the inhibitor concentration  $\widehat{\beta}$ ,  $d_2$ . The model considers the permanent supply of inhibitors, modelled by  $\tilde{f}$  and localized on a small region  $\omega_0$  inside the tumor. This term  $\tilde{f}$  was introduced in Díaz and Tello [8] to control the growth of the tumor. Then  $\widehat{\beta}$  satisfies

$$\frac{\partial \widehat{\beta}}{\partial t} - d_2 \Delta \widehat{\beta} = -r_2 \widehat{\beta} H(\widehat{\sigma} - \sigma_n) + \widehat{g}_2(\widehat{\sigma}, \widehat{\beta}) + \tilde{f} \chi_{\omega_0}, \quad (1.3)$$

adding initial and boundary conditions we obtain

$$\widehat{\sigma}(\tilde{x}, t) = \overline{\sigma}, \quad \widehat{\beta}(\tilde{x}, t) = \overline{\beta}, \quad |\tilde{x}| = R(t), \quad (1.4)$$

$$\widehat{\sigma}(\tilde{x}, 0) = \sigma_0(\tilde{x}), \quad \widehat{\beta}(\tilde{x}, 0) = \beta_0(\tilde{x}), \quad |\tilde{x}| < R_0. \quad (1.5)$$

In this formulation, the presence of the maximal monotone graph  $H$  is the reason why the symbol  $\in$  appears in equations (1.2) and (1.3) instead of the equal sign (a precise notion of weak solution will be presented later). Different constants appears in the equations and boundary conditions which lead to a wide variety of special cases:  $\sigma_n$  is the level of concentration of nutrients above which the cells can live (below this level the cells die by *necrosis*),  $\overline{\sigma}$  and  $\overline{\beta}$  are the concentration of nutrients and inhibitors in the exterior of the tumor. The diffusion operator  $\Delta$  is the Laplacian operator and  $\chi_{\omega_0}$  denotes the characteristic function of the set  $\omega_0$  (i.e.  $\chi_{\omega_0}(\tilde{x}) = 1$ , if  $\tilde{x} \in \omega_0$ , and  $\chi_{\omega_0}(\tilde{x}) = 0$ , otherwise).

Notice that the above formulation is of global nature and that the inner free boundary  $\rho(t)$  is defined implicitly as the boundary of the set  $\{r \in [0, R(t)) : \widehat{\sigma} \leq \sigma_n\}$ . So, if for instance, the initial datum  $\sigma_0$  satisfies  $\sigma_0(\tilde{x}) = \sigma_n$  on  $[0, \rho_0]$ , for some  $\rho_0 > 0$  and  $\widehat{g}_1(\sigma_n, \widehat{\beta}) \in [0, r_1(\sigma_B - \sigma_n) - \lambda\sigma_n]$  for any  $\widehat{\beta} \geq 0$ , the above formulation leads to the associate double free boundary formulation in which  $\widehat{\sigma}$  satisfies

$$\left\{ \begin{array}{ll} \frac{\partial \widehat{\sigma}}{\partial t} - d_1 \Delta \widehat{\sigma} + \lambda \widehat{\sigma} = r_1(\sigma_B - \widehat{\sigma}) + \widehat{g}_1(\widehat{\sigma}, \widehat{\beta}), & \rho(t) < |\tilde{x}| < R(t), \\ \widehat{\sigma}(\tilde{x}, t) = \sigma_n, & |\tilde{x}| \leq \rho(t), \\ \widehat{\sigma}(\tilde{x}, t) = \overline{\sigma}, & |\tilde{x}| = R(t), \\ R(0) = R_0, \rho(0) = \rho_0, \widehat{\sigma}(\tilde{x}, 0) = \sigma_0(\tilde{x}), & \rho_0 < |\tilde{x}| < R_0. \end{array} \right.$$

The free boundary  $R(t)$  is described by the ODE presented in (1.1).

We prove the solvability of the model equations: (1.1)-(1.5) and establish uniqueness of solutions under additional conditions. The existence result is present in Section 2 and proved by using a Galerkin approximation based on a weak formulation of the problem.

We have mentioned that the study of the approximate controllability problem is considered in Díaz and Tello [8], where  $f$  is understood as a local control and the goal is to made the final nutrient concentration  $\widehat{\sigma}(\tilde{x}, T)$  as closed as desired (in a suitable sense) to a given profile  $\widehat{\sigma}_d(\tilde{x})$ .

## 2. Existence of Solutions

We shall assume that the reaction terms  $\widehat{g}_i$  and the mass of the tumor balance  $\widehat{S}$  satisfy:

$$\widehat{g}_i \text{ are piecewise continuous, } |\widehat{g}_i(a, b)| \leq c_0 + c_1(|a| + |b|), \quad (2.1)$$

$$\widehat{S} \text{ is continuous and } -\lambda_0 \leq \widehat{S}(a, b) \leq c_0 + c_1(|a|^2 + |b|^2), \quad (2.2)$$

for some positive constants  $\lambda_0$ ,  $c_0$  and  $c_1$ .

The above assumptions ((2.1) and (2.2)) do not constitute biological restrictions, and previous models satisfy them provided  $\sigma$  and  $\beta$  are bounded. They are introduced in order to carry out the mathematical treatment, and its great generality allows us to handle all the special cases from the literature previously mentioned. They are relevant due to its generality. It is possible to show that the absence of one (or both) of the conditions implies the occurrence of very complicated mathematical pathologies, and much more sophisticated approaches would be needed for proving that the model admits a solution (in some very delicate sense).

We introduce the variables

$$x = (x_1, x_2, x_3) = \frac{\tilde{x}}{R(t)}, \quad (2.3)$$

$$u(x, t) = \widehat{\sigma}(R(t)x, t) - \overline{\sigma} \quad (2.4)$$

and

$$v(x, t) = \widehat{\beta}(R(t)x, t) - \overline{\beta}. \quad (2.5)$$

The unit ball  $\{x \in \mathbb{R}^3, |x| < 1\}$  is denoted by  $B$  and we define the (multivalued) functions from  $\mathbb{R}^2$  to  $2^{\mathbb{R}^2}$  by

$$\begin{cases} g_1(\widehat{\sigma} - \overline{\sigma}, \widehat{\beta} - \overline{\beta}) := r_1((\sigma_B - \widehat{\sigma}) - \lambda\widehat{\sigma} - \widehat{\beta})H(\widehat{\sigma} - \sigma_n) + \widehat{g}_1(\widehat{\sigma}, \widehat{\beta}), \\ g_2(\widehat{\sigma} - \overline{\sigma}, \widehat{\beta} - \overline{\beta}) := -r_2\widehat{\beta}H(\widehat{\sigma} - \sigma_n) + \widehat{g}_2(\widehat{\sigma}, \widehat{\beta}), \end{cases} \quad (2.6)$$

$$S(\widehat{\sigma} - \overline{\sigma}, \widehat{\beta} - \overline{\beta}) := \frac{4}{3\pi}\widehat{S}(\widehat{\sigma}, \widehat{\beta}) \quad (2.7)$$

and

$$f(x, t) := \tilde{f}(xR(t), t), \quad \tilde{\omega}_0^t = \{(x, t) \in B \times [0, T] \text{ such that } R(t)x \in \omega_0\}.$$

Problem (1.2) becomes

$$\left\{ \begin{array}{ll} \frac{\partial u}{\partial t} - \frac{d_1}{R(t)^2} \Delta u - \frac{R'(t)}{R(t)} x \cdot \nabla u \in g_1(u, v), & x \in B \ t > 0, \\ \frac{\partial v}{\partial t} - \frac{d_2}{R(t)^2} \Delta v - \frac{R'(t)}{R(t)} x \cdot \nabla v \in g_2(u, v) + f \chi_{\tilde{\omega}_0^t}, & x \in B \ t > 0, \\ R(t)^{-1} \frac{dR(t)}{dt} = \int_B S(u, v) dx, & t > 0, \\ u(x, t) = v(x, t) = 0, & x \in \partial B \ t > 0, \\ R(0) = R_0, \ u(x, 0) = u_0(x), \ v(x, 0) = v_0(x), & x \in B. \end{array} \right. \quad (2.8)$$

We introduce the Hilbert spaces

$$\mathbf{H}(B) := L^2(B)^2, \quad \mathbf{V}(B) = H_0^1(B)^2,$$

and define inner products by

$$\langle \Phi, \Psi \rangle_{\mathbf{H}(B)} = \int_B \Phi \cdot \Psi^t dx,$$

$$\langle \Phi, \Psi \rangle_{\mathbf{V}(B)} = \sum_{i=1,2} d_i \int_B (\nabla \Phi_i)^t \cdot \nabla \Psi_i dx,$$

for all  $\Phi = (\Phi_1, \Phi_2)$ ,  $\Psi = (\Psi_1, \Psi_2)$ .

For the sake of notational simplicity we use  $\mathbf{H} = \mathbf{H}(B)$  and  $\mathbf{V} = \mathbf{V}(B)$ . Given  $T > 0$ , we introduce  $U = (u, v)$ ,  $U_0 = (u_0, v_0)$  and define  $G : \mathbb{R}^2 \rightarrow 2^{\mathbb{R}^2} \times 2^{\mathbb{R}^2}$  and  $F : (0, T) \times B \rightarrow \mathbb{R}^2$  by

$$G(U) = (g_1(u, v), g_2(u, v)), \quad F(t, x) = (0, f(t, x) \chi_{\tilde{\omega}_0^t}).$$

We have:

$$|G(U)| = |g_1(u, v)| + |g_2(u, v)| \leq C_0 + C_1|U| = C_0 + C_1(|u| + |v|). \quad (2.9)$$

**Definition 2.1.**  $(U, R) \in L^2(0, T : \mathbf{V}) \times W^{1, \infty}(0, T : \mathbb{R})$  is a weak solution of the problem (2.8) if there exists  $g^* = (g_1^*, g_2^*) \in L^2(0, T : \mathbf{H})$  with  $g^*(x, t) \in G(U(x, t))$  a.e.  $(x, t) \in B \times (0, T)$  and

$$\int_0^T - \langle U, \Phi_t \rangle_{\mathbf{H}} dt + \int_0^T \tilde{a}(R(t), U, \Phi) dt = \int_0^T \langle g^*, \Phi \rangle_{\mathbf{H}} dt \\ + \langle U_0, \Phi(0) \rangle_{\mathbf{H}} + \int_0^T \langle F(t), \Phi \rangle_{\mathbf{H}} dt,$$

$\forall \Phi \in L^2(0, T : \mathbf{V}) \cap H^1(0, T : \mathbf{H})$  with  $\Phi(T) = 0$ , where

$$\tilde{a}(R(t), U, \Phi) := \frac{1}{R^2(t)} \langle U, \Phi \rangle_{\mathbf{V}} - \frac{R'(t)}{R(t)} \langle x \cdot \nabla U, \Phi \rangle_{\mathbf{H}} \quad (2.10)$$

and  $R(t)$  is strictly positive and given by

$$R(t)^{-1} \frac{dR(t)}{dt} = \int_B S(U(x, t)) dx \text{ for } t \in (0, T).$$

**Definition 2.2.**  $(\sigma, \beta, R)$  is a weak solution of (1.2) if

$$\sigma(\tilde{x}, t) = u\left(\frac{\tilde{x}}{R(t)}, t\right) + \bar{\sigma} \text{ and } \beta(\tilde{x}, t) = v\left(\frac{\tilde{x}}{R(t)}, t\right) + \bar{\beta},$$

for  $t \in (0, T)$  and  $\tilde{x} \in \mathbb{R}^3, |\tilde{x}| \leq R(t)$ , where  $(U = (u, v), R)$  is the weak solution of (2.8) for any  $T > 0$ .

**Remark 2.1.** The definition of weak solution and the structural assumptions on  $G$  imply that  $\frac{\partial U}{\partial t} \in L^2(0, T : \mathbf{V}'(B))$  and the equation holds in  $D'(B \times (0, T))$ .

**Theorem 2.1.** Assume (2.1), (2.2),  $R_0 > 0$  and  $\sigma_0, \beta_0 \in L^2(0, R_0)$  then (1.2) has at least a weak solution for each  $T > 0$ .

*Proof.* We shall use a Galerkin method to construct a weak solution. Let  $R(t) \in W^{1, \infty}(0, T : \mathbb{R})$  such that  $\frac{R'(t)}{R(t)} \geq -\lambda_0$  a.e.  $t \in (0, T)$ . For fixed  $t \in (0, T)$ , we consider the operator  $\mathbf{A}(t) \equiv \mathbf{A}(R(t)) : \mathbf{V} \rightarrow \mathbf{V}'$  defined by

$$\mathbf{A}(R(t))(U) = \begin{pmatrix} -\frac{d_1}{R(t)^2} \Delta u - \frac{R'(t)}{R(t)} x \cdot \nabla u & 0 \\ 0 & -\frac{d_2}{R(t)^2} \Delta v - \frac{R'(t)}{R(t)} x \cdot \nabla v \end{pmatrix}.$$

$\mathbf{A}(t)$  defines a continuous, bilinear form on  $\mathbf{V} \times \mathbf{V}$

$$\tilde{a}(t : \cdot, \cdot) : \mathbf{V} \times \mathbf{V} \longrightarrow \mathbb{R},$$

for a.e.  $t \in (0, T)$  (see (2.10)). Since  $\frac{R'(t)}{R(t)} \geq -\lambda_0$ ,  $\tilde{a}(t, \cdot, \cdot)$  satisfies

$$\begin{aligned} \tilde{a}(t, U, U) &= \frac{1}{R^2(t)} \langle U, U \rangle_{\mathbf{V}} - \frac{R'(t)}{R(t)} \langle x \cdot \nabla U, U \rangle_{\mathbf{H}} \\ &= \frac{1}{R^2(t)} \langle U, U \rangle_{\mathbf{V}} + \frac{R'(t)}{2R(t)} \langle U, U \rangle_{\mathbf{H}} \geq (\max_{0 < t < T} \{R(t)\})^{-2} \|U\|_{\mathbf{V}}^2 \\ &\quad - \frac{\lambda_0}{2} \|U\|_{\mathbf{H}}^2. \quad \square \end{aligned}$$

Now we establish some *a priori estimates* which will be used later. In fact, those estimates can be applied even for other existence methods, different from the Galerkin type one, as, for instance, iterative methods, fixed point methods, etc.

$$\mathbf{Lemma\ 2.1.} \quad \|U\|_{\mathbf{H}}^2 \leq \frac{1}{2} C_0^2 (\exp\{(\frac{\lambda_0}{2} + C_1 + 1)T\} - 1) + \frac{1}{2} \|F\|_{L^2(0, T; \mathbf{H})}^2 + \|U_0\|_{\mathbf{H}}^2.$$

*Proof.* Inserting  $U^t$  as test function into the weak formulation of (2.8), one obtains

$$\frac{d}{dt} \int_B \frac{1}{2} U^2 dx + \tilde{a}(R(t), U, U) + \int_B g^*(U) U^t dx = \int_B F \cdot U^t dx$$

for some  $g^* \in L^2((0, T) \times B)^2$  and  $g^*(x, t) \in G(U(x, t))$  for a.e.  $(x, t) \in B \times (0, T)$ . The definition of  $\tilde{a}$  yields

$$\frac{1}{2} \frac{d}{dt} \|U\|_{\mathbf{H}}^2 - \frac{\lambda_0}{2} \|U\|_{\mathbf{H}}^2 \leq (\|g^*\|_{\mathbf{H}} + \|F\|_{\mathbf{H}}) \|U\|_{\mathbf{H}}. \quad (2.11)$$

Thus Young inequality and (2.9) imply

$$\frac{1}{2} \frac{d}{dt} \|U\|_{\mathbf{H}}^2 - (\frac{\lambda_0}{2} + C_1 + 1) \|U\|_{\mathbf{H}}^2 \leq \frac{1}{4} (C_0^2 + \|F\|_{\mathbf{H}}^2).$$

Integrating with respect to time, we get

$$\begin{aligned} \frac{1}{2} \|U\|_{\mathbf{H}}^2 - \frac{1}{2} \|U_0\|_{\mathbf{H}}^2 - (\frac{\lambda_0}{2} + C_1 + 1) \|U\|_{L^2(0, T; \mathbf{H})}^2 \\ \leq \frac{1}{4} (C_0^2 T + \|F\|_{L^2(0, T; \mathbf{H})}^2) \end{aligned}$$



and by Gronwall Lemma

$$\begin{aligned} \|U\|_{\mathbf{H}}^2 &\leq \frac{1}{2}C_0^2(\exp\{(\frac{\lambda_0}{2} + C_1 + 1)T\} - 1) \\ &\quad + \frac{1}{2}\|F\|_{L^2(0,T;\mathbf{H})}^2 + \|U_0\|_{\mathbf{H}}^2 \leq C. \quad \square \quad (2.12) \end{aligned}$$

**Remark 2.2.** Since  $U$  is bounded in  $\mathbf{H}$  (by (2.12)),  $R$  satisfies

$$R(t) = R_0 \exp\left\{\int_0^t \int_0^1 S(U) dx dt\right\} \leq R_0 e^{K_1 t}, \quad (2.13)$$

and

$$R(t) \geq R_0 \exp\{-\lambda_0 t\}, \quad (2.14)$$

consequently  $R \in W^{1,\infty}(0, T)$ .

**Lemma 2.2.**  $\|U\|_{L^2(0,T;\mathbf{V})} \leq K(T, F, G, U_0)$ .

*Proof.* Selecting  $U$  as test function in (2.12), we have

$$\begin{aligned} \frac{D}{R_0^2 e^{2K_1 T}} \|U\|_{L^2(0,T;\mathbf{V})}^2 - \frac{\lambda_0}{2} \|U\|_{L^2(0,T;\mathbf{H})}^2 &\leq C_1 \|U\|_{L^2(0,T;\mathbf{H})}^2 \\ &\quad + (C_0 + \|F\|_{L^2(0,T;\mathbf{H})}) \|U\|_{L^2(0,T;\mathbf{H})}. \end{aligned}$$

By (2.12) we get

$$\|U\|_{L^2(0,T;\mathbf{V})} \leq K(F, G, U_0, T). \quad \square \quad (2.15)$$

**Remark 2.3.** By Lemma 2.2 and Remark 2.2 we get that

$$u_t - \frac{d_1}{R^2} \Delta u \in L^2(0, T; L^2(B)), \quad v_t - \frac{d_2}{R^2} \Delta v \in L^2(0, T; L^2(B)),$$

to obtain the extra regularity

$$U_t, \Delta U \in [L^2(0, T; L^2(B))]^2. \quad (2.16)$$

Now, as previously in the proof of Theorem 2.1, we consider the approximate problem

$$\begin{aligned} \frac{\partial U^\epsilon}{\partial t} + A(R^\epsilon(t))U^\epsilon &= G^\epsilon(U^\epsilon) + F(t) \text{ on } B \times (0, T), \\ U^\epsilon(0, x) &= U_0, \quad U^\epsilon = 0 \text{ on } \partial B, \end{aligned} \quad (2.17)$$

$$\frac{1}{R^\epsilon(t)} \frac{dR^\epsilon}{dt} = \int_B S(U^\epsilon) dx,$$

where  $G^\epsilon = (g_1^\epsilon, g_2^\epsilon)$  is a Lipschitz continuous function such that

$$G^\epsilon \longrightarrow G \text{ when } \epsilon \rightarrow 0 \text{ a.e. in } \mathbb{R}^2$$

where  $H$  have been replaced by

$$H^\epsilon(s) = \begin{cases} 0 & \text{if } s < 0, \\ \frac{s}{\epsilon} & \text{if } 0 \leq s \leq \frac{1}{\epsilon}, \\ 1 & \text{if } s > \frac{1}{\epsilon}. \end{cases}$$

Now, we apply the Galerkin method to the approximated problem. Let  $\lambda_n$  and  $\phi_n \in H_0^1(B)$  for  $n \in \mathbb{N}$  be the eigenvalues and eigenfunctions associated to  $-\Delta$  satisfying

$$-\Delta \phi_n = \lambda_n \phi_n.$$

We consider  $V_m$  the finite dimensional vector space spanned by  $\{\phi_1, \dots, \phi_m\}$ . We search for a solution  $U_m^\epsilon \in L^2(0, T : V_m)$  of the problem

$$\frac{d}{dt} U_m^\epsilon + A(R_m^\epsilon(t)) U_m^\epsilon = G^\epsilon(U_m^\epsilon) + F_m(t), \quad (2.18)$$

with

$$R_m^\epsilon(t)^{-1} \frac{dR_m^\epsilon(t)}{dt} = \int_B S(U_m^\epsilon(x, t)) dx.$$

Then

$$R_m^\epsilon(t) = R_0 \exp\left\{ \int_0^t \int_B S(U_m^\epsilon(x, s)) dx ds \right\}$$

and the initial conditions  $U_m^\epsilon(0) = P_m(U_0)$  (where  $P_m$  is the orthogonal projection from  $L^2(B)$  onto  $V_m$ ) and  $F_m = P_m(F)$ .

**Proposition 2.1.** (2.18) has a unique solution  $U_m^\epsilon$  for any  $T < \infty$ .

*Proof.* Problem (2.18) can be written as a suitable nonlinear ordinary differential system. Let  $U_m^\epsilon = (u_m^\epsilon, v_m^\epsilon)$  be defined by

$$u_m^\epsilon(t) = \sum_{n=1, \dots, m} a_n^{\epsilon m}(t) \phi_n, \quad v_m^\epsilon(t) = \sum_{n=1, \dots, m} b_n^{\epsilon m}(t) \phi_n,$$

and denote

$$a^{\epsilon m} = (a_1^{\epsilon m}, a_2^{\epsilon m}, \dots, a_m^{\epsilon m}), \quad b^{\epsilon m} = (b_1^{\epsilon m}, b_2^{\epsilon m}, \dots, b_m^{\epsilon m}),$$

$$\lambda_a = (\lambda_1 a_1^{\epsilon m}, \dots, \lambda_m a_m^{\epsilon m}),$$

and  $\lambda_b = (\lambda_1 b_1^{\epsilon m}, \dots, \lambda_m b_m^{\epsilon m})$ . Then  $a^{\epsilon m}$ ,  $b^{\epsilon m}$  and  $R_m^\epsilon$  satisfy

$$\dot{a}^{\epsilon m} + \frac{\lambda_a}{(R_m^\epsilon(t))^2} + \phi_\epsilon(a^{\epsilon m}, b^{\epsilon m}) L_1^m(a^{\epsilon m}, b^{\epsilon m}) + g_1^m(a^{\epsilon m}, b^{\epsilon m}) = 0,$$

$$\dot{b}^{\epsilon m} + \frac{\lambda_b}{(R_m^\epsilon(t))^2} + \phi_\epsilon(a^{\epsilon m}, b^{\epsilon m}) L_2^m(a^{\epsilon m}, b^{\epsilon m}) + g_2^m(a^{\epsilon m}, b^{\epsilon m}) = F^m(t),$$

$$\frac{\dot{R}_m^\epsilon}{R_m^\epsilon(t)} = \phi_\epsilon(a^{\epsilon m}, b^{\epsilon m}),$$

where

$$\phi_\epsilon(a^{\epsilon m}, b^{\epsilon m}) = \int_B S(U_m^\epsilon) dx,$$

$$L_1^m(a^{\epsilon m}, b^{\epsilon m}) = \int_B x \cdot \nabla u_m^\epsilon \phi_n dx \text{ for } n = 1, \dots, m,$$

$$L_2^m(a^{\epsilon m}, b^{\epsilon m}) = \int_B x \cdot \nabla v_m^\epsilon \phi_n dx \text{ for } n = 1, \dots, m,$$

$$g_1^m(a^{\epsilon m}, b^{\epsilon m}) = \int_B g_1^\epsilon(u_m^\epsilon, v_m^\epsilon) \phi_n dx \text{ for } n = 1, \dots, m,$$

$$g_2^m(a^{\epsilon m}, b^{\epsilon m}) = \int_B g_2^\epsilon(u_m^\epsilon, v_m^\epsilon) \phi_n dx \text{ for } n = 1, \dots, m.$$

Since  $G_\epsilon$  is a Lipschitz function we obtain that there exists a unique solution  $a^{\epsilon m}, b^{\epsilon m}, R_m^\epsilon$  to the system for  $T$  small enough. Moreover, (2.12) and (2.14) hold, and we get the existence of a solution of (2.18) for any  $T < \infty$ .

By (2.16) and (2.15) we obtain that  $\{(U_m^\epsilon, \frac{d}{dt}U_m^\epsilon)\}_{m=1,\infty}$  is uniformly bounded in  $L^2(0, T : \mathbf{V}) \times L^2(0, T : \mathbf{V}')$ . So, there exists a subsequence  $U_{m_i}^\epsilon \in L^2(0, T : \mathbf{V})$  with  $\frac{d}{dt}U_{m_i}^\epsilon \in L^2(0, T : \mathbf{V}')$  such that

$$(U_{m_i}^\epsilon, \frac{d}{dt}U_{m_i}^\epsilon) \rightharpoonup (U^\epsilon, \frac{d}{dt}U^\epsilon) \text{ weakly in } L^2(0, T : \mathbf{V}) \times L^2(0, T : \mathbf{V}').$$

Taking limits when  $m_i \rightarrow \infty$  we get the existence of solution to (2.17) for any  $T < \infty$ .

To end the proof of Theorem 2.1, we take limits in the equation when  $\epsilon \rightarrow 0$ . We employ (2.12) and (2.14) and the compact embedding  $\mathbf{H}_0^1(B) \subset L^s(B)$  (for  $s < 6$ ) in order to obtain the existence of a subsequence  $U^{\epsilon_i}$  such that

$$U^{\epsilon_i} \rightarrow U \text{ in } L^2(0, T : [L^s(B)]^2)$$

and in particular

$$U^{\epsilon_i} \rightarrow U \text{ in } L^2(0, T : \mathbf{H})$$

(see e.g. Simon [15]). Since

$$H^\epsilon(u^\epsilon + \bar{\sigma}) \rightarrow h \in H(u + c) \text{ weakly in } L^2(0, T : L^s(B))$$

and

$$v^\epsilon \rightarrow v \text{ strong in } L^2(0, T : L^s(B))$$

(see Lemma 3.4.1 of Vrabie [16]) we have

$$G^{\epsilon_i}(U^{\epsilon_i}) \rightarrow g^* \in G(U) \text{ weakly in } L^1(0, T : \mathbf{H}).$$

Since  $|R'| \leq C$  there exists a subsequence  $R_{\epsilon_{ij}}$  such that

$$R_{\epsilon_{ij}} \rightarrow R \text{ weakly in } W^{1,p}(0, T), \quad p < \infty.$$

By (2.11) we deduce that  $R_{\epsilon_{ij}} \rightarrow R$  in  $C^0([0, T])$ . Finally, taking limits in the weak formulation of the problem (2.12) we get

$$\begin{aligned} \int_0^T \langle U_t, \Phi \rangle_{\mathbf{H}} dt + \int_0^T \tilde{a}(R(t), U, \Phi) dt + \int_0^T \langle g^*, \Phi \rangle_{\mathbf{H}} dt \\ = \int_0^T \langle F, \Phi \rangle_{\mathbf{H}} dt, \end{aligned}$$

for all  $\Phi \in L^2(0, T : V)$  and moreover,

$$R(t)^{-1} \frac{dR(t)}{dt} = \int_B S(U(x, t)) dx.$$

Notice that

$$\begin{aligned} \int_0^T \frac{R'_{\epsilon ij}}{R_{\epsilon ij}} \int_B x \cdot \nabla u_{\epsilon ij} \psi dx dt \\ = \int_0^T \frac{R'_{\epsilon ij}}{R_{\epsilon ij}} \int_B u_{\epsilon ij} \psi dx dt - \int_0^T \frac{R'_{\epsilon ij}}{R_{\epsilon ij}} \int_B u_{\epsilon ij} x \cdot \nabla \psi dx dt \end{aligned}$$

and

$$\begin{aligned} \int_0^T \frac{R'_{\epsilon ij}}{R_{\epsilon ij}} \int_B x \cdot \nabla v_{\epsilon ij} \psi dx dt \\ = \int_0^T \frac{R'_{\epsilon ij}}{R_{\epsilon ij}} \int_B v_{\epsilon ij} \psi dx dt - \int_0^T \frac{R'_{\epsilon ij}}{R_{\epsilon ij}} \int_B v_{\epsilon ij} x \cdot \nabla \psi dx dt. \end{aligned}$$

We conclude that  $(\sigma, \beta, R)$  defined by

$$\sigma(t, \tilde{x}) = u(t, \frac{\tilde{x}}{R(t)}) + \bar{\sigma} \text{ and } \beta(t, \tilde{x}) = v(t, \frac{\tilde{x}}{R(t)}) + \bar{\beta}$$

is a weak solution of problem (1.2). The additional regularity

$$\hat{\sigma}_t - d_1 \Delta \hat{\sigma} \text{ and } \hat{\beta}_t - d_2 \Delta \hat{\beta} \in L^2(\cup_{t \in [0, T]} (0, R(t)) \times \{t\})$$

follows from the fact that

$$\frac{\partial U}{\partial t}(t) + \mathbf{A}(R(t))U(t) \in L^2(0, T : L^2(B)^2). \quad \square$$

### 3. Uniqueness of Solutions with Radial Symmetry

In this section we shall prove the uniqueness of radial symmetric weak solutions. We start by pointing out that if, for instance,  $\sigma_n \geq \frac{r_1 \sigma_B}{r_1 + \lambda}$ ,  $r_1 \sigma_B > 0$ ,  $\hat{g}_1(\hat{\sigma}, \hat{\beta})$  is a decreasing function of  $\hat{\sigma}$  and independent of  $\hat{\beta}$

and the initial datum  $\sigma_0(\bar{x})$  is such that  $\sigma_0'(\rho_0) = \sigma_0''(\rho_0) = 0$ , then it is possible to adapt the arguments of Díaz and L. Tello [7] in order to construct more than one solution of problem (1.1)-(1.5). This and the presence of non-Lipschitz terms at both equations clarify that any possible uniqueness result will require an significant set of additional conditions.

Cui and Friedman [5] prove uniqueness of solution for the non necrotic case (i.e. linear functions  $g_i$ ).

As in previous section, we can prove that there exists at least one radial symmetric solution  $(\widehat{\sigma}, \widehat{\beta})$  to (1.2). We define  $\sigma = \widehat{\sigma} - \bar{\sigma}$ ,  $\beta = \widehat{\beta} - \bar{\beta}$  and  $r = |x|$ . Then  $(\sigma, \beta)$  verifies

$$\left\{ \begin{array}{ll} \frac{\partial \sigma}{\partial t} - \frac{d_1}{r^2} \frac{\partial}{\partial r} (r^2 \frac{\partial \sigma}{\partial r}) \in g_1(\sigma, \beta), & 0 < r < R(t) \quad 0 < t < T, \\ \frac{\partial \beta}{\partial t} - \frac{d_2}{r^2} \frac{\partial}{\partial r} (r^2 \frac{\partial \beta}{\partial r}) \in g_2(\sigma, \beta), & 0 < r < R(t) \quad 0 < t < T, \\ R(t)^2 \frac{dR(t)}{dt} = \int_0^{R(t)} S(\sigma, \beta) r^2 dr, & 0 < t < T, \\ \frac{\partial \sigma}{\partial r}(0, t) = 0, \quad \frac{\partial \beta}{\partial r}(0, t) = 0, & 0 < t < T, \\ \sigma(R(t), t) = 0, \quad \beta(R(t), t) = 0, & 0 < t < T, \\ R(0) = R_0, \\ \sigma(r, 0) = \sigma_0(r), \quad \beta(r, 0) = \beta_0(r), & 0 < r < R_0, \end{array} \right. \quad (3.1)$$

where  $g_i$  are given by (2.6) when  $\hat{g}_i = 0$ , i.e.

$$g_1(\sigma, \beta) = -[(r_1 + \lambda)(\sigma + \bar{\sigma}) - r_1 \sigma_B + (\beta + \bar{\beta})]H(\sigma + \bar{\sigma} - \sigma_n), \quad (3.2)$$

$$g_2(\sigma, \beta) = -r_2(\beta + \bar{\beta}). \quad (3.3)$$

We will assume throughout this section that

$$S(\sigma, \beta) \in W_{loc}^{1, \infty}(\mathbb{R}^2), \quad (3.4)$$

$$S \text{ is an increasing function in } \sigma \text{ and decreasing in } \beta \quad (3.5)$$

$$\sigma_n \geq \frac{r_1 \sigma_B - \bar{\beta}}{r_1 + \lambda}, \quad (3.6)$$

and the initial data  $(\sigma_0 = \widehat{\sigma} - \overline{\sigma}, \beta_0 = \widehat{\beta}_0 - \overline{\beta})$  belong to  $H^2(0, R_0)$  and satisfy

$$\frac{\partial \sigma_0}{\partial r}(0, t) = 0, \quad \frac{\partial \beta}{\partial r}(0, t) = 0 \quad 0 < t < T, \quad (3.7)$$

$$\sigma(R(t), t) = 0, \quad \beta(R(t), t) = 0 \quad 0 < t < T. \quad (3.8)$$

**Theorem 3.1.** *There is, at most, one solution to (3.1).*

Let us introduce the functions

$$T_0(s) = \begin{cases} s & \text{if } s \geq 0, \\ 0 & \text{otherwise} \end{cases} \quad T^0(s) = \begin{cases} s & \text{if } s \leq 0, \\ 0 & \text{otherwise} \end{cases}$$

which we will use in the proof of the theorem.

**Lemma 3.1.** *Every solution  $(\sigma, \beta)$  to (3.1) is bounded and satisfies  $\sigma_n \leq \sigma \leq \sigma_B$  and  $-\overline{\beta} \leq \beta \leq \max\{\beta_0\}$  (provided  $\sigma_n \leq \sigma_0 \leq \sigma_B$  and  $-\overline{\beta} \leq \beta_0$ ).*

*Proof.* By the “integrations by parts formula” (justifying the multiplication of the equation by  $T_0(\sigma - \sigma_B)$  and posterior integration in time and space, see Alt and Luckhaus [1] Lemma 1.5) we have

$$\frac{1}{2} \int_0^{R(t)} [T_0(\sigma - \sigma_B)]^2 r^2 dr \leq \int_0^t \int_0^{R(s)} g_1(\sigma, \beta) T_0(\sigma - \sigma_B) r^2 dr ds.$$

Since

$$\begin{aligned} & -[(r_1 + \lambda)(\sigma + \overline{\sigma}) - r_1 \sigma_B + (\beta + \overline{\beta})] H(\sigma + \overline{\sigma} - \sigma_n) T_0(\sigma - \sigma_B) \\ &= -(r_1 + \lambda) T_0(\sigma - \sigma_B)^2 - [(r_1 + \lambda)(\sigma_B + \overline{\sigma}) - r_1 \sigma_B + (\beta - \overline{\beta})] T_0(\sigma - \sigma_B) \\ &\quad \leq -[(\lambda \sigma_B + (r_1 + \lambda) \overline{\sigma} + (\beta + \overline{\beta}))] T_0(\sigma - \sigma_B) \\ &\quad \leq T^0(\beta + \overline{\beta}) T_0(\sigma - \sigma_B) \leq \frac{1}{2} ([T^0(\beta + \overline{\beta})]^2 + [T_0(\sigma - \sigma_B)]^2), \end{aligned}$$

we obtain

$$\begin{aligned} & \int_0^{R(t)} T_0(\sigma - \sigma_B)^2 r^2 dr \\ &\quad \leq \int_0^t \int_0^{R(s)} [T^0(\beta + \overline{\beta})^2 + T_0(\sigma - \sigma_B)^2] r^2 dr ds. \quad (3.9) \end{aligned}$$

In the same way, we consider  $T^0(\beta + \bar{\beta})$  and since

$$r_2(\beta + \bar{\beta})H(\sigma + \bar{\sigma} - \sigma_n)T^0(\beta + \bar{\beta}) \leq r_2[T^0(\beta + \bar{\beta})]^2,$$

it follows that

$$\int_0^{R(t)} [T^0(\beta + \bar{\beta})]^2 r^2 dr \leq \int_0^t \int_0^{R(s)} r_2 T^0(\beta + \bar{\beta}) r^2 dr ds. \quad (3.10)$$

Adding (3.9) and (3.10), we obtain thanks to Gronwall Lemma,

$$\sigma \leq \sigma_B \text{ and } \beta \geq -\bar{\beta}.$$

Notice that  $\beta \geq -\bar{\beta}$  implies  $\hat{\beta} \geq 0$ .

Let us consider  $\epsilon > 0$  and take  $T^0(\sigma - \sigma_n - \epsilon)$  as test function in the weak formulation, then

$$\frac{1}{2} \int_0^{R(t)} [T^0(\sigma - \sigma_n - \epsilon)]^2 r^2 dr \leq 0.$$

Now, taking limits as  $\epsilon \rightarrow 0$ , one concludes

$$\frac{1}{2} \int_0^{R(t)} [T^0(\sigma - \sigma_n)]^2 r^2 dr \leq 0,$$

which proves  $\sigma \geq \sigma_n$ .

Knowing  $\sigma$  and  $R$ ,  $\beta$  is well defined as the unique solution of the equation

$$\frac{\partial \beta}{\partial t} - \frac{d_2}{r^2} \frac{\partial}{\partial r} (r^2 \frac{\partial \beta}{\partial r}) = -r_2(\beta + \bar{\beta}), \quad 0 < r < R(t), \quad 0 < t < T$$

$$\beta(R(t), t) = 0, \quad \frac{\partial \beta}{\partial r} = 0 \text{ on } 0 < t < T.$$

Since  $\beta_0 \geq -\bar{\beta}$  it results that

$$\frac{\partial \beta}{\partial t} - \frac{d_2}{r^2} \frac{\partial}{\partial r} (r^2 \frac{\partial \beta}{\partial r}) \leq 0$$

and we obtain by maximum principle that  $\beta \leq \max\{\beta_0\}$ .  $\square$

**Corollary 3.1.** *There exists a positive constant  $M$  such that  $R(t) \leq R_0 e^{Mt}$  and  $R'(t) \leq R_0 M e^{Mt}$ .*



*Proof.* The above result shows  $(\sigma(r, t), \beta(r, t)) \in [\sigma_n, \sigma_B] \times [-\bar{\beta}, \max\{\beta_0\}]$  and by (3.4) we get the conclusion.  $\square$

**Lemma 3.2.** *The solution  $(\sigma, \beta)$  of (3.1) satisfies*

$$\int_0^T (\|\sigma\|_{W^{1,\infty}(\epsilon, R(t))}^2 + \|\beta\|_{W^{1,\infty}(\epsilon, R(t))}^2) dt \leq C_1,$$

for all  $\epsilon > 0$ .

*Proof.* The pair  $(u(x, t), v(x, t)) = (\sigma(R(t)|x|, t), \beta(R(t)|x|, t))$  is a solution to (2.8) and so  $(u, v) \in [L^2(0, T : H^1(B))]^2$ . By (2.3) and

$$\tau(t) = \int_0^t R^{-2}(\rho) d\rho, \quad (3.11)$$

we obtain that  $\tau(t) \in C^1$ . By the Implicit Function Theorem,  $t(\tau) \in C^1$  and then  $(u, v) \in L^2(0, T : H^2(B))^2$  (see e.g. Brezis [3]). Since  $(u, v)$  are symmetric we define

$$\tilde{u}(|x|, t) := u(x, t) \quad \text{and} \quad \tilde{v}(|x|, t) := v(x, t),$$

which belong to  $L^2(0, T : H^2(\epsilon_0, 1)) \subset L^2(0, T : W^{1,\infty}(\epsilon_0, 1))$  for all  $\epsilon_0 > 0$ . Doing the change of variable  $r = R(t)|x|$  we obtain

$$\begin{aligned} & \int_0^T (\|\sigma\|_{W^{1,\infty}(\epsilon, R(t))}^2 + \|\beta\|_{W^{1,\infty}(\epsilon, R(t))}^2) dt \\ &= \int_0^T R^2(t) (\|\tilde{u}\|_{W^{1,\infty}(\frac{\epsilon}{R(t)}, 1)}^2 + \|\tilde{v}\|_{W^{1,\infty}(\frac{\epsilon}{R(t)}, 1)}^2) dt \\ &\leq \int_0^T R^2(t) (\|\tilde{u}\|_{W^{1,\infty}(\epsilon_0, 1)}^2 + \|\tilde{v}\|_{W^{1,\infty}(\epsilon_0, 1)}^2) dt \leq C_1 \end{aligned}$$

and the proof ends.  $\square$

*Proof of Theorem 3.1.* We argue by contradiction and assume that  $(\sigma_1, \beta_1, R_1)$  and  $(\sigma_2, \beta_2, R_2)$  are two solutions of the problem. Let  $R(t) := \min\{R_1(t), R_2(t)\}$ ,  $\sigma := \sigma_1 - \sigma_2$  and  $\beta := \beta_1 - \beta_2$  be the

solution to

$$\left\{ \begin{array}{ll} \frac{\partial \sigma}{\partial t} - \frac{d_1}{r^2} \frac{\partial}{\partial r} (r^2 \frac{\partial \sigma}{\partial r}) = g_1(\sigma_1, \beta_1) - g_1(\sigma_2, \beta_2) & 0 < r < R(t) \quad 0 < t < T, \\ \frac{\partial \beta}{\partial t} - \frac{d_2}{r^2} \frac{\partial}{\partial r} (r^2 \frac{\partial \beta}{\partial r}) = g_2(\sigma_1, \beta_1) - g_2(\sigma_2, \beta_2) & 0 < r < R(t) \quad 0 < t < T, \\ \frac{\partial \sigma}{\partial r}(0, t) = 0, \quad \frac{\partial \beta}{\partial r}(0, t) = 0 & 0 < t < T, \\ \sigma(R(t), t) = \sigma_1(R(t), t) - \sigma_2(R(t), t) & 0 < t < T, \\ \beta(R(t), t) = \beta_1(R(t), t) - \beta_2(R(t), t) & 0 < t < T, \\ \sigma(r, 0) = 0, \quad \beta(r, 0) = 0 & 0 < r < R_0. \end{array} \right. \quad (3.12)$$

Now, we state a technical lemma.

**Lemma 3.3.**  $|\beta|$  takes the maximum on the boundary  $R(t)$  and  $\sigma$  satisfies

$$\int_0^{R(t)} [T_0(\sigma - \sigma^*)]^2 r^2 dr \leq TC [\max_{t \in [0, T]} \{\beta\}]^2,$$

where

$$\sigma^* = \max_{t \in [0, T]} \{\sigma(R(t), t)\}.$$

*Proof.* Let us consider  $\beta_* = \min\{0, \beta(R(t), t)\}$  and

$$g_2(\beta_1) - g_2(\beta_2) = -r_2 [(\beta_1 - \bar{\beta}) - (\beta_2 - \bar{\beta})] = -r_2 \beta,$$

then

$$(g_2(\beta_1) - g_2(\beta_2))T^0(\beta - \beta_*) = -r_2 \beta T^0(\beta - \beta_*) \leq 0.$$

Multiply the equation by  $T^0(\beta - \beta_*)$ , we get

$$\int_0^{R(t)} [T^0(\beta - \beta_*)]^2 r^2 dr \leq 0$$

and obtain  $\beta \geq \beta_*$ . In the same way, we prove that  $\beta$  takes its maximum on  $R(t)$ .

$$\begin{aligned}
g_1(\sigma_1, \beta_1) - g_1(\sigma_2, \beta_2) &= -\left[ (r_1 + \lambda)(\sigma_1 + \bar{\sigma}) - r_1\sigma_B + (\beta_1 + \bar{\beta}) \right] \\
&\times H(\sigma_1 + \bar{\sigma} - \sigma_n) - \left[ (r_1 + \lambda)(\sigma_2 + \bar{\sigma}) - r_1\sigma_B + (\beta_2 + \bar{\beta}) \right] H(\sigma_2 + \bar{\sigma} - \sigma_n) \\
&= -(r_1 + \lambda)[(\sigma_1 + \bar{\sigma} - \sigma_n)H(\sigma_1 + \bar{\sigma} - \sigma_n) - (\sigma_2 + \bar{\sigma} - \sigma_n)H(\sigma_2 + \bar{\sigma} - \sigma_n)] \\
&\quad + (-(r_1 + \lambda)\sigma_n + r_1\sigma_B - \bar{\beta})(H(\sigma_1 + \bar{\sigma} - \sigma_n) - H(\sigma_2 + \bar{\sigma} - \sigma_n)) \\
&\quad \quad - [\beta_1 H(\sigma_1 + \bar{\sigma} - \sigma_n) - \beta_2 H(\sigma_2 + \bar{\sigma} - \sigma_n)].
\end{aligned}$$

Since  $(\sigma + \bar{\sigma} - \sigma_n)H(\sigma + \bar{\sigma} - \sigma_n)$  is an increasing function of  $\sigma$ , we obtain that

$$\begin{aligned}
&- [(\sigma_1 + \bar{\sigma} - \sigma_n)H(\sigma_1 + \bar{\sigma} - \sigma_n) \\
&\quad - (\sigma_2 + \bar{\sigma} - \sigma_n)H(\sigma_2 + \bar{\sigma} - \sigma_n)]T_0(\sigma_1 - \sigma_2 - \sigma^*) \leq 0.
\end{aligned}$$

Since  $-(r_1 + \lambda)\sigma_n + r_1\sigma_B - \bar{\beta} \leq 0$ , it follows that

$$\begin{aligned}
&(- (r_1 + \lambda)\sigma_n + r_1\sigma_B - \bar{\beta})(H(\sigma_1 + \bar{\sigma} - \sigma_n) \\
&\quad - H(\sigma_2 + \bar{\sigma} - \sigma_n))T_0(\sigma_1 - \sigma_2 - \sigma^*) \leq 0.
\end{aligned}$$

Then

$$\begin{aligned}
&[g_1(\sigma_1, \beta_1) - g_1(\sigma_2, \beta_2)]T_0(\sigma_1 - \sigma_2 - \sigma^*) \\
&\leq -[\beta_1 H(\sigma_1 + \bar{\sigma} - \sigma_n) - \beta_2 H(\sigma_2 + \bar{\sigma} - \sigma_n)]T_0(\sigma_1 - \sigma_2 - \sigma^*) \\
&\leq -(\beta_1 - \beta_2)H(\sigma_2 + \bar{\sigma} - \sigma_n)T_0(\sigma_1 - \sigma_2 - \sigma^*) \\
&\leq -T^0(\beta_1 - \beta_2)T_0(\sigma_1 - \sigma_2 - \sigma^*) \leq -\beta_*T_0(\sigma_1 - \sigma_2 - \sigma^*).
\end{aligned}$$

Multiplying the equation, as before, by  $T_0(\sigma - \sigma^*)$ , we get

$$\begin{aligned}
&\int_0^{R(t)} [T_0(\sigma - \sigma^*)]^2 r^2 dr + \int_0^t \int_0^{R(s)} \left[ \frac{\partial}{\partial r} T_0(\sigma - \sigma^*) \right]^2 r^2 dr ds \\
&= \int_0^t \int_0^{R(s)} (g_1(\sigma_1, \beta_1) - g_1(\sigma_2, \beta_2)) T_0(\sigma - \sigma^*) r^2 dr ds
\end{aligned}$$

$$\begin{aligned}
&\leq - \int_0^t \int_0^{R(s)} \beta_* T_0(\sigma - \sigma^*) r^2 dr ds \\
&\leq \frac{TC}{\lambda} \beta_*^2 + \lambda \int_0^t \int_0^{R(s)} [T_0(\sigma_1 - \sigma_2 - \sigma^*)]^2 r^2 dr ds.
\end{aligned}$$

Now, choose  $\lambda$  such that

$$\begin{aligned}
&\lambda \int_0^{R(s)} [T_0(\sigma_1 - \sigma_2 - \sigma^*)]^2 r^2 dr \\
&\quad - \int_0^{R(s)} \left[ \frac{\partial}{\partial r} T_0(\sigma - \sigma^*) \right]^2 r^2 dr \leq 0 \quad \text{a.e. } t \in (0, T),
\end{aligned}$$

then

$$\int_0^{R(t)} [T_0(\sigma - \sigma^*)]^2 r^2 dr \leq TC \beta_*^2$$

holds, which ends the proof.  $\square$

*End of the proof of Theorem 3.1.* Let us define

$$\delta = \max_{t \in [0, T]} \{|R_1(t) - R_2(t)|\} \geq 0,$$

and consider

$$\begin{aligned}
&R_1^2(t)R_1'(t) - R_2^2(t)R_2'(t) = \int_0^{R(t)} (S(\sigma_1, \beta_1) - S(\sigma_2, \beta_2)) r^2 dr \\
&+ \int_{R(t)}^{R_1(t)} S(\sigma_1, \beta_1) r^2 dr - \int_{R(t)}^{R_2(t)} S(\sigma_2, \beta_2) r^2 dr.
\end{aligned} \tag{3.13}$$

By (3.5) and Lemma 3.1, we obtain

$$\left| \int_{R(t)}^{R_i(t)} S(\sigma_i, \beta_i) r^2 dr \right| \leq M \delta \quad (\text{for } i = 1, 2), \tag{3.14}$$

where

$$M = \max\{S(\sigma, \beta) \text{ for any } (\sigma, \beta) \in [\sigma_n, \sigma_B] \times [\bar{\beta}, \max\{\beta_0\}]\}.$$

(3.4) and (3.5) imply

$$\int_0^{R(t)} (S(\sigma_1, \beta_1) - S(\sigma_2, \beta_2)) r^2 dr \leq C \int_0^{R(t)} (T_0(\sigma) - T_0(\beta)) r^2 dr.$$

Since  $T_0(\sigma) \leq T_0(\sigma - \sigma^*) + \sigma^*$  and  $-T^0(\beta) \leq -\beta_*$  then

$$\begin{aligned} & \int_0^{R(t)} (S(\sigma_1, \beta_1) - S(\sigma_2, \beta_2))r^2 dr \\ & \leq C \int_0^{R(t)} (T_0(\sigma - \sigma^*) + \sigma^* - \beta_*)r^2 dr \\ & \leq C' \left( \left[ \int_0^{R(t)} T_0(\sigma - \sigma^*)^2 r^2 dr \right]^{\frac{1}{2}} + \sigma^* - \beta_* \right). \end{aligned}$$

By Lemma 3.3 it follows that

$$C' \left( \left[ \int_0^{R(t)} T_0(\sigma - \sigma^*)^2 r^2 dr \right]^{\frac{1}{2}} + \sigma^* - \beta_* \right) \leq C''(\sigma^* - (T + 1)\beta_*).$$

Since  $\sigma_i(R_i(t), t) = 0$  (for  $j = 1$  or  $2$ ),  $\sigma$  and  $\beta$  satisfies

$$|\sigma(R(t), t)| \leq \left( \sum_{i=1,2} \|\sigma_i\|_{W^{1,\infty}(R(t), R_i(t))} \right) |R_1(t) - R_2(t)|,$$

$$|\beta(R(t), t)| \leq \left( \sum_{i=1,2} \|\beta_i\|_{W^{1,\infty}(R(t), R_i(t))} \right) |R_1(t) - R_2(t)|$$

and then

$$\int_0^{R(t)} (S(\sigma_1, \beta_1) - S(\sigma_2, \beta_2))r^2 dr \leq C(T + 2)\delta. \quad (3.15)$$

Integrating in time in (3.13), we get thanks to (3.14) and (3.15) that

$$R_1^3(t) - R_2^3(t) \leq TC(T + 2)\delta + 2TM\delta. \quad (3.16)$$

On the other hand, one has

$$R_1^3(t) - R_2^3(t) = (R_1(t) - R_2(t))(R_1^2 + R_1R_2 + R_2^2).$$

We can assume without loss of generality that  $\delta = R_1(t_0) - R_2(t_0)$  (for some  $t_0 \in [0, T]$ ), hence

$$R_1^3(t) - R_2^3(t) \geq 4R^2\delta.$$

Substituting this into (3.16) leads to  $\delta \leq k_0\delta T$ . Furthermore, taking  $T_1 < \frac{1}{k_0}$  necessitates  $R_1(t) = R_2(t)$  for any  $t \in [0, T_1]$ . Since  $|\beta|$  takes its

maximum at  $R(t) = R_1(t) = R_2(t)$  (and this maximum is 0), we get that  $\beta = 0$ . Substituting in (3.12) and taking  $\sigma$  as test function we obtain

$$\int_0^{R(t)} \sigma^2 r^2 dr \leq \int_0^t \int_0^{R(s)} (g_1(\sigma_1, 0) - g_1(\sigma_2, 0)) \sigma r^2 dr ds.$$

As in Lemma 3.3, since  $(\sigma_i + \bar{\sigma}_i - \sigma_n)H(\sigma_i + \bar{\sigma} - \sigma_n)$  is an increasing function of  $\sigma$  we obtain by (3.5) that  $(g_1(\sigma_1, 0) - g_1(\sigma_2, 0))\sigma \leq 0$ , which proves  $\sigma = 0$ .

Repeating the above process, starting now from  $T_1$ , we get the uniqueness of solutions for arbitrary  $T > 0$ , provided  $R(T) > 0$ .  $\square$

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