Abstract

Multiple myeloma is a plasma cell cancer that leads to a dysregulated bone remodeling process. We present a partial differential equation model describing the dynamics of bone remodeling with the presence of myeloma tumor cells. The model explicitly takes into account the roles of osteoclasts, osteoblasts, precursor cells, stromal cells, osteocytes, and tumor cells. Previous models based on ordinary differential equations make the simplifying assumption that the bone and tumor cells are adjacent to each other. However, in actuality, these cell populations are separated by the bone marrow. Our model takes this separation into account by including the diffusion of chemical factors across the marrow, which can be viewed as communication between the tumor and bone. Additionally, this model incorporates the growth of the tumor and the diminishing bone mass by utilizing a “moving boundary.” We present numerical simulations that qualitatively validate our model’s description of the cell population dynamics.

1 Introduction

Multiple myeloma is a plasma cell cancer characterized by an excess of malignant plasma cells in the bone marrow. The disease has a significant impact on the bones, the immune system, and the kidneys (American Cancer Society 2015). In the bone, patients experience pain, hypercalcemia, fractures, and spinal cord compression (Drake 2014). Spinal cord compression can lead to severe back pain, numbness, and muscle weakness. Hypercalcemia, or high levels of calcium in the blood, can result in dehydration, excessive urination, constipation, loss of appetite, weakness, drowsiness, confusion, and even kidney failure or coma. When the kidneys begin to fail and lose the ability to remove waste from the body, symptoms like weakness, shortness of breath, itching, and leg swelling can arise. The American Cancer Society expects that in 2015 the United States will see approximately 26,850 new multiple myeloma diagnoses and 11,240 deaths from the disease. Survival times range from 29 to 62 months once treatment has started (American Cancer Society 2015).

Some risk factors associated with multiple myeloma include age, gender, race, family history, and obesity. There are very few myeloma patients under the age of 35 (less than one percent), and most multiple myeloma patients are 65 years of age or older. Women are less likely to have myeloma than men, and African Americans develop the disease at least twice as often as Caucasian Americans. While the majority of multiple myeloma patients have no family history of the disease, individuals with a sibling or parent who has had multiple myeloma are four times as likely to have the disease. Other risk factors include radiation exposure and solitary plasmacytoma (American Cancer Society 2015).
1.1 Biological Background

Multiple myeloma bone disease disrupts the body’s ability to maintain a healthy skeleton (American Cancer Society 2015). Healthy bone continuously remodels itself in order to repair damaged bone, to adapt to mechanical strains, and to gain access to minerals stored in the bone (Burr 2002; Parfitt 2002). The bone remodeling process involves the removal of old, and perhaps damaged, bone and its replacement with new bone. The primary actors in this process are cells called osteoclasts, osteoblasts, and osteocytes. Together, the osteoclasts (which destroy bone) and osteoblasts (which form new bone) are called the basic multicellular unit, or BMU (Bellido et al 2014).

Osteoclasts are responsible for bone removal (also called osteolysis or bone resorption). They are multinucleated descendants of the hematopoietic monocyte-macrophage lineage. Once the remodeling process has begun, hematopoietic precursor cells are recruited to the BMU. Once there, the precursor cells differentiate into preosteoclasts. Then, the mononuclear preosteoclasts join to form the multinucleated mature osteoclast. These osteoclast precursor cells are recruited from their myeloid progenitors by macrophage colony-stimulating factor (M-CSF), tumor necrosis factors (TNF), interleukin-6 (IL-6), receptor activator of nuclear factor kappa-B ligand (RANKL), growth factors (GFs), and Activin A. Then mature osteoclast recruitment from the preosteoclast population is regulated by osteocyte-secreted RANKL and osteoblast secreted osteoprotegerin (OPG). Once bone resorption is complete, osteoclasts undergo apoptosis. While the factors that stimulate apoptosis have not yet been completely determined, in vitro experiments have shown that high calcium levels lead to osteoclast apoptosis (Bellido et al 2014).

Osteoblasts are responsible for the creation of new bone; they carry out bone matrix protein secretion and bone mineralization. Osteoblasts are the descendants of mesenchymal stem cells and are characterized by a cuboidal shape and a large nucleus located at the edge of the cell (Bellido et al 2014). As with osteoclasts, osteoblast formation is regulated by chemical factors. Osteoclast-derived coupling factors recruit osteoblast precursors from a pool of mesenchymal stem cells. Then the formation of mature osteoblasts is promoted by insulin-like growth factor (IGF), transforming growth factor-β (TGFβ), and bone morphogenetic proteins (BMPs) secreted by osteoblasts (Parfitt 1994; Bonewald and Dallas 1994; Plotkin and Bivi 2014). Once new bone has been formed, 60%–80% of osteoblasts undergo apoptosis. Some of the remaining osteoblasts flatten and become lining cells. The rest become osteocytes (Bellido et al 2014; Bonewald 2011).

Approximately 5%-20% of osteoblasts become trapped in the bone and differentiate into osteocytes. They are regularly dispersed throughout the mineralized bone and account for over 90% of the cells in the bone matrix and on the surface of the bone (Bellido et al 2014; Bonewald 2011). Osteocytes are located in lacunae and are connected by a network of dendritic processes, which are found in the canaliculi in the bone matrix. The proteins produced by osteocytes are transported through this network of lacunae and canaliculi. Thus, osteocytes can influence other cells within the bone matrix and on the surface of the bone (Buenzli 2015).

Recent studies have also shown that osteocytes play a key role in the regulation of osteoclasts and osteoblasts (Bonewald 2011). They are able to identify damaged bone and induce osteoclastogenesis with RANKL (Bellido et al 2014; Bonewald 2011). This happens in two ways. First, osteocytes going through apoptosis cause osteoclasts and stromal cells to produce RANKL, thereby stimulating osteoclast recruitment. Second, osteocytes can secrete RANKL themselves (Bellido et al 2014). Osteocytes also produce and secrete sclerostin, which inhibits osteoblast recruitment by blocking the Wnt signaling pathway (Bonewald 2011; Neve et al 2012; Kular et al 2012).

In healthy bone, the destruction of bone by osteoclasts is matched by the replacement of bone by osteoblasts so that bone mass is returned to its original state. However, in multiple myeloma patients, the bone remodeling process is out of balance. In this unhealthy bone, bone destruction outpaces bone replacement, leaving patients with bone lesions. These lesions are quite common in multiple myeloma patients; over ninety percent of patients suffer from them. They occur most often in the spine, skull, pelvis, and ribs. Bone lesions lead to pathologic fractures, bone pain, hypercalcemia, and spinal cord compression (Drake 2014). Even in complete remission, multiple myeloma patients usually do not show reduction of skeletal lesions (Wahlin et al 2009).

Multiple myeloma leads to bone lesions because myeloma tumor cells cause increased osteoclast production, increased osteoclast activity levels, and decreased osteoblast production (Mundy et al 1974; Bataille et al 1991; Valentin-Opran et al 1982; Evans et al 1989; Bataille et al 1990). This causes increased bone resorption, which in turn encourages tumor growth. This is called the multiple myeloma “vicious cycle”
Myeloma tumor cells encourage this vicious cycle through several chemical factors. Several of these factors encourage osteoclast production. Through adhesion between vascular cell adhesion protein 1 (VCAM-1) located on the stromal cells and very late antigen-4 (VLA4) located on the tumor cells, myeloma cells stimulate stromal RANKL production. This, in turn, simulates osteoclast formation (Michigami et al. 2000; Mori et al. 2004). Myeloma cells further encourage osteoclast recruitment through the production of macrophage inflammatory protein-1α (MIP-1α), tumor necrosis factor-α (TNF-α), and interleukin-3 (IL-3) (Silbermann and Roodman 2013). Myeloma also causes osteocytes to secrete additional interleukin-11 (IL-11), stimulating osteoclastogenesis (Giuliani et al. 2012).

Myeloma cells also suppress the recruitment of osteoblasts. Some chemical factors secreted by myeloma tumor cells that decrease osteoblast production are Dickkopf-related protein 1 (DKK1), IL-3, sclerostin, and secreted frizzled-related proteins (sFRPs) (Drake 2014; Tian et al. 2003; Ehrlich and Roodman 2005; Colucci et al. 2011; Oshima et al. 2005). Additionally, tumor cells increase stromal cell production of Activin A, leading to further decreased osteoblast production (Vallet et al. 2010).

The other half of the multiple myeloma “vicious cycle” is the promotion of tumor growth by osteoclast signaling. Osteoclasts secrete B-cell activating factor (BAFF) and a proliferation-inducing ligand (APRIL), which lead to increased tumor growth (Abe et al. 2006).

1.2 Mathematical Background

Power law approximations are a method of representing biological systems pioneered by Savageau for biochemical systems (Savageau 1969a,b, 1970, 1976; Voit 2000). They are equations of the form

\[
\frac{dX_i}{dt} = \sum_j \gamma_i \prod_k X_{g_{ij}}^k,
\]

where the \(X_j\) are the populations present in the biological system and the \(\gamma_i\) and \(g_{ij}\) are parameters that control the growth and decay of the populations. By expressing the power law instead as

\[
\frac{dX_i}{dt} = \sum_j \alpha_j \prod_k X_k^{h_{ik}} - \sum_j \beta_j \prod_k X_k^{h_{ik}},
\]

we separate the equation into two parts: one that promotes growth of the population and another that contributes to decay. Each part of the equation is the product of a constant (\(\alpha_j\) or \(\beta_j\)) and the cell populations that contribute to the growth or decay raised to powers (\(h_{ik}\)). This method is used by Komarova et al. (2003); Ryser et al. (2009, 2010); Ayati et al. (2010); Graham et al. (2013), and it is used in the model we present here.

We choose the more qualitative or phenomenological power law approach over mechanistic models with explicit biochemistry (Wang et al. 2011; Eudy et al. 2015; Ji et al. 2015) for a number of reasons: the models are much simpler mathematically; eventually they will be easier to parameterize from patient data; and the fundamental relationships involved are more robust to changes in the understanding of the underlying biochemistry. This last point is critical. A high fidelity mechanistic model, where the parameters are mostly estimated, would indeed provide valuable and quantitatively precise information about the underlying rate constants. However, if the mechanistic model is based on assumptions that later turn out to be false, whatever claims that are made about the underlying rate constants will also turn out to be false. We are operating under the assumption that the current consensus on the mechanisms underlying multiple myeloma bone disease are subject to change.

The model in this paper advances prior work in two main ways. First, we add a number of additional components we anticipate are necessary if a model is to be able to be used to predict patient outcomes (compare Fig. 10 with Figs. 1, 3, 6). Second, we have a spatial model that includes cytokine diffusion and explicit presence of the tumor; the model presented by Graham et al. (2012) used an implicit tumor not located in any particular part of space. Other models based on ordinary differential equations have no spatial heterogeneity (Ryser et al. 2009, 2010; Wang et al. 2011; Eudy et al. 2015; Ji et al. 2015).
2 Zero-Dimensional Models

Komarova et al (2003) used Savageau’s power law approximations to describe the dynamics of osteoclasts and osteoblasts during a healthy bone remodeling event (without the presence of multiple myeloma tumor cells). This model takes into account the autocrine and paracrine factors that contribute to the growth and decay of these two cell populations. The model, based on the cell dynamics described in Figure 1, is

\[
\frac{d}{dt} C(t) = \alpha_1 C(t)^{g_{11}} B(t)^{g_{21}} - \beta_1 C(t),
\]

(1) where \( C(t) \) is the density of osteoclasts, \( B(t) \) is the density of osteoblasts, and \( z \) is the total bone mass. \( \bar{C} \) and \( \bar{B} \) represent the steady states for osteoclasts and osteoblasts, respectively. The steady state is given by

\[
\bar{C} = \left( \frac{\beta_1}{\alpha_1} \right)^{(1-g_{11})/\Gamma} \left( \frac{\beta_2}{\alpha_2} \right)^{g_{21}/\Gamma},
\]

\[
\bar{B} = \left( \frac{\beta_1}{\alpha_1} \right)^{g_{12}/\Gamma} \left( \frac{\beta_2}{\alpha_2} \right)^{(1-g_{11})/\Gamma},
\]

where \( \Gamma = g_{12}g_{21} - (1 - g_{11})(1 - g_{22}). \) Figure 2 shows the total bone mass (as a percentage) during a simulation of a bone remodeling event initiated by an increase in osteoclasts.

Ayati et al (2010) expanded on Komarova et al.’s model by including the presence of a multiple myeloma tumor. The new variables in this model are \( T(t) \) (the density of the tumor cells), \( L_T \) (the maximum tumor size), and \( \gamma_T \) (the tumor growth constant). The equations are

\[
\frac{d}{dt} C(t) = \alpha_1 C(t)^{g_{11}} \left( 1 + r_{11} T(t) \frac{L_T}{T(t)} \right) B(t)^{g_{21}} \left( 1 + r_{21} T(t) \frac{L_T}{T(t)} \right) - \beta_1 C(t),
\]

(4)

\[
\frac{d}{dt} B(t) = \alpha_2 C(t)^{g_{12}} \left( 1 + r_{12} T(t) \frac{L_T}{T(t)} \right) B(t)^{g_{22}} \left( 1 + r_{22} T(t) \frac{L_T}{T(t)} \right) - \beta_2 B(t),
\]

(5)

\[
\frac{d}{dt} T(t) = \gamma_T T(t) \log \left( \frac{L_T}{T(t)} \right),
\]

(6) Gompertz form

\[
\frac{d}{dt} z(t) = -k_1 \max\{0, C(t) - \bar{C}\} + k_2 \max\{0, B(t) - \bar{B}\}.
\]

(7)

The parameters \( r_{11}, r_{12}, r_{21}, \) and \( r_{21} \) are all nonnegative. Thus, the addition of the tumor to this model increases osteoclast production and decreases osteoblast production. The steady state solution of this model
\( C = \left( \frac{\beta_1}{\alpha_1} \right)^{1-g_{22}+r_{22}}/\Lambda \left( \frac{\beta_2}{\alpha_2} \right)^{g_{21}(1+r_{21})}/\Lambda, \)
\( B = \left( \frac{\beta_1}{\alpha_1} \right)^{g_{12}/(\Lambda(1+r_{12}))} \left( \frac{\beta_2}{\alpha_2} \right)^{(1-g_{11}(1+r_{11}))}/\Lambda, \)
\( T = LT, \)

where \( \Lambda = \left( g_{12}/(1 + r_{12}) \right) \left( g_{21}(1 + r_{21}) \right) - \left( 1 - g_{11}(1 + r_{11}) \right) \left( 1 - g_{22} + r_{22} \right). \) Computational results for this model are shown in Figure 4. These results show increasing tumor size accompanied by increased osteoclast activity (bone removal) and decreased osteoblast activity (bone replacement).

Ayati et al (2010) also introduce a model that includes treatment functions. These treatment functions, \( V_1(t) \) and \( V_2(t) \), model the effects of proteasome inhibitors. Proteasome inhibitors promote osteoblast production and inhibit tumor growth, thereby breaking the multiple myeloma “vicious cycle.” The treatment model is

\[
\frac{d}{dt}C(t) = \alpha_1 C(t)^{g_{11}(1+r_{11})/T(t)} B(t)^{g_{21}(1+r_{21})/T(t)} - \beta_1 C(t), \tag{8}
\]

\[
\frac{d}{dt}B(t) = \alpha_2 C(t)^{g_{12}/(1+r_{12})} B(t)^{g_{22}-r_{22}} T(t) - \beta_2 - V_1(t) B(t), \tag{9}
\]

\[
\frac{d}{dt}T(t) = (\gamma_T - V_2(t)) T(t) \log \left( \frac{LT}{T(t)} \right), \tag{10}
\]

\[
\frac{d}{dt}z(t) = -k_1 \max\{0, C(t) - \bar{C}\} + k_2 \max\{0, B(t) - \bar{B}\}. \tag{11}
\]

The treatment functions used in this model are given by

\[
V_1(t) = \begin{cases} 
0, & t < t_{\text{start}} \\
v_1, & t \geq t_{\text{start}}
\end{cases}
\]

\[
V_2(t) = \begin{cases} 
0, & t < t_{\text{start}} \\
v_2, & t \geq t_{\text{start}}.
\end{cases}
\]

Figure 5 shows computational results for this model. These results are similar to Figure 4 until \( t = 600 \), when the treatment is introduced. At this time, the tumor density begins to shrink. At the same time, the number of osteoclasts decreases and the number of osteoblasts increases. This leads to recovery of lost bone mass.

3 Incorporating Osteocytes

Graham et al (2013) present a mathematical model of healthy bone remodeling that incorporates two additional cell populations: osteocytes (\( Y(t) \)) and osteoblast precursors (\( B_P(t) \)). The biological details of this model are summarized in Figure 6. The equations for this model are
respectively. The equations for this model are

\[
\frac{dY}{dt} = \alpha_1 B^{231} \left( 1 - \frac{Y}{K_Y} \right)_+ \]

recruitment of osteocytes from osteoblasts that become embedded in the bone

\[
\frac{dB_P}{dt} = \alpha_2 Y^{231} \left( 1 - \frac{Y}{K_Y} \right)^{922}_+ + \alpha_3 B^{232} \left( 1 - \frac{Y}{K_Y} \right)_+ - \beta_1 B^{232} C^{f14} - \delta B_P \]

differentiation of osteoblast precursors to osteoblasts by osteoclast signaling

\[
\frac{dC}{dt} = \alpha_4 Y^{941} B^{232} (\epsilon + B)^{913} \left( 1 - \frac{Y}{K_Y} \right)_+^{944} - \beta_3 C^{f34} \]

differentiation of osteoblast precursors to osteoblasts by the RANK/RANKL/OPG pathway

\[
\frac{dz}{dt} = \frac{k_1 C}{\text{amout of bone removed proportional to the number of osteoblasts}} + \frac{k_2 B}{\text{amount of bone formed proportional to the number of osteoblasts}}
\]

where \((x)_+ = \max\{x, 0\}\).

In this model, \(K_Y\) represents the relationship between osteocyte apoptosis and the decrease in sclerostin inhibition. The term \(1 - \frac{Y}{K_Y}\) represents the effects of sclerostin and the Wnt/\(\beta\)-catenin pathway. That is, when the number of osteocytes reaches \(K_Y\), the sclerostin level is sufficient to block Wnt signaling. This model assumes that osteocyte death is primarily governed by the initiation of the remodeling process. Thus, no osteocyte apoptosis term is included.

### 4 One-Dimensional Bone Remodeling with Multiple Myeloma

Here we present a one-dimensional model of bone remodeling with the presence of multiple myeloma tumor cells. Figure 9 is a simplified two-dimensional representation of a cross section of a bone marrow biopsy core. A section of bone and a myeloma tumor lay within the marrow. Additionally, a remodeling site is located on the edge of the bone. For our model we consider a one-dimensional representation of this spatial environment, also shown in Figure 9.

This model builds upon the model presented in Graham et al. (2013) by including additional cell populations, specifically osteoclast precursors \((C_P(t))\), stromal cells \((S(t))\), and myeloma tumor cells \((T(t))\). The interactions of the various cell populations included in this model are detailed in Figure 10. This model also incorporates the effects of chemical factors that diffuse across the marrow during the remodeling process:

- \(L_{C(t)}\): BAFF and APRIL, diffusing from the osteoclasts to the tumor cells
- \(L_{T_1(t)}\): MIP-1\(\alpha\), IL-3, and TNF\(\alpha\), diffusing from the tumor cells to the osteoclasts
- \(L_{T_2(t)}\): DKK1, IL-3, sclerostin, and sFRPs, diffusing from the tumor cells to the osteoclasts
- \(L_{S_1(t)}\): IL-6, RANKL, GFs, and Activin A, diffusing from the stromal cells to the osteoclasts
- \(L_{S_2(t)}\): Activin A, diffusing from the stromal cells to the osteoblasts

Additionally, the model includes a “moving boundary.” That is, the positions of the left and right endpoints of the marrow \((l(t)\) and \(r(t)\), respectively) are governed by the change in the bone mass and tumor density, respectively. The equations for this model are
\[ \frac{\partial S}{\partial t} = \alpha_1 S g_{11} T + \beta_1 S \]  
recruitment of stromal cells by tumor signaling  

\[ \frac{\partial T}{\partial t} = \alpha_2 [S g_{21}]_{P} + \alpha_3 T g_{32} L_C + \beta_2 T f_{21} \]  
recruitment of tumor cells by stromal cell signaling  
recruitment of tumor cells by BAFF and APRIL signaling  

\[ \frac{\partial C_P}{\partial t} = \alpha_4 L g_{41} S x + \alpha_5 C g_{51} \]  
recruitment of osteoclast precursors by stromal cell signaling  
recruitment of osteoclast precursors by osteoclasts  

\[ \frac{\partial C}{\partial t} = \gamma_1 (\epsilon + B) h_{11} + \gamma_2 C h_{13} \]  
differentiation of osteoclast precursors into osteoclasts  
differentiation of osteoclasts into osteocytes  

\[ \frac{\partial B_P}{\partial t} = \alpha_6 z g_{52} \]  
differentiation of osteoblast precursors into osteoclasts  
differentiation of osteoblasts into osteocytes  

\[ \frac{\partial Y}{\partial t} = \gamma_3 B h_{31} \]  
differentiation of osteoblasts into osteocytes  

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(17)  
(18)  
(19)  
(20)  
(21)  
(22)  
(23)
\[ \frac{\partial L_C}{\partial t} = \delta_{11} \nabla^2 L_C - \delta_{12} L_C \]  
\[ \text{diffusion of BAFF and APRIL from osteoclasts to tumor cells} \]  

\[ \frac{\partial L_{T_1}}{\partial t} = \delta_{21} \nabla^2 L_{T_1} - \delta_{22} L_{T_1} \]  
\[ \text{diffusion of MIP-1\alpha, IL-3, and TNF\alpha from tumor cells to osteoclasts} \]  

\[ \frac{\partial L_{T_2}}{\partial t} = \delta_{31} \nabla^2 L_{T_2} - \delta_{32} L_{T_2} \]  
\[ \text{diffusion of DKK1, IL-3, sclerostin, and sFRP's from tumor cells to osteoblasts} \]  

\[ \frac{\partial L_{S_1}}{\partial t} = \delta_{41} \nabla^2 L_{S_1} - \delta_{42} L_{S_1} \]  
\[ \text{diffusion of IL-6, RANKL, GFs and Activin A from stromal cells to osteoclast precursors} \]  

\[ \frac{\partial L_{S_2}}{\partial t} = \delta_{51} \nabla^2 L_{S_2} - \delta_{52} L_{S_2} \]  
\[ \text{diffusion of Activin A from stromal cells to osteoblasts} \]  

\[ \frac{dz}{dt} = -k_1 C + k_2 B \]  
\[ \text{amount of bone removed is proportional to the number of osteoclasts} \]  
\[ \text{amount of bone formed is proportional to the number of osteoblasts} \]

\[ \frac{d\ell}{dt} = \frac{dz}{at} \]  
\[ \text{movement of the left boundary is proportional to the change in bone mass} \]

\[ \frac{dr}{dt} = \frac{dT}{at} \]  
\[ \text{movement of the right boundary is proportional to the change in the tumor} \]

where \((x)_+ = \max\{x, 0\} = \begin{cases} x, & x \geq 0 \\ 0, & x < 0 \end{cases}\)  
The boxed numbers correspond with the cell signaling represented in Figure 10.

Equation 17 describes the dynamics of the stromal cell population. The stromals (the connective tissue cells of the bone marrow) are recruited by tumor cell signaling at a rate \(\alpha_1\). Stromal cell apoptosis occurs at a rate \(\beta_1\).

Equation 18 describes the dynamics of the tumor cell population. Myeloma tumor cells are recruited by stromal cell signaling at the right endpoint of the marrow. This recruitment occurs at a rate \(\alpha_2\). Tumor cells are also recruited by osteoclast signaling of BAFF and APRIL as a part of the multiple myeloma “vicious cycle.” This recruitment occurs at a rate \(\alpha_3\) and is due to the amount go these ligands present at the right endpoint of the marrow. Finally, tumor cell apoptosis occurs at a rate \(\beta_2\).

The dynamics of the osteoclast precursor cells are described in equation 19. This equation states that osteoclast precursors descend from a pool of myeloid progenitors at a rate \(\alpha_4\). This differentiation is largely influenced by stromal cell signaling at the left boundary point of the marrow. Additionally, this equation states that osteoclast precursors differentiate into osteoclasts by the RANK/RANKL/OPG pathway at a rate \(\gamma_1\). Finally, we have osteoclast precursor death at a rate \(\beta_3\).
Equation 20 describes the dynamics of the osteoclast population. This equation states the osteoclasts differentiate from the pool of osteoclast precursors by the RANK/RANKL/OPG pathway \([\gamma_1]\) at a rate \(\gamma_1\). Additionally, osteoclasts undergo apoptosis \([\beta_3]\) at a rate \(\beta_3\).

Equation 21 describes the dynamics of the osteoclast precursor population. Osteoblast precursors differentiate from a pool of mesenchymal stem cells due to osteoclast \([\alpha_2]\) and bone matrix \([\alpha_3]\) signaling. Osteoblast precursors are recruited by osteoclasts at a rate \(\alpha_5\) and by IGF-1 (secreted by the bone matrix) at a rate \(\alpha_6\). Additionally, osteoblast precursors differentiate into mature osteoblasts \([\alpha_4]\) at a rate \(\alpha_7\). Finally, osteoblast precursor undergo apoptosis \([\beta_6]\) at a rate \(\beta_6\).

The dynamics of mature osteoblasts are described by Equation 22. This equation states that osteoblast precursors are differentiated into osteoblasts \([\alpha_4]\) at a rate \(\alpha_7\). Additionally, under this model, mature osteoblasts have one of two fates: differentiation into osteocytes \([\beta_5]\) or cell death \([\beta_5]\). Osteoblasts differentiate into osteocytes at a rate \(\gamma_3\) and undergo apoptosis at a rate \(\beta_5\).

Equation 23 describes the dynamics of the osteocyte population. This equation states that osteocytes differentiate from the pool of osteoblasts \([\alpha_3]\) at a rate \(\gamma_3\). These cells undergo apoptosis \([\beta_7]\) at a rate \(\beta_7\).

Equations 24-28 describe the movement of the bone/marrow interface and the marrow/tumor interface, respectively. The bone/marrow interface \((\ell(t))\) moves to the left as the bone mass decreases. Similarly, the marrow/tumor interface \((r(t))\) move to the left as the tumor grows.

5 Results

Equations 17-31 were solved using MATLAB’s \texttt{pdepe} function CITEx with the parameter and initial condition values listed in Table 4. The diffusion values \((\delta_{i1})\) were computed based on the relationship between the size of the peptides (Stokes radius) and the known diffusion values:

\[
(\text{Stokes Radius}) = 0.0156(\text{molecular weight}) + 1.527 \\
(\text{diffusion constant}) = -4 \times 10^{-7}(\text{Stokes Radius}) + 2 \times 10^{-6}
\]

The computed values for each ligand are given in Table 4. The simulation represents a myeloma-dysregulated bone remodeling event taking place over 75 days. The results are shown in Figures 11, 12, and 13.

Figure 11 gives the bone cell counts and bone mass percentage for the simulated bone remodeling event. Figure 11(a) shows the dynamics of the stromal cell population at position \(x = 0\). The dynamics of this population at other positions are similar to those shown in Figure 11(a). Throughout the remodeling event, we see an increase in the number of stromal cells. Figure 11(b) shows the dynamics of the multiple myeloma tumor cell population. For the first fifty days of the bone remodeling event, there is no significant change in the number of tumor cells. However, in the last twenty-five days of the event we see an increase in the tumor cell population due to the multiple myeloma “vicious cycle.” Figures 11(c) and 11(d) show the dynamics of the osteoclast precursor and mature osteoclast cell populations. Both populations decrease as the remodeling event continues. Figures 11(e) and 11(f) show the dynamics of the osteoblast precursor and mature osteoblast cell populations. The osteoblast precursor population decreases in size quickly as osteoblast precursors are recruited to the mature osteoblast population. Figure 11(g) shows the dynamics of the osteocyte population. There is an initial decrease in the number of osteocytes due to the initiation of the bone remodeling event. However, as the event continues, the number of osteocytes begins to increase due to the creation of new bone. Figure 11(h) shows the percentage of bone mass throughout the bone remodeling event. As the remodeling event progresses and the tumor cell population grows, the bone mass percentage decreases.

Figure 12 shows the diffusion of ligands across the marrow. Figure 12(a) shows the diffusion of BAFF and APRIL from the osteoclasts to the tumor cells. Figure 12(b) shows the diffusion of MIP-1\(\alpha\), IL-3, and TNF\(\alpha\) from the tumor cells to the osteoclasts. Figure 12(c) shows the diffusion of DKK1, IL-3, sclerotin,
and sFRPs from the tumor cells to the osteoblasts. Figure 12(d) shows the diffusion of IL-6, RANKL, GFs, and Activin A from the stromal cells to the osteoclasts. Figure 12(e) shows the diffusion of Activin A from the stromal cells to the osteoblasts.

Figure 13 shows the movement of the bone/marrow interface and the marrow/tumor interface. At time $t = 0$, the bone/marrow interface is at $x = -1$ and the marrow/tumor interface is at $x = -1$. As time progresses, the bone recedes and the tumor grows. At time $t = 75$, the bone/marrow interface is at $x = -1.2999$ and the marrow/tumor interface is at $x = 0.5820$.

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Wang Y, Pivonka P, Buenzli PR, Smith DW, Dunstan CR (2011) Computational modeling of interactions between multiple myeloma and the bone microenvironment. PloS one 6(11):e27,494, DOI 10.1371/journal.pone.0027494, URL http://journals.plos.org/plosone/article?id=10.1371/journal.pone.0027494
Figure 1: Diagram of the chemical signals between osteoclasts and osteoblasts, as described by Komarova et al. (2003). The parameters are also taken from Komarova et al. (2003): $g_{11}$ (autocrine promotion of osteoclasts), $g_{12}$ (paracrine promotion of osteoclasts), $g_{21}$ (paracrine inhibition of osteoclasts), and $g_{22}$ (autocrine promotion of osteoblasts).

| Symbol | Definition |
|--------|------------|
| $C(t)$ | Number of osteoclasts at time $t$ |
| $B(t)$ | Number of osteoblasts at time $t$ |
| $\bar{C}$ | Number of osteoclasts in the steady-state |
| $\bar{B}$ | Number of osteoblasts in the steady-state |
| $z(t)$ | Percentage of bone mass at time $t$ |
| $T(t)$ | Number of tumor cells at time $t$ |
| $L_T$ | Maximum tumor size |
| $\gamma_T$ | Tumor growth constant |
| $V_1(t), V_2(t)$ | Treatment functions |

Table 1: Definitions of symbols used in Section 2

| Symbol | Definition |
|--------|------------|
| $Y(t)$ | Number of osteocytes at time $t$ |
| $B_P(t)$ | Number of osteoblast precursors at time $t$ |
| $B(t)$ | Number of osteoblasts at time $t$ |
| $C(t)$ | Number of osteoclasts at time $t$ |
| $z(t)$ | Percentage of bone mass at time $t$ |
| $K_Y$ | Osteocyte population threshold for sclerostin production |

Table 2: Definitions of symbols used in Section 3
Figure 2: Simulation of a healthy bone remodeling event (Equations 1-3) using the following parameter values: $\alpha_1 = 3, \alpha_2 = 4, \beta_1 = 0.2, \beta_2 = 0.02, g_{11} = 0.5, g_{12} = 1, g_{21} = -0.5, g_{22} = 0, k_1 = 0.24, k_2 = 0.0017$. The simulation was completed with MATLAB’s ode15s with initial conditions $C(0) = 15, B(0) = 316, z(0) = 100$ (Komarova et al. 2003).

| Symbol | Definition |
|--------|------------|
| $S(t)$ | Number of stromal cells at time $t$ |
| $T(t)$ | Number of tumor cells at time $t$ |
| $C_P(t)$ | Number of osteoclast precursors at time $t$ |
| $C(t)$ | Number of osteoclasts at time $t$ |
| $B_P(t)$ | Number of osteoblast precursors at time $t$ |
| $B(t)$ | Number of osteoblasts at time $t$ |
| $Y(t)$ | Number of osteocytes at time $t$ |
| $L_C(t)$ | BAFF and APRIL, diffusing from the osteoclasts to the tumor cells |
| $L_{T_1}(t)$ | MIP-1$\alpha$, IL-3, and TNF$\alpha$, diffusing from the tumor cells to the osteoclasts |
| $L_{T_2}(t)$ | DKK1, IL-3, sclerostin, and sFRPs, diffusing from the tumor cells to the osteoblasts |
| $L_{S_1}(t)$ | IL-6, RANKL, GFs, and Activin A, diffusing from the stromal cells to the osteoclasts |
| $L_{S_2}(t)$ | Activin A, diffusing from the stromal cells to the osteoblasts |
| $z(t)$ | Percentage of bone mass at time $t$ |
| $\ell(t)$ | Position of the bone/marrow interface at time $t$ |
| $r(t)$ | Position of the marrow/tumor interface at time $t$ |
| $K_Y$ | Osteocyte population threshold for sclerostin production |

Table 3: Definitions of symbols used in Section 4
Figure 3: Diagram of the chemical signals between osteoclasts, osteoblasts, and myeloma tumor cells, as described by Ayati et al (2010). The parameters $g_{11}$, $g_{12}$, $g_{21}$, and $g_{22}$ are as in Figure 1. Arrow (i) represents the suppression of osteoblast production by myeloma tumor cells. Arrow (ii) represents the increased osteoclast production and activity levels resulting from tumor signaling. Arrow (iii) represents the increased tumor growth resulting from osteoclast activity. Together, arrows (ii) and (iii) comprise the multiple myeloma “vicious cycle”
Figure 4: Simulation of a bone remodeling event with the presence of multiple myeloma tumor cells (Equations 4-7) with the following parameter values: $\alpha_1 = 3$, $\alpha_2 = 4$, $\beta_1 = 0.2$, $\beta_2 = 0.02$, $g_{11} = 1.1$, $g_{12} = 1$, $g_{21} = -0.5$, $g_{22} = 0$, $k_1 = 0.0748$, $k_2 = 0.0006395$, $\gamma_T = 0.005$, $L_T = 100$, $r_{11} = 0.005$, $r_{21} = 0$, $r_{12} = 0$, and $r_{22} = 0.2$. The simulation was completed with MATLAB’s ode23t with initial conditions $C(0) = 15$, $B(0) = 316$, $z(0) = 100$, and $T(0) = 1$ (Ayati et al. 2010).
Figure 5: Simulation of a bone remodeling event with the presence of multiple myeloma tumor cells and treatment (Equations 8-11) with the following parameter values: $\alpha_1 = 3$, $\alpha_2 = 4$, $\beta_1 = 0.2$, $\beta_2 = 0.02$, $g_{11} = 1.1$, $g_{12} = 1$, $g_{21} = -0.5$, $g_{22} = 0$, $k_1 = 0.0748$, $k_2 = 0.0006395$, $\gamma_T = 0.005$, $L_T = 100$, $r_{11} = 0.005$, $r_{21} = 0$, $r_{12} = 0$, $r_{22} = 0.2$, $t_{\text{start}} = 600$, $v_1 = 0.001$, and $v_2 = 0.008$. The simulation was completed with MATLAB’s ode15s with initial conditions $C(0) = 13$, $B(0) = 300$, $z(0) = 100$, and $T(0) = 1$. The steady states are taken to be $C = 5$ and $B = 316$ (Ayati et al. 2010).

Figure 6: Wiring Diagram used by Graham et al. (2013)
Figure 7: Population dynamics during a bone remodeling event (without the presence of a tumor), as simulated by [Graham et al. (2013)].
Figure 8: Bone volume dynamics during a bone remodeling event (without the presence of a tumor), as simulated by Graham et al. (2013).
Figure 9: Diagram of the bone marrow microenvironment. A section of the bone and a multiple myeloma tumor are separated by the marrow. A remodeling site (with osteoclasts, osteoblasts, and osteocytes) is located on the edge of the bone.
Figure 10: Wiring diagram used in Equations 17-31

- Stromal cell
- Myeloma cell
- Osteoblast precursor
- Osteoblast
- Osteoclast precursor
- Osteoclast
- Bone matrix

Key nodes and interactions:
1. M-CSF, TNF
2. IL-6, RANKL, GFRs
3. Activin A
4. VCAM-1
5. Apoptosis
6. OPG
7. RANK
8. BMPs
9. Apoptosis
10. Sclerostin
11. Apoptosis
12. Activin A
13. IGF-1
14. BMPs
15. Sclerostin
16. IL-6, GFRs
17. Resorption
18. RANKL
19. IL-11
20. Osteoblast
21. Sclerostin

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Figure 10: Wiring diagram used in Equations 17-31
Figure 11: Computational results for Equations 17 - 31. This simulation represents a myeloma-dysregulated remodeling event taking place over 75 days.
Figure 12: Computational results for Equations 17 - 31. This simulation represents a myeloma-dysregulated remodeling event taking place over 75 days.
Figure 13: Computational results for Equations [17, 31]. This simulation represents a myeloma-dysregulated remodeling event taking place over 75 days.
| Parameter | Value  | Parameter | Value |
|-----------|--------|-----------|-------|
| $\alpha_1$ | 0.005  | $K_Y$     | 10    |
| $\alpha_2$ | 0.0003 | $\epsilon$ | 1     |
| $\alpha_3$ | 0.0001 | $g_{11}$ | 1     |
| $\alpha_4$ | 0.01   | $g_{12}$ | 1     |
| $\alpha_5$ | 0.01   | $g_{21}$ | 1     |
| $\alpha_6$ | 0.01   | $g_{22}$ | 1     |
| $\beta_1$  | 0.01   | $g_{31}$ | 1     |
| $\beta_2$  | 0.008  | $g_{32}$ | 1     |
| $\beta_3$  | 0.01   | $g_{41}$ | 1     |
| $\beta_4$  | 0.01   | $g_{51}$ | 1     |
| $\beta_5$  | 0.01   | $g_{52}$ | 1     |
| $\beta_6$  | 0.01   | $h_{11}$ | -1    |
| $\beta_7$  | 0.1    | $h_{12}$ | 1     |
| $\gamma_1$ | 0.01   | $h_{13}$ | 1     |
| $\gamma_2$ | 38.4   | $h_{14}$ | 1     |
| $\gamma_3$ | 0.00390625 | $h_{15}$ | 1     |
| $\delta_{11}$ | 0.1037 | $h_{21}$ | 1     |
| $\delta_{12}$ | 0.01   | $h_{22}$ | -0.8  |
| $\delta_{21}$ | 0.1210 | $h_{23}$ | -0.8  |
| $\delta_{22}$ | 0.01   | $h_{24}$ | -0.8  |
| $\delta_{31}$ | 0.1037 | $h_{25}$ | -0.8  |
| $\delta_{32}$ | 0.01   | $h_{26}$ | 1     |
| $\delta_{41}$ | 0.1728 | $h_{31}$ | 1     |
| $\delta_{42}$ | 0.01   | $h_{32}$ | 0.5   |
| $\delta_{51}$ | 0.1063 | $f_{21}$ | 0.65  |
| $\delta_{52}$ | 0.01   | $a$ | 0.01  |
| $k_1$    | 0.1    | $b$ | -0.03 |
| $k_2$    | 0.01   |       |       |

Table 4: Parameter values used in Section 4
| Peptide   | Molecular Weight (kDa) | Diffusion Coefficient (Literature) (cm²/sec) | Stokes Radius (nm) | Diffusion Coefficient (Calculated) (cm²/sec) |
|-----------|------------------------|---------------------------------------------|--------------------|---------------------------------------------|
| IL-2      | 15.1                   | $1.0 \times 10^{-6}$                        | 1.602              | $1.36 \times 10^{-6}$                       |
| IL-3      | 16.2                   | $1.0 \times 10^{-6}$                        | 1.6328             | $1.35 \times 10^{-6}$                       |
| IL-6      | 23.7                   | $9.0 \times 10^{-7}$                        | 1.9628             | $1.21 \times 10^{-6}$                       |
| RANKL     | 35                     |                                             | 2.073              | $1.17 \times 10^{-6}$                       |
| OPG       | 60                     |                                             | 2.463              | $1.01 \times 10^{-6}$                       |
| BAFF      | 31                     |                                             | 2.0106             | $1.20 \times 10^{-6}$                       |
| APRIL     | 28                     |                                             | 1.9638             | $1.21 \times 10^{-6}$                       |
| MIP-1α    | 8                      |                                             | 1.272              | $1.49 \times 10^{-6}$                       |
| TNF       | 17                     |                                             | 1.668              | $1.33 \times 10^{-6}$                       |
| DKK-1     | 26                     |                                             | 1.9326             | $1.23 \times 10^{-6}$                       |
| Sclerostin| 23                     | $9.0 \times 10^{-7}$                        | 1.932              | $1.23 \times 10^{-6}$                       |
| sFRP-1    | 33                     |                                             | 2.0418             | $1.18 \times 10^{-6}$                       |
| GFs       |                        |                                             |                    | $2.00 \times 10^{-6}$                       |
| Activin A | 26.2                   | $6.9 \times 10^{-7}$                        | 1.93572            | $1.23 \times 10^{-6}$                       |
| Hemoglobin| 68                     | $6.9 \times 10^{-7}$                        |                    |                                             |

Table 5: Table of known and calculated diffusion coefficients.