# MODELING OF THE KINETICS OF VITAMIN $D_{3}$ IN OSTEOBLASTIC CELLS 

Robert P. Gilbert, Philippe Guyenne and Ying Liu<br>Department of Mathematical Sciences<br>University of Delaware<br>Newark, DE 19716, USA

(Communicated by Stephen Cantrell)


#### Abstract

A one-dimensional model for the transport of vitamin $D_{3}$ in an osteoblast cell is proposed, from its entry through the membrane to its activation of RANKL synthesis in the nucleus. In the membrane and cytoplasm, the transport of $\mathrm{D}_{3}$ and RANKL is described by a diffusion process, while their interaction in the nucleus is modeled by a reaction-diffusion process. For the latter, an integral equation involving the boundary conditions, as well as an asymptotic solution in the regime of small concentrations, are derived. Numerical simulations are also performed to investigate the kinetics of $D_{3}$ and RANKL through the entire cell. Comparison between the asymptotics and numerics in the nucleus shows an excellent agreement. To our knowledge, this is the first time, albeit using a simple model, a description of the complete passage of $D_{3}$ through the cell membrane, the cytoplasm, into the cell nucleus, and finally the production of RANKL with its passage to the exterior of the cell, has been modeled.


1. Introduction. Bone is a composite structure of living cells embodied in an organic, highly mineralized matrix. Bone is built during body growth by bonebuilding cells, called osteoblasts, that synthesize and release organic molecules, largely collagen. This constitutes the matrix onto which a variety of calcium salts are deposited. The mature calcium salt is carbanato-hydroxyapatite. Bone is resorbed by osteoclasts, cells that form a podosome, a tent-like membrane cover over the bone surface into the space of which they secrete enzymes and hydrogen ions, the combined action of which leads to mineral dissolution and matrix destruction. Adult bone undergoes remodeling, i.e. resorption by osteoclasts, followed by renewed matrix synthesis and mineral deposition initiated by osteoblasts. Two types of bone mineral structure are known: trabecular and compact. Trabecular bone, approximately $20 \%$ of total adult bone, has struts and undergoes renewal at about twice the rate of compact bone. It is trabecular bone that is primarily destroyed in postmenopausal osteoporosis. Total bone renewal, i.e. trabecular and compact averaged out, approximates $3.6 \%$ per year.

Bone formation is tightly regulated by the cells present inside the bone. Especially two key players, osteoclasts (bone resorbing cells) and osteoblasts (bone forming cells), tightly regulate the amount of bone matrix. The balance between

[^0]active osteoblasts and osteoclasts is crucial for proper bone formation and maintenance. 1,25-Dihydroxyvitamin $\mathrm{D}_{3}$, the active form of vitamin $\mathrm{D}_{3}$, plays a critical role during bone formation. As illustrated in Figure 1, it regulates the differentiation of hematopoietic stem cells and preosteoblasts into osteoclast progenitor cells and mature osteoblasts (see [3] for further details). It also induces osteoclast differentiation from the osteoclast progenitor cell to the prosteoclast by affecting the RANKL/RANK/OPG axis, by upregulation of RANKL. This enables the binding of RANKL to RANK leading to the differentiation of the osteoclast progenitor. 1,25 -Dihydroxyvitamin $\mathrm{D}_{3}$ also stimulates the secretion of collagenase, osteopontin, C3, MGP and plasminogen, which in turn affect osteoclast differentiation.

In this paper, we propose a reasonably complete biological model of the preosteoblastic cell and a simplified mathematical version for the transport of vitamin $D_{3}$ from its entry through the membrane surface to its activation of RANKL synthesis in the nucleus. We indicate where we deviate from the biological system in this initial attempt and where we might provide a more complete mathematical version. The outgoing transport of RANKL from the nucleus to the membrane is also described. Assuming spherical symmetry, the cell is defined by a one-dimensional domain divided into three regions: the nucleus, the cytoplasm and the membrane. In each region, the transport of $\mathrm{D}_{3}$ and RANKL is modeled by a diffusion process with distinct properties. For the interaction problem in the nucleus, we derive a simple integral equation involving the boundary conditions. We also identify an asymptotic regime of parameters that allows us to linearize the nonlinear partial differential equations of the nuclear model and derive an approximate analytical solution. For the kinetics through the entire cell, we perform numerical simulations based on a finite-difference scheme, which show that the various transport and synthesis processes are well reproduced in the context of our model. The asymptotic solution for the nucleus is also compared with the numerical solution, and a very good agreement is found.

The remainder of the paper is organized as follows. In Section 2, we describe the mathematical model for the kinetics of $\mathrm{D}_{3}$ and RANKL in an osteoblast cell, including the diffusion equations in the membrane and cytoplasm, and the reactiondiffusion equations in the nucleus. The integral equation together with the asymptotic solution of the nuclear model are derived in Section 3, and details are given in the Appendix. Numerical simulations of the full cellular model are presented in Section 4, including a description of the numerical methods and a discussion of the numerical results. Finally, concluding remarks are given in Section 5.
2. Mathematical model. Typically, the vitamin D serum binding proteins, exterior to the cell, are present at high concentrations and have high off rates relative to the membrane receptors, $\mathrm{VDR}_{m},{ }^{1}$ that are present at much lower concentrations but have higher binding affinities. These receptors are fast (seconds-minutes) and mediate catabolic effects. There is also a nuclear receptor, $\mathrm{VDR}_{n}$, in osteoblast cells that responds to $\mathrm{D}_{3}$, which is a slow (hours), non-calcemic receptor mediating anabolic effects.

Vitamin $\mathrm{D}_{3}$ dissociates from the serum binding vitamin $\mathrm{D}_{3}$ protein (DBP) to the membrane receptor, acting as a molecular switch for the activation of $\mathrm{VDR}_{n}$. This

[^1]then activates responsive genes. In the model, for simplicity, we choose $\mathrm{VDR}_{n}$ to be located in the nucleus rather than in the cytoplasm.
$\mathrm{VDR}_{m}$ then signals $\mathrm{VDR}_{n}$ to shut down the production of OPG and promotes transcription of the gene encoding receptor activator of RANKL [23]. Osteoclastogenesis can be inhibited by osteoprotegerin, OPG or RANKL, in their capacity acting as a decoy receptor for RANKL [23]. Shutting down OPG is essential to decreasing the maturation and metabolic effects of osteoclasts, resulting in a decrease of bone resorption. The RNA encoding RANKL or OPG is exported to the cytoplasm where it is translated into protein on the ribosome. The new protein is then inserted into the secretory pathway for export. The mechanism by which RANKL is converted from a membrane to a soluble structure is unclear.

### 2.1. Assumptions concerning the model.

- The process of serum binding proteins passing along the hormone from one to the next through a sequence of low affinity binding and release events ${ }^{2}$ (the buckets) until reaching the target (the receptor) is modelled by a sequence of short distance diffusional events with a pause in between each. The length of the pause (binding event) determines the rate of the movement of the diffusing molecule. When the molecule binds to the receptor, the number of molecules that enters is determined by the number of liganded receptors. In this paper we eliminated this step as the mathematical model was sufficiently complicated for an initial attempt. It will be included, however in subsequent models.
- When $\mathrm{D}_{3}$ interacts with MARRS, the membrane diffusion coefficient changes from a high off-rate to a lower value that allows a specific amount of $D_{3}$ to cross the membrane. Once transfer is completed, the diffusion coefficient reverts to the high off-value and transmembrane transfer is turned off.
- The saturation curve for the membrane receptor is biphasic.This means two separate events may be initiated, such as opening a calcium channel and permitting $\mathrm{D}_{3}$ to pass through the membrane. Calcium activates and, at greater concentration, deactivates transcription. These events are symbolized by changes in the kinetic coefficients.
- Calcium entering the cells initiates cross talk between the vesicles and the nuclear receptor via signal transduction pathways. We assume that this switching mechanism is located in the nuclear receptor, as nuclear receptor regulated genes still turn on in 24 hrs when the membrane system is blocked. This is modeled using a threshold concerning the number of occupied nuclear receptors.
- The translocation of vitamin $\mathrm{D}_{3}$ to the nucleus occurs using a chaperon, DBP, that moves along cytoskeletal tracks through the cytoplasm [17] and then diffuses into the nucleus to reach the $\mathrm{VDR}_{n}$. However, in the present model this is described by a diffusion process.
- The binding with $\mathrm{VDR}_{n}$ is described by a reaction-diffusion equation that is initiated by the signal cross-talk with MARRS. Signaling from the membrane receptor to the nuclear receptor depends on a threshold effect that induces a change in the kinetics of the nuclear response, with a time lag before the signal is received. In the present mathematical model the reaction-diffusion

[^2]equation becomes valid upon the arrival of the $D_{3}$. This is appropriate as we use a diffusion model in the cytoplasm rather than transportation along microtubules.

- $\mathrm{D}_{3}$ combines with the nuclear receptor and diffuses to the nucleus where it initiates, by a hormone-like action, the target gene response. The effect of this response may be slow, requiring hours.
- Biosynthesis of soluble RANKL is a multistep process, but will be modeled as a single step. RANKL is a soluble decoy that is assumed to diffuse through the nuclear space into the cytoplasm from which it exits through the membrane into the body fluids.
2.2. Membrane-cytoplasm model. Assume that the cell is a small ball with radius $r_{m}^{e}>0$ (Figure 2), the cell membrane is a spherical shell, and also assume that the diffusion of vitamin $\mathrm{D}_{3}\left(1,25(\mathrm{OH})_{2} \mathrm{D}_{3}\right)$ from the membrane to the nucleus is independent of the spherical angles $(\theta, \phi)$. Then we can model the interaction of the $1,25(\mathrm{OH})_{2} \mathrm{D}_{3}$ and the receptors by a system of partial differential equations in one space dimension.

In the following, we represent the membrane receptor, $\mathrm{VDR}_{m}$, by $M$ and the compound formed from $\mathrm{D}_{3}$ and $\mathrm{VDR}_{m}$ by $D * M$. The genes for both OPG and RANKL are regulated by the nuclear receptor for $D_{3}$ and this is described in a subsequent section. It is proposed that the $\mathrm{VDR}_{m}$ complex that regulates calcium signaling also regulates the intake of $\mathrm{D}_{3}$ into the cell.

One possibility is to model the entry of $\mathrm{D}_{3}$ into the cell by having the diffusion coefficient for the membrane dependent on the concentration of $D * M$. This is a transient phenomenon, namely increasing $\mathrm{D}_{3}$ in the serum suddenly above its basal concentration leads MARRS to signal ER, allowing an increased flow of $\mathrm{Ca}^{2+}$ into the cell. This in turn signals a temporary change in the membrane diffusion coefficient

$$
\kappa_{m}(D * M)= \begin{cases}\kappa_{m}^{H} & \text { for high intake } \\ \kappa_{m}^{L} & \text { for low intake }\end{cases}
$$

which then after a fixed time reverts to its non-entry state. Another possibility is to have $\mathrm{D}_{3}$ collect in caveolae and have MARRS using $\mathrm{Ca}^{2+}$ to signal dynamin to snip the vesicle containing $D_{3}$. At this point, it is unknown exactly how $D_{3}$ enters the cell. We assume that MARRS signals the nuclear receptor using $\mathrm{Ca}^{2+}$, which also signals the entry of $\mathrm{D}_{3}$. Hence, in some way, the concentration of $\mathrm{Ca}^{2+}$ is tied to the entry of $D_{3}$ into the cell. We make the simplifying assumption that the boundary condition for $\mathrm{D}_{3}$ is proportional to $\mathrm{Ca}^{2+}$. We use the single-pool model for intracellular $\mathrm{Ca}^{2+}$ from [1] to describe the calcium spiking. These spikes, when they become large enough, switch the membrane diffusion coefficient from off to on, etc. In this paper, we simplify the mechanism by assuming the flow of $D_{3}$ into the membrane is proportional to that of $\mathrm{Ca}^{2+}$.

In the membrane, we represent the concentration of $\mathrm{D}_{3}$ by the function $D_{m}(r, t)$. We model the transport of $\mathrm{D}_{3}$ through the membrane of the cell by a diffusion process where the diffusion coefficient is chosen to match the time scale necessary to pass the correct amount through the membrane during the time it is open. After this time, the diffusion coefficient changes to permit essentially no $D_{3}$ to pass through the membrane. As mentioned in the previous section, this is controlled by MARRS. In this section, we provide the analysis for the case where we either increase or decrease the concentration of $D_{3}$ in the serum. When the membrane diffusion
coefficient has switched to its higher value $\kappa_{m}^{H}, D_{m}(r, t)$ satisfies the equation

$$
\begin{equation*}
\kappa_{m} \frac{\partial^{2} D_{m}}{\partial r^{2}}=\frac{\partial D_{m}}{\partial t}, \quad r_{m}^{i}<r<r_{m}^{e} \tag{2.1}
\end{equation*}
$$

In this paper, we assume that $\mathrm{D}_{3}$ obeys a diffusion law in the cytoplasm rather than being transported along microtubules in the cytoplasm, namely

$$
\kappa_{c} \frac{\partial^{2} D_{c}}{\partial r^{2}}=\frac{\partial D_{c}}{\partial t}, \quad r_{n}<r<r_{m}^{i}
$$

Accurate measurements of $\mathrm{D}_{3}$ transport in osteoblast cells are usually difficult to obtain. According to [18], the effective diffusion constant might be described by a two-parameter power law. In [22], transport along microtubules or actin filaments is compared. The microtubules are for long-range transport whereas the actin filaments are for local movement of organelles. This is a possibility we might try in the future; however for the present, we use diffusion transport, estimating $\kappa_{c}$ by Einstein's formula

$$
\begin{equation*}
\kappa_{c}=\frac{k_{B} T}{6 \pi \eta r_{D}} \tag{2.2}
\end{equation*}
$$

where $k_{B}$ is Boltzmann's constant, $\eta$ the dynamic viscosity of water and $T=309$ K (internal body temperature). Assuming spherical symmetry for the $D$ molecule, its radius is given by

$$
\begin{align*}
r_{D} & =0.066 \mathrm{~m}_{D}^{1 / 3}  \tag{2.3}\\
& =4.9294 \times 10^{-10} \mathrm{~m}=0.4929 \mathrm{~nm}
\end{align*}
$$

where $m_{D}=416.64 \mathrm{~g} \mathrm{~mol}^{-1}$ is the molecular mass of $1,25(\mathrm{OH})_{2} \mathrm{D}_{3}$. As discussed above, because diffusion is strong in the membrane during the $\mathrm{Ca}^{2+}$ influx, we choose $\kappa_{m}=10 \kappa_{c}$.

We match the membrane and cytoplasm solutions at the inner membrane surface $r=r_{m}^{i}$, using the transmission conditions:

$$
\begin{equation*}
D_{c}\left(r_{m}^{i,-}, t\right)=D_{m}\left(r_{m}^{i,+}, t\right), \quad \kappa_{c} \frac{\partial D_{c}}{\partial r}\left(r_{m}^{i,-}, t\right)=\kappa_{m} \frac{\partial D_{m}}{\partial r}\left(r_{m}^{i,+}, t\right) \tag{2.4}
\end{equation*}
$$

At the external membrane surface $r=r_{m}^{e}$, we assume the effective concentration is proportional to the $\mathrm{Ca}^{2+}$ spikes as described in [1]. These $\mathrm{Ca}^{2+}$ spikes obey the system of ordinary differential equations

$$
\begin{align*}
\frac{d D_{m}}{d t} & =J_{\text {channel }}-J_{\text {pump }}+J_{l e a k}  \tag{2.5}\\
\tau_{n} \frac{d n}{d t} & =n_{\infty}\left(D_{m}\right)-n \tag{2.6}
\end{align*}
$$

where

$$
\begin{align*}
J_{\text {channel }} & =k_{\text {flux }} \mu\left(\left[I P_{3}\right]\right) n\left(b+\frac{V_{1} D_{m}}{k_{1}+D_{m}}\right) \\
J_{\text {pump }} & =\frac{\gamma D_{m}}{k_{\gamma}+D_{m}}, \\
J_{\text {leak }} & =\beta \\
n_{\infty}\left(D_{m}\right) & =1-\frac{D_{m}^{2}}{k_{2}^{2}+D_{m}^{2}}, \\
\mu\left(\left[I P_{3}\right]\right) & =\mu_{0}+\frac{\mu_{1}\left[I P_{3}\right]}{k_{\mu}+\left[I P_{3}\right]} . \tag{2.7}
\end{align*}
$$

| Parameter | Value |
| :---: | :---: |
| $b$ | 0.111 |
| $V_{1}$ | 0.889 |
| $\beta$ | $0-0.02 \mu \mathrm{M} \mathrm{s}^{-1}$ |
| $\gamma$ | $2.0 \mu \mathrm{M} \mathrm{s}^{-1}$ |
| $\tau_{n}$ | 2.0 s |
| $k_{1}$ | $0.7 \mu \mathrm{M} \mathrm{s}$ |
| $k_{\gamma}$ | $0.1 \mu \mathrm{M}$ |
| $k_{2}$ | $0.7 \mu \mathrm{M}$ |
| $k_{\text {flux }}$ | $8.1 \mu \mathrm{M} \mathrm{s}$ |

TABLE 1. Parameter values in the boundary conditions (2.5)-(2.7) at the external membrane surface. $b$ represents a basal current through the $\mathrm{Ca}^{2+}$ channel. $V_{1}$ is the proportion of $\mathrm{IP}_{3} \mathrm{Rs}$ that are activated by the binding of $\mathrm{Ca}^{2+} . \beta$ is the constant rate of $\mathrm{Ca}^{2+}$ influx into the cytosol. $\gamma$ is the maximum rate of $\mathrm{Ca}^{2+}$ pumping from the cytosol. $k_{\gamma}$ is the concentration of $\mathrm{Ca}^{2+}$ at half-maximum pumping. $\tau_{n}$ is the time constant for the dynamics of $n . k_{\text {flux }}$ is the maximum total $\mathrm{Ca}^{2+}$ flux through all $\mathrm{IP}_{3}$ Rs. Further details can be found in [1].

Each $J$ term represents a concentration flux, $n$ is the dimensionless variable representing the proportion of receptor $\mathrm{IP}_{3} \mathrm{Rs}$ that have been filled by $\mathrm{Ca}^{2+}, n_{\infty}$ denotes the steady state of $n$ as a function of $D_{m}, \mu\left(\left[I P_{3}\right]\right)$ is the proportion of $\mathrm{IP}_{3} \mathrm{Rs}$ that have their $\mathrm{IP}_{3}$ binding domain activated, and $b+V_{1}=1$. The values of the constants in (2.6)-(2.7) are taken from [1] and listed in Table 1. A numerical simulation of (2.5)-(2.7), starting from zero initial conditions and using the 4th-order Runge-Kutta method, is depicted in Figure 3.
2.3. Nuclear model. It is assumed that the nucleus is a smaller concentric ball of radius $r_{n}$. Let $D_{n}$ denote the concentration of $D_{3}$ in the nucleus, $N$ the concentration of $\mathrm{VDR}_{n}$ and $V_{n}=D_{n} * N$ the compound formed with the nuclear receptor. Their kinetics may be described by the reaction diagram

> influx


Considering again isotropy, we model their dynamics as a reaction-diffusion process of the form

$$
\begin{align*}
\frac{\partial D_{n}}{\partial t} & =\kappa_{n} \frac{\partial^{2} D_{n}}{\partial r^{2}}+k_{4} V_{n}-k_{3}\left(R_{N}-V_{n}\right) D_{n}  \tag{2.8}\\
\frac{\partial V_{n}}{\partial t} & =\kappa_{v n} \frac{\partial^{2} V_{n}}{\partial r^{2}}-k_{4} V_{n}+k_{3}\left(R_{N}-V_{n}\right) D_{n} \tag{2.9}
\end{align*}
$$

| Parameter | Value |
| :---: | :---: |
| $k_{3}$ | $6.7 \times 10^{-9} \mathrm{pM}^{-1} \mathrm{~s}^{-1}$ |
| $k_{4}$ | $1.96 \times 10^{-7} \mathrm{~s}^{-1}$ |
| $R_{N}$ | 10 pM |

Table 2. Parameter values in the nuclear model (2.8)-(2.9). $k_{3}$ is the rate of RANK-RANKL binding. $k_{4}$ is the rate of RANKRANKL unbinding. $R_{N}$ is a fixed concentration of RANK. Further details can be found in [13].
where $k_{3}, k_{4}$ and $R_{N}$ are taken from [13]. The values of these parameters, together with their description, are given in Table 2. We choose $\kappa_{n}=0.1 \kappa_{c}$ to allow for a sufficient level of $D$ in the nucleus so it can produce enough $V$ to diffuse out, in a reasonable amount of time.

At the surface $r=r_{n}$ of the nucleus, we have the transmission conditions

$$
D_{n}\left(r_{n}^{-}, t\right)=D_{c}\left(r_{n}^{+}, t\right), \quad \kappa_{n} \frac{\partial D_{n}}{\partial r}\left(r_{n}^{-}, t\right)=\kappa_{c} \frac{\partial D_{c}}{\partial r}\left(r_{n}^{+}, t\right)
$$

and, at the center $r=0$, we impose the reflecting condition

$$
\kappa_{n} \frac{\partial D_{n}}{\partial r}(0, t)=\kappa_{v n} \frac{\partial V_{n}}{\partial r}(0, t)=0
$$

which is a suitable choice of boundary condition at this location given the spherical symmetry adopted in the problem.

The initial conditions are simply

$$
\begin{equation*}
D_{n}(r, 0)=D_{c}(r, 0)=D_{m}(r, 0)=0, \quad 0<r<r_{m}^{e} \tag{2.10}
\end{equation*}
$$

2.4. Outgoing compound. In the previous section, we define the compound RANKL formed with the nuclear receptor by $V_{n}:=D_{n} * N$. Assume that $D_{n}$ combines with the nuclear receptor and diffuses to the cytoplasm when it initiates. Let $V_{c}$ denote the concentration of RANKL in the cytoplasm and $V_{m}$ the concentration in the membrane. This step is modelled as a diffusion process by the following equations:

$$
\kappa_{v c} \frac{\partial^{2} V_{c}}{\partial r^{2}}=\frac{\partial V_{c}}{\partial t}, \quad r_{n}<r<r_{m}^{i}
$$

and

$$
\kappa_{v m} \frac{\partial^{2} V_{m}}{\partial r^{2}}=\frac{\partial V_{m}}{\partial t}, \quad r_{m}^{i}<r<r_{m}^{e}
$$

where $\kappa_{v c}$ and $\kappa_{v m}$ are the diffusion coefficients in the cytoplasm and membrane, respectively. Applying again formula (2.3) to the $V$ molecule, we estimate its radius to be

$$
r_{V}=1.8465 \times 10^{-9} \mathrm{~m}=1.8465 \mathrm{~nm}
$$

which is about four times larger than $r_{D}$, using the molecular mass $m_{V}=21900 \mathrm{~g}$ $\mathrm{mol}^{-1}$ of RANKL. Since the $V$ molecule is bigger than the $D$ molecule, we expect its dispersion to be slower. We thus choose

$$
\kappa_{v n}=\frac{1}{4} \kappa_{n}, \quad \kappa_{v c}=\frac{1}{4} \kappa_{c}, \quad \kappa_{v m}=\frac{1}{4} \kappa_{m},
$$

| Parameter | Value |
| :---: | :---: |
| $\kappa_{n}$ | $6.559 \times 10^{-11} \mathrm{~m}^{2} \mathrm{~s}^{-1}$ |
| $\kappa_{c}$ | $6.559 \times 10^{-10} \mathrm{~m}^{2} \mathrm{~s}^{-1}$ |
| $\kappa_{m}$ | $6.559 \times 10^{-9} \mathrm{~m}^{2} \mathrm{~s}^{-1}$ |
| $\kappa_{v n}$ | $1.639 \times 10^{-11} \mathrm{~m}^{2} \mathrm{~s}^{-1}$ |
| $\kappa_{v c}$ | $1.639 \times 10^{-10} \mathrm{~m}^{2} \mathrm{~s}^{-1}$ |
| $\kappa_{v m}$ | $1.639 \times 10^{-9} \mathrm{~m}^{2} \mathrm{~s}^{-1}$ |

Table 3. Diffusion coefficients for $D$ and $V$ based on Einstein's formula (2.2). We use $k_{B}=1.3806488 \times 10^{-23} \mathrm{~J} \mathrm{~K}^{-1}, T=309 \mathrm{~K}$ and $\eta=0.7 \times 10^{-3} \mathrm{~N} \mathrm{~s} \mathrm{~m}^{-2}$.
because the diffusion coefficient is inversely proportional to the molecular radius by virtue of (2.2). The values of the various diffusion coefficients for $D$ and $V$ are summarized in Table 3.

At the interfaces between the nucleus and cytoplasm, and between the cytoplasm and membrane, we have the following transmission conditions:

$$
\begin{equation*}
V_{n}\left(r_{n}^{-}, t\right)=V_{c}\left(r_{n}^{+}, t\right), \quad \kappa_{v n} \frac{\partial V_{n}}{\partial r}\left(r_{n}^{-}, t\right)=\kappa_{v c} \frac{\partial V_{c}}{\partial r}\left(r_{n}^{+}, t\right) \tag{2.11}
\end{equation*}
$$

and

$$
V_{c}\left(r_{m}^{i,-}, t\right)=V_{m}\left(r_{m}^{i,+}, t\right), \quad \kappa_{v c} \frac{\partial V_{c}}{\partial r}\left(r_{m}^{i,-}, t\right)=\kappa_{v m} \frac{\partial V_{m}}{\partial r}\left(r_{m}^{i,+}, t\right)
$$

At the external boundary of the membrane, we consider the fast diffusion case assuming that there is no build-up, namely

$$
V_{m}\left(r_{m}^{e}, t\right)=0
$$

The initial conditions are given by

$$
\begin{equation*}
V_{n}(r, 0)=V_{c}(r, 0)=V_{m}(r, 0)=0, \quad 0<r<r_{m}^{e} \tag{2.12}
\end{equation*}
$$

3. Asymptotic solution of the nuclear model. The reaction-diffusion model (2.8)-(2.10), (2.11) and (2.12) for $D_{n}$ and $V_{n}$ in the nucleus is a nonlinear system of partial differential equations. In this section, we derive a first-order analytical solution in an asymptotic regime with prescribed fluxes through the surface of the nucleus. We first examine the case of constant fluxes and then extend the solution to time-dependent fluxes.

Consider a closed version of this reaction-diffusion model in the form

$$
\begin{aligned}
\frac{\partial D_{n}}{\partial t} & =\kappa_{n} \frac{\partial^{2} D_{n}}{\partial r^{2}}+k_{4} V_{n}-k_{3}\left(R_{N}-V_{n}\right) D_{n} \\
\frac{\partial V_{n}}{\partial t} & =\kappa_{v n} \frac{\partial^{2} V_{n}}{\partial r^{2}}-k_{4} V_{n}+k_{3}\left(R_{N}-V_{n}\right) D_{n}
\end{aligned}
$$

with initial conditions

$$
\begin{equation*}
D_{n}(r, 0)=V_{n}(r, 0)=0 \tag{3.1}
\end{equation*}
$$

and boundary conditions

$$
\begin{gather*}
\frac{\partial D_{n}}{\partial r}(0, t)=\frac{\partial V_{n}}{\partial r}(0, t)=0  \tag{3.2}\\
\frac{\partial D_{n}}{\partial r}\left(r_{n}, t\right)=g(t), \quad \frac{\partial V_{n}}{\partial r}\left(r_{n}, t\right)=h(t) \tag{3.3}
\end{gather*}
$$

where $g(t)$ and $h(t)$ are prescribed fluxes at $r=r_{n}$, which are assumed to be smooth functions.

The first step is to identify a suitable asymptotic regime with a small parameter. For this purpose, we non-dimensionalize the equations by introducing the dimensionless variables

$$
D^{\prime}=\frac{D_{n}}{\mathcal{D}}, \quad V^{\prime}=\frac{V_{n}}{\mathcal{D}}, \quad t^{\prime}=\frac{t}{\mathcal{T}}, \quad r^{\prime}=\frac{r}{\mathcal{R}}
$$

with the characteristic values

$$
\begin{equation*}
\mathcal{D}=\mathcal{D}_{0} \frac{k_{4}}{k_{3}}, \quad \mathcal{R}=r_{n}, \quad \mathcal{T}=\frac{\mathcal{R}^{2}}{\kappa_{n}} \tag{3.4}
\end{equation*}
$$

where $\mathcal{D}_{0} \ll 1$ is a dimensionless scaling factor. Dropping the primes, the nondimensionalized version of (2.8)-(2.9) reads

$$
\begin{align*}
\frac{\partial D}{\partial t} & =\frac{\partial^{2} D}{\partial r^{2}}-a D+b V+\epsilon D V  \tag{3.5}\\
\frac{\partial V}{\partial t} & =\alpha \frac{\partial^{2} V}{\partial r^{2}}+a D-b V-\epsilon D V \tag{3.6}
\end{align*}
$$

where

$$
\alpha=\frac{\kappa_{v n}}{\kappa_{n}}, \quad a=\frac{R_{N} k_{3} \mathcal{R}^{2}}{\kappa_{n}}, \quad b=\frac{k_{4} \mathcal{R}^{2}}{\kappa_{n}}, \quad \epsilon=\frac{\mathcal{D}_{0} k_{4} \mathcal{R}^{2}}{\kappa_{n}} .
$$

The values of these dimensionless constants are listed in Table 4. Clearly, diffusion is the dominant process for both $D$ and $V$, the other effects being much weaker. We note that $a$ and $b$ are comparable while $\epsilon \ll a, b$ (because $\epsilon=\mathcal{D}_{0} b$ and $\mathcal{D}_{0} \ll 1$ ). This suggests that $\epsilon$ can be used as a small parameter in perturbation calculations, and thus the nonlinear terms in (3.5)-(3.6) can be neglected as compared to the linear terms. We could in fact only retain the diffusive terms and neglect all other terms in good approximation. It turns out however that the linear terms in factor of $a$ and $b$ can be solved exactly together with the diffusive terms, as shown below. Moreover, these linear terms in factor of $a$ and $b$ represent the leading-order contributions to the reaction process, and thus they should be taken into account. The scaling $\mathcal{D}_{0} \ll 1$ may be viewed as a regime of 'small' concentration, which is a reasonable choice given the fact that only a fraction of the vitamin $D_{3}$ serum eventually gets into the nucleus to activate the production of RANKL.

In order to solve the pair of nonlinear partial differential equations (3.5)-(3.6), we need to find a scheme by which to do so. As there is not one in the mathematical literature, we list our results here formally as Theorems (see in particular Theorem 3.1)These results are formal and may be of slight interest to biologists, in which case we urge the reader to pass over them. They are added to the paper, nevertheless, for completeness.
Theorem 3.1. Let $g(t)$ and $h(t)$ in (3.3) be two integrable functions of $t$, and define

$$
E(t)=\int_{0}^{r_{n}}(D+V) d r
$$

Then the solution $(D, V)$ of (3.5)-(3.6) with initial conditions (3.1) and boundary conditions (3.2)-(3.3) satisfies the integral equation

$$
\begin{equation*}
E(t)=\int_{0}^{t}[g(\tau)+\alpha h(\tau)] d \tau \tag{3.7}
\end{equation*}
$$

Proof. The coupling terms can be eliminated by adding (3.5)-(3.6) together, yielding

$$
\frac{\partial D}{\partial t}+\frac{\partial V}{\partial t}=\frac{\partial^{2} D}{\partial r^{2}}+\alpha \frac{\partial^{2} V}{\partial r^{2}}
$$

Then integrating in $r$ and using the boundary conditions (3.2)-(3.3), we obtain

$$
\begin{aligned}
\int_{0}^{r_{n}} \frac{\partial}{\partial t}(D+V) d r & =\int_{0}^{r_{n}}\left(\frac{\partial^{2} D}{\partial r^{2}}+\alpha \frac{\partial^{2} V}{\partial r^{2}}\right) d r \\
& =\left[\frac{\partial D}{\partial r}+\alpha \frac{\partial V}{\partial r}\right]_{0}^{r_{n}} \\
& =g(t)+\alpha h(t)
\end{aligned}
$$

Differentiation in $t$ and integration in $r$ can be interchanged on the left-hand side of the above equation. Finally, integrating in $t$ and using the initial conditions (3.1), we arrive at

$$
\int_{0}^{r_{n}}(D+V) d r=\int_{0}^{t}[g(\tau)+\alpha h(\tau)] d \tau
$$

In the case where $g(t)=h(t)=1$, this equation reduces to

$$
\int_{0}^{r_{n}}(D+V) d r=(1+\alpha) t
$$

which concludes the proof of the theorem.
An asymptotic solution can be found by regular perturbation in the form of a power series in $\epsilon$,

$$
\begin{align*}
D(r, t) & =\sum_{j=0}^{\infty} \epsilon^{j} D^{(j)}(r, t)  \tag{3.8}\\
V(r, t) & =\sum_{j=0}^{\infty} \epsilon^{j} V^{(j)}(r, t) \tag{3.9}
\end{align*}
$$

Plugging (3.8)-(3.9) in (3.5)-(3.6), and in the initial and boundary conditions, the first-order solution $\left(D^{(0)}, V^{(0)}\right)$ satisfies the linear system

$$
\begin{align*}
\frac{\partial D^{(0)}}{\partial t} & =\frac{\partial^{2} D^{(0)}}{\partial r^{2}}-a D^{(0)}+b V^{(0)}  \tag{3.10}\\
\frac{\partial V^{(0)}}{\partial t} & =\alpha \frac{\partial^{2} V^{(0)}}{\partial r^{2}}+a D^{(0)}-b V^{(0)} \tag{3.11}
\end{align*}
$$

with

$$
\begin{equation*}
D^{(0)}(r, 0)=V^{(0)}(r, 0)=0 \tag{3.12}
\end{equation*}
$$

and

$$
\begin{gather*}
\frac{\partial D^{(0)}}{\partial r}(0, t)=\frac{\partial V^{(0)}}{\partial r}(0, t)=0 \\
\frac{\partial D^{(0)}}{\partial r}\left(r_{n}, t\right)=g(t), \quad \frac{\partial V^{(0)}}{\partial r}\left(r_{n}, t\right)=h(t) . \tag{3.13}
\end{gather*}
$$

Note the non-homogeneity in the boundary conditions (3.13). Although the linear system (3.10)-(3.13) is a simplification of the nuclear model, it still exhibits important features of the interaction problem, due to the linear coupling terms. Clearly, the asymptotic solution $\left(D^{(0)}, V^{(0)}\right)$ also satisfies the integral equation (3.7), as can be shown by following the proof of Theorem 3.1.

In the case of constant fluxes at $r=r_{n}$, say $g(t)=h(t)=1$ for simplicity, we have the following result:

| Parameter | Value |
| :---: | :---: |
| $\alpha$ | 0.25 |
| $a$ | $4.08 \times 10^{-7}$ |
| $b$ | $1.19 \times 10^{-6}$ |
| $\epsilon$ | $1.19 \times 10^{-8}$ |

Table 4. Parameter values in the non-dimensionalized nuclear model (3.5)-(3.6). We use $r_{n}=2 \times 10^{-5} \mathrm{~m}$ and $\mathcal{D}_{0}=10^{-2}$.

Theorem 3.2. Let $g(t)=h(t)=1$ in (3.13). Assume $\left(D^{(0)}, V^{(0)}\right)$ is a series solution of the linear problem (3.10)-(3.13). Then this solution can be written as

$$
\begin{equation*}
\binom{D^{(0)}(r, t)}{V^{(0)}(r, t)}=\sum_{j=0}^{\infty} W_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right)+\frac{r^{2}}{2 r_{n}}\binom{1}{1} \tag{3.14}
\end{equation*}
$$

where

$$
W_{0}(t)=\binom{-\left[\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}+\frac{(a-b) r_{n}}{6(a+b)}\right] e^{-(a+b) t}+\frac{b \gamma_{0} t}{6 r_{n}(a+b)^{2}}+\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}-\frac{b r_{n}}{3(a+b)}}{\left[\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}+\frac{(a-b) r_{n}}{6(a+b)}\right] e^{-(a+b) t}+\frac{a \gamma_{0} t}{6 r_{n}(a+b)^{2}}-\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}-\frac{a r_{n}}{3(a+b)}}
$$

with

$$
\beta_{0}=\left(b^{2}-a^{2}\right) r_{n}^{2}+6 a-6 a b, \quad \gamma_{0}=6(a+b)(1+\alpha)
$$

and the higher Fourier coefficients $W_{j}(t)(j>0)$ are given in the Appendix.
It can checked that the series expansion (3.14) of $\left(D^{(0)}, V^{(0)}\right)$ satisfies (3.7) for $g(t)=h(t)=1$. Indeed, since

$$
\int_{0}^{r_{n}} \cos \left(\frac{j \pi r}{r_{n}}\right) d r=0, \quad j>0
$$

we have

$$
\begin{aligned}
\int_{0}^{r_{n}}\left(D^{(0)}+V^{(0)}\right) d r & =\left[\frac{\gamma_{0} t}{6 r_{n}(a+b)}-\frac{r_{n}}{3}\right] \int_{0}^{r_{n}} d r+\int_{0}^{r_{n}} \frac{r^{2}}{r_{n}} d r \\
& =\frac{\gamma_{0} t}{6(a+b)} \\
& =(1+\alpha) t
\end{aligned}
$$

The extension to the more general case of time-dependent fluxes is stated next:
Corollary 1. Let $g(t)$ and $h(t)$ be two smooth functions of $t$. Then a series solution of (3.10)-(3.13) can be written in the form

$$
\binom{D^{(0)}(r, t)}{V^{(0)}(r, t)}=\sum_{j=0}^{\infty} W_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right)+\frac{r^{2}}{2 r_{n}}\binom{g(t)}{h(t)}
$$

where

$$
W_{j}(t)=e^{t A_{j}} \int_{0}^{t} e^{-\tau A_{j}} Y_{j}(\tau) d \tau+e^{t A_{j}} W_{j}(0), \quad j \geq 0
$$

and the coefficients $W_{j}(0)$ and $Y_{j}(t)$ are given in the Appendix.
Proofs: The proofs of the theorem and corollary are lengthy and are therefore relegated to the Appendix.
4. Numerical results for the full model. In this section, we present numerical simulations of the full cellular model to investigate the kinetics of $D_{3}$ from the membrane to the nucleus, together with that of RANKL from the nucleus to the membrane. We also take this opportunity to test and validate the asymptotic solution, derived previously, against the numerical solution in the nucleus.
4.1. Numerical methods. The full cellular model (2.1)-(2.12) is solved numerically by a finite-difference method. Since it describes diffusion-dominated processes, we use a second-order implicit scheme of Crank-Nicolson type. Let

$$
\left(u_{a}\right)_{j}^{\delta}=D_{a}\left(r_{j}, t_{\delta}\right), \quad\left(v_{a}\right)_{j}^{\delta}=V_{a}\left(r_{j}, t_{\delta}\right), \quad a=\{m, c, n\}
$$

where $r_{j}=j \Delta r\left(j=\left\{0, \ldots, J_{a}\right\}\right)$ and $t_{\delta}=\delta \Delta t\left(\delta=\left\{0, \ldots, I_{a}\right\}\right)$.
In the membrane and cytoplasm $(a=\{m, c\})$, the discretized form of the bulk equation reads

$$
\begin{aligned}
\frac{\left(u_{a}\right)_{j}^{\delta+1}-\left(u_{a}\right)_{j}^{\delta}}{\Delta t}= & \kappa_{a}\left[\frac{\left(u_{a}\right)_{j+1}^{\delta+1}-2\left(u_{a}\right)_{j}^{\delta+1}+\left(u_{a}\right)_{j-1}^{\delta+1}}{2(\Delta r)^{2}}\right. \\
& \left.+\frac{\left(u_{a}\right)_{j+1}^{\delta}-2\left(u_{a}\right)_{j}^{\delta}+\left(u_{a}\right)_{j-1}^{\delta}}{2(\Delta r)^{2}}\right]
\end{aligned}
$$

A similar discretized equation holds for $v_{a}$. In the nucleus $(a=n)$, we have

$$
\begin{align*}
\frac{\left(u_{a}\right)_{j}^{\delta+1}-\left(u_{a}\right)_{j}^{\delta}}{\Delta t}= & \kappa_{a}\left[\frac{\left(u_{a}\right)_{j+1}^{\delta+1}-2\left(u_{a}\right)_{j}^{\delta+1}+\left(u_{a}\right)_{j-1}^{\delta+1}}{2(\Delta r)^{2}}\right. \\
& \left.+\frac{\left(u_{a}\right)_{j+1}^{\delta}-2\left(u_{a}\right)_{j}^{\delta}+\left(u_{a}\right)_{j-1}^{\delta}}{2(\Delta r)^{2}}\right] \\
& -k_{3} R_{N}\left(u_{a}\right)_{j}^{\delta+1}+k_{4}\left(v_{a}\right)_{j}^{\delta+1} \\
& +k_{3}\left(u_{a}\right)_{j}^{\delta}\left(v_{a}\right)_{j}^{\delta},  \tag{4.1}\\
\frac{\left(v_{a}\right)_{j}^{\delta+1}-\left(v_{a}\right)_{j}^{\delta}}{\Delta t}= & \kappa_{v a}\left[\frac{\left(v_{a}\right)_{j+1}^{\delta+1}-2\left(v_{a}\right)_{j}^{\delta+1}+\left(v_{a}\right)_{j-1}^{\delta+1}}{2(\Delta r)^{2}}\right. \\
& \left.+\frac{\left(v_{a}\right)_{j+1}^{\delta}-2\left(v_{a}\right)_{j}^{\delta}+\left(v_{a}\right)_{j-1}^{\delta}}{2(\Delta r)^{2}}\right] \\
& +k_{3} R_{N}\left(u_{a}\right)_{j}^{\delta+1}-k_{4}\left(v_{a}\right)_{j}^{\delta+1} \\
& -k_{3}\left(u_{a}\right)_{j}^{\delta}\left(v_{a}\right)_{j}^{\delta} . \tag{4.2}
\end{align*}
$$

Note that the linear terms in (4.1)-(4.2) are treated implicitly while the nonlinear terms are treated explicitly. Otherwise, if all the terms were treated implicitly, we would need to solve a nonlinear system at each time step, which is a more demanding computational task. As mentioned in Section 3, we consider more particularly the regime of small concentrations in which the nonlinear contributions are weaker than the linear ones. Therefore, the explicit treatment of the nonlinear terms is not expected to significantly deteriorate the stability of the Crank-Nicolson scheme, provided the time step $\Delta t$ is selected sufficiently small.

For the reflecting and transmission conditions, we use second-order backward and forward finite-difference formulas. For example, at the interface $r=r_{m}^{i}$ between
the cytoplasm and membrane (say at $j=j_{1}$ ), we have

$$
\begin{aligned}
& \kappa_{m} \frac{-\left(u_{m}\right)_{j_{1}+2}^{\delta}+4\left(u_{m}\right)_{j_{1}+1}^{\delta}-3\left(u_{m}\right)_{j_{1}}^{\delta}}{2 \Delta r} \\
= & \kappa_{c} \frac{3\left(u_{c}\right)_{j_{1}}^{\delta}-4\left(u_{c}\right)_{j_{1}-1}^{\delta}+\left(u_{c}\right)_{j_{1}-2}^{\delta}}{2 \Delta r}
\end{aligned}
$$

and, at the interface $r=r_{n}$ between the nucleus and cytoplasm (say at $j=j_{0}$ ), we have

$$
\begin{aligned}
& \kappa_{c} \frac{-\left(u_{c}\right)_{j_{0}+2}^{\delta}+4\left(u_{c}\right)_{j_{0}+1}^{\delta}-3\left(u_{c}\right)_{j_{0}}^{\delta}}{2 \Delta r} \\
= & \kappa_{n} \frac{3\left(u_{n}\right)_{j_{0}}^{\delta}-4\left(u_{n}\right)_{j_{0}-1}^{\delta}+\left(u_{n}\right)_{j_{0}-2}^{\delta}}{2 \Delta r}
\end{aligned}
$$

Similar formulas are used for $v_{a}$. In this way, the full numerical scheme is second order in both space and time.

Collecting all the discretized equations and expressing the resulting algebraic system in matrix form, we obtain

$$
\begin{equation*}
\mathbf{A} \mathbf{w}^{\delta+1}=\mathbf{B} \mathbf{w}^{\delta}+\mathbf{b}^{\delta} \tag{4.3}
\end{equation*}
$$

where

$$
\mathbf{w}^{\delta}=\left(\mathbf{D}_{\mathbf{n}}, \mathbf{D}_{\mathbf{c}}, \mathbf{D}_{\mathbf{m}}, \mathbf{V}_{\mathbf{n}}, \mathbf{V}_{\mathbf{c}}, \mathbf{V}_{\mathbf{m}}\right)^{\top}
$$

with

$$
\begin{aligned}
& \mathbf{D}_{\mathbf{a}}=\left(\left(u_{a}\right)_{1}^{\delta},\left(u_{a}\right)_{2}^{\delta}, \ldots,\left(u_{a}\right)_{J_{a}}^{\delta}\right)^{\top} \\
& \mathbf{V}_{\mathbf{a}}=\left(\left(v_{a}\right)_{1}^{\delta},\left(v_{a}\right)_{2}^{\delta}, \ldots,\left(v_{a}\right)_{J_{a}}^{\delta}\right)^{\top}
\end{aligned}
$$

The sparse vector $\mathbf{b}^{\delta}$ on the right-hand side of (4.3) contains contributions from the nonlinear terms in the nuclear model, and from the boundary condition (2.5)(2.6) at the external surface of the membrane. The coefficient matrices $\mathbf{A}$ and $\mathbf{B}$ are sparse matrices resulting from the Crank-Nicolson discretization. Given the solution $\mathbf{w}^{\delta}$ at time $t_{\delta}$, the solution $\mathbf{w}^{\delta+1}$ at the next time $t_{\delta+1}$ is found by solving the linear system (4.3) through direct Gaussian elimination.
4.2. Discussion of numerical results. Figure 4 shows snapshots of $D(r, t)$ and $V(r, t)$ over the entire cell, $0<r<r_{m}^{e}$, at various times during the first two $\mathrm{Ca}^{2+}$ spikes (Figure 3). For graphical purposes, the relative sizes of the nucleus, cytoplasm and membrane are not imposed exactly. We particularly zoom in the nuclear region where the $D-V$ interaction takes place. The variables are non-dimensionalized according to (3.4) with $\mathcal{D}_{0}=10^{-2}$, except that $\mathcal{R}=r_{m}^{e}$ so that $r \in[0,1]$ spans the entire cell. We choose $r_{n}=0.4$ and $r_{m}^{i}=0.8$. The computational domain is discretized into $J_{a}=100$ grid points and the time step is set to be $\Delta t=10^{-3}$.

We first observe that both curves are continuous across the whole domain at all times, which indicates that the transmission conditions at $r_{n}$ (nucleus-cytoplasm interface) and $r_{m}^{i}$ (cytoplasm-membrane interface) are well simulated by the numerical scheme. The difference in curve slope is representative of the different diffusivities specified in the three regions of the cell.

We clearly see that, as $D$ diffuses in, it activates the production of $V$ in the nucleus. During the increasing phase of the first $\mathrm{Ca}^{2+}$ spike $(t \lesssim 12)$, the influx of $D$ through the membrane is high and, accordingly, the ingoing diffusion is strong. Due to the reflecting boundary at $r=0$, there is a gradual build-up of $D$ in the nucleus, which persists even during the $\mathrm{Ca}^{2+}$ decreasing phase ( $12 \lesssim t \lesssim 17$ ). On
the other hand, $V$ steadily builds up in the nucleus during the whole $\mathrm{Ca}^{2+}$ spike, while diffusing out to the membrane. The successive increasing and decreasing phases of $\mathrm{Ca}^{2+}$ spiking are clearly indicated by the changes in $D$ level at $r=r_{m}^{e}$. The zero Dirichlet boundary condition for $V$ at $r=r_{m}^{e}$, corresponding to instant diffusion out of the cell, is also well reproduced numerically.

During the second $\mathrm{Ca}^{2+}$ spike $(17 \lesssim t \lesssim 29)$, this reaction-diffusion process repeats itself. The new influx further raises the $D$ concentration in the nucleus before it has a chance to diffuse all out. As a result, more $V$ is created there. This suggests that tying the boundary conditions to the calcium spikes may not be correct and that a more complicated mechanical procedure is necessary to monitor the intake of $\mathrm{D}_{3}$, such as receptor-mediated endocis or perhaps through the closing and snipping of caveolae suggested earlier. These models will be investigated in a future work. Note the resemblance of concentration profiles for $D$ between $t=20$ and $t=30$ which correspond to a periodic $\mathrm{Ca}^{2+}$ cycle (Figure 3). In contrast, the $V$ concentration in the nucleus continually increases during those two $\mathrm{Ca}^{2+}$ spikes. It remains five orders of magnitude lower than the $D$ concentration up to $t=30$. According to our non-dimensionalization (3.4), the typical diffusion time through the cell is $O(1)$. Therefore, the simulation time up to $t=30$ is sufficiently long to illustrate the reaction-diffusion process between $D$ and $V$ through the entire cell during at least two $\mathrm{Ca}^{2+}$ spikes.

Furthermore, a version of the numerical scheme described in the previous section, was implemented to solve the nuclear model (3.5)-(3.6) with $g(t)=h(t)=1$, in view of testing the asymptotic solution $\left(D^{(0)}, V^{(0)}\right)$ of Theorem 3.2. Figure 5 shows the comparison between the asymptotic and numerical solutions (for both $D$ and $V$ ) at various times. Here we choose $\mathcal{R}=r_{n}$ so that $r \in[0,1]$ only spans the nucleus. Again $J_{a}=100$ and $\Delta t=10^{-3}$. A number of 10 terms is used in the Fourier series (3.14) of $\left(D^{(0)}, V^{(0)}\right)$. Overall, an excellent agreement is found. After $t=1$, the asymptotic and numerical curves are indistinguishable at the graphical scale of Figure 5. This result not only verifies the derivation of our asymptotic solution, but also validates a posteriori the choice of $\epsilon$ as the perturbation parameter in our asymptotic calculations, and hence it validates the choice of the scaling regime defined by (3.4).

We also note from Figure 5 that the amplitude of both $D$ and $V$ keeps increasing with time, which is consistent with the fact that the first Fourier coefficient $W_{0}$ in the asymptotic solution tends to grow linearly with $t$ as $t \rightarrow+\infty$ (see Theorem 3.2). In contrast, the higher Fourier coefficients $W_{j}(j>0)$ decay exponentially in time because the corresponding eigenvalues $\lambda_{j}^{ \pm}$are all negative (see Eq. (6.10) in the Appendix). The characteristic parabolic curve for both $D$ and $V$ is reminiscent of the term $r^{2} /\left(2 r_{n}\right)$ in the asymptotic solution, which accommodates the Neumann boundary conditions (3.13) at $r=0$ and $r=r_{n}$.

The growth of $D$ and $V$ in the nucleus is further revealed in Figure 6 which shows numerical values of $E(t)$ as a function of time. The trapezoidal rule was used to evaluate the integral in $r$. According to Theorem 3.1, $E(t)$ grows linearly with $t$ at rate $1+\alpha$ for $g(t)=h(t)=1$. This behavior is well reproduced by the numerical solution, as indicated by the excellent agreement observed in Figure 6.

Finally, the convergence of the Fourier series (3.14) in $\left(D^{(0)}, V^{(0)}\right)$ is examined in Figure 7 which plots the relative $L_{2}$ error between the asymptotic and numerical solutions in the nucleus at $t=0.01$ and $t=10$, as a function of the number of terms in the series. Overall, the errors for $D$ and $V$ are both found to be very
small. At $t=0.01$ (early time), they rapidly decrease as the number of terms in the series increases, to plateau around $10^{-2}$. At $t=10$ (later time), these errors remain pretty much constant around $10^{-4}$. This result is not surprising since, as mentioned above, the first term $W_{0}$ tends to prevail as $t$ increases. The higher terms $(j>0)$ decay exponentially with $t$ and thus, very quickly, do not contribute further to the convergence of the Fourier series.
5. Conclusions. In this paper, we have provided a simplified mathematical model of the pre-osteoblastic cell. We have indicated where this model deviates from the biological system. To our knowledge, this is the first time the entire process of entry of $D_{3}$ through the membrane and across the cytoplasm into the nucleus, and the transcription of the gene encoding receptor activator of RANKL, has been mathematically modeled. The graphs indicate the arrival times of $\mathrm{D}_{3}$ and the exit times of RANKL through the membrane. Arrival time in the nucleus is usually too quick compared with the only known experimental data presented in [3]. This can be fixed using an idea discussed in [18] where an effective diffusivity is introduced. This uses the method of homogenization with which we are well aware of ( $[6,7]$ and the references therein). Another important finding in [19] was the approximate formula for the effective diffusion coefficient

$$
\begin{equation*}
D^{H}=D_{0}\left(1-\frac{\phi}{\phi_{c}}\right)^{\alpha \phi_{c}} \tag{5.1}
\end{equation*}
$$

where $D_{0}$ is the diffusion coefficient of the solvent, $\phi$ is the volume fraction of obstacles, $\phi_{c}$ is the critical volume fraction that is the minimal volume fraction of obstacles at which a tracer particle is trapped, and $\alpha$ is the empirical constant. This is one possible way in which to correct the arrival time. Another is to distribute the transport from a purely diffusional process to one in which $\mathrm{D}_{3}$ diffuses until meeting a microtubule and then is transported to the nuclear pore. This would require a more complicated geometry than used here. Numerical experiments conducted in [19] show that, for all relevant volume fractions, (5.1) agrees very well with the effective diffusion obtained by the classical homogenization procedure (see e.g. [8]). Our tying of the boundary conditions for $\mathrm{D}_{3}$ at the membrane surface to the production of calcium waves was rather simplistic. We could rather choose to have the entry of a calcium spike instigating a probabilistic procedure for the passing of a quantity of $D_{3}$ across the membrane. This is envisioned for future work.

Acknowledgments. This work was funded in part by the National Science Foundation through Mathematical Biology grant No. DMS-0920850 and by a grant from the Simons Foundation (No. 246170 to P. Guyenne). P. Guyenne would also like to thank the Institute for Advanced Study (Princeton, NJ) for its hospitality during the academic year 2011-2012.

## 6. Appendix.

6.1. Proof of Theorem 3.2. In this section, we give details on the proof of Theorem 3.2.

We examine the problem (3.10)-(3.13) with $g(t)=h(t)=1$. The nonhomogeneous boundary conditions (3.13) suggest the change of variables

$$
\begin{aligned}
D^{(0)}(r, t) & =u(r, t)+\frac{r^{2}}{2 r_{n}} \\
V^{(0)}(r, t) & =v(r, t)+\frac{r^{2}}{2 r_{n}}
\end{aligned}
$$

so Eqs. (3.10)-(3.13) become

$$
\begin{align*}
\frac{\partial u}{\partial t} & =\frac{\partial^{2} u}{\partial r^{2}}-a u+b v-(a-b) \frac{r^{2}}{2 r_{n}}+\frac{1}{r_{n}}  \tag{6.1}\\
\frac{\partial v}{\partial t} & =\alpha \frac{\partial^{2} v}{\partial r^{2}}+a u-b v+(a-b) \frac{r^{2}}{2 r_{n}}+\frac{\alpha}{r_{n}} \tag{6.2}
\end{align*}
$$

with initial conditions

$$
\begin{equation*}
u(r, 0)=v(r, 0)=-\frac{r^{2}}{2 r_{n}} \tag{6.3}
\end{equation*}
$$

and homogeneous boundary conditions

$$
\begin{equation*}
\frac{\partial u}{\partial r}(0, t)=\frac{\partial v}{\partial r}(0, t)=0, \quad \frac{\partial u}{\partial r}\left(r_{n}, t\right)=\frac{\partial v}{\partial r}\left(r_{n}, t\right)=0 \tag{6.4}
\end{equation*}
$$

By the superposition principle and separation of variables, we look for a solution of (6.1)-(6.4) in terms of Fourier cosine series

$$
\begin{aligned}
& u(r, t)=\sum_{j=0}^{\infty} u_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right) \\
& v(r, t)=\sum_{j=0}^{\infty} v_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right)
\end{aligned}
$$

which satisfy (6.4). Plugging these expressions in (6.1)-(6.2), we obtain

$$
\begin{align*}
& u_{0}^{\prime}=-a u_{0}+b v_{0}+\frac{1}{r_{n}}-(a-b) \frac{r_{n}}{6}  \tag{6.5}\\
& v_{0}^{\prime}=a u_{0}-b v_{0}+\frac{\alpha}{r_{n}}+(a-b) \frac{r_{n}}{6} \tag{6.6}
\end{align*}
$$

for $j=0$, and

$$
\begin{align*}
u_{j}^{\prime} & =-\left(a+\frac{j^{2} \pi^{2}}{r_{n}^{2}}\right) u_{j}+b v_{j}-\frac{2(-1)^{j}(a-b) r_{n}}{j^{2} \pi^{2}}  \tag{6.7}\\
v_{j}^{\prime} & =a u_{j}-\left(b+\alpha \frac{j^{2} \pi^{2}}{r_{n}^{2}}\right) v_{j}+\frac{2(-1)^{j}(a-b) r_{n}}{j^{2} \pi^{2}} \tag{6.8}
\end{align*}
$$

for $j \geq 1$, where we have used

$$
\begin{equation*}
\frac{r^{2}}{2 r_{n}}=\frac{r_{n}}{6}+\sum_{j=1}^{\infty} \frac{2(-1)^{j} r_{n}}{j^{2} \pi^{2}} \cos \left(\frac{j \pi r}{r_{n}}\right) \tag{6.9}
\end{equation*}
$$

and the primes stand for differentiation. Let

$$
W_{j}(t)=\binom{u_{j}(t)}{v_{j}(t)}, \quad A_{j}=\left(\begin{array}{cc}
-a-\sigma_{j} & b \\
a & -b-\alpha \sigma_{j}
\end{array}\right), \quad \sigma_{j}=\frac{j^{2} \pi^{2}}{r_{n}^{2}}
$$

for $j \geq 0$, and

$$
B_{0}=\binom{\frac{1}{r_{n}}-\frac{(a-b) r_{n}}{6}}{\frac{\alpha}{r_{n}}+\frac{(a-b) r_{n}}{6}}, \quad B_{j}=\frac{2(-1)^{j}(a-b) r_{n}}{j^{2} \pi^{2}}\binom{-1}{1},
$$

for $j \geq 1$. The eigenvalues of $A_{j}$ are

$$
\lambda_{j}^{ \pm}=\frac{1}{2}\left(-p_{j} \pm \sqrt{p_{j}^{2}-4 q_{j}}\right),
$$

with

$$
p_{j}=(1+\alpha) \sigma_{j}+a+b, \quad q_{j}=\left(\alpha \sigma_{j}+b+\alpha a\right) \sigma_{j}
$$

and, in particular, they reduce to $\lambda_{0}^{+}=0$ and $\lambda_{0}^{-}=-(a+b)$ for $j=0$. For $j>0$, both $\lambda_{j}^{ \pm}<0$ because the parameters $a, b, \alpha$ and $\sigma_{j}$ are all positive. These evanescent modes are representative of the underlying diffusion process.

Equations (6.5)-(6.6) can be rewritten as

$$
W_{0}^{\prime}=A_{0} W_{0}+B_{0}
$$

whose solution is given by

$$
W_{0}(t)=e^{t A_{0}}\left(\int_{0}^{t} e^{-\tau A_{0}} B_{0} d \tau+C_{0}\right)
$$

using the integrating factor technique. By diagonalization,

$$
e^{ \pm t A_{0}}=\frac{1}{a+b}\left(\begin{array}{ll}
b+a e^{\mp(a+b) t} & b-b e^{\mp(a+b) t} \\
a-a e^{\mp(a+b) t} & a+b e^{\mp(a+b) t}
\end{array}\right)
$$

and thus

$$
W_{0}(t)=\frac{1}{6 r_{n}(a+b)^{2}}\binom{-\beta_{0} e^{-(a+b) t}+b \gamma_{0} t+\beta_{0}}{\beta_{0} e^{-(a+b) t}+a \gamma_{0} t-\beta_{0}}+e^{t A_{0}} C_{0}
$$

where

$$
\beta_{0}=\left(b^{2}-a^{2}\right) r_{n}^{2}+6 a-6 a b, \quad \gamma_{0}=6(a+b)(1+\alpha)
$$

The constant of integration

$$
C_{0}=W_{0}(0)=-\frac{r_{n}}{6}\binom{1}{1}
$$

is determined from the initial condition (6.3) together with the first term of the Fourier series (6.9). Collecting all the contributions, we find

$$
W_{0}(t)=\binom{-\left[\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}+\frac{(a-b) r_{n}}{6(a+b)}\right] e^{-(a+b) t}+\frac{b \gamma_{0} t}{6 r_{n}(a+b)^{2}}+\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}-\frac{b r_{n}}{3(a+b)}}{\left[\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}+\frac{(a-b) r_{n}}{6(a+b)}\right] e^{-(a+b) t}+\frac{a \gamma_{0} t}{6 r_{n}(a+b)^{2}}-\frac{\beta_{0}}{6 r_{n}(a+b)^{2}}-\frac{a r_{n}}{3(a+b)}} .
$$

Similarly, Eqs. (6.7)-(6.8) for $j \geq 1$ can be expressed as

$$
W_{j}^{\prime}=A_{j} W_{j}+B_{j}
$$

whose solution is given by

$$
W_{j}(t)=e^{t A_{j}}\left(\int_{0}^{t} e^{-\tau A_{j}} B_{j} d \tau+C_{j}\right)
$$

where

$$
C_{j}=W_{j}(0)=\frac{2(-1)^{j+1} r_{n}}{j^{2} \pi^{2}}\binom{1}{1}
$$

as derived from the Fourier coefficients of (6.9), and

$$
e^{ \pm t A_{j}}=\frac{1}{\lambda_{j}^{+}-\lambda_{j}^{-}}\left(\begin{array}{cc}
-m_{j}^{-} e^{ \pm \lambda_{j}^{+} t}+m_{j}^{+} e^{ \pm \lambda_{j}^{-} t} & b\left(e^{ \pm \lambda_{j}^{+} t}-e^{ \pm \lambda_{j}^{-} t}\right)  \tag{6.10}\\
-\frac{m_{j}^{+} m_{j}^{-}}{b}\left(e^{ \pm \lambda_{j}^{+} t}-e^{ \pm \lambda_{j}^{-} t}\right) & m_{j}^{+} e^{ \pm \lambda_{j}^{+} t}-m_{j}^{-} e^{ \pm \lambda_{j}^{-} t}
\end{array}\right)
$$

with

$$
m_{j}^{ \pm}=\lambda_{j}^{ \pm}+\sigma_{j}+a .
$$

This leads to

$$
\begin{aligned}
I_{j} & =\int_{0}^{t} e^{-\tau A_{j}} B_{j} d \tau \\
& =\frac{2(-1)^{j}(a-b) r_{n}}{\left(\lambda_{j}^{+}-\lambda_{j}^{-}\right) j^{2} \pi^{2}} \\
& \times\binom{\quad-\frac{m_{j}^{-}+b}{\lambda_{j}^{+}} e^{-\lambda_{j}^{+} t}+\frac{m_{j}^{+}+b}{\lambda_{j}^{-}} e^{-\lambda_{j}^{-} t}+\frac{m_{j}^{-}+b}{\lambda_{j}^{+}}-\frac{m_{j}^{+}+b}{\lambda_{j}^{-}}}{-\frac{m_{j}^{+} m_{j}^{-}+b m_{j}^{+}}{b \lambda_{j}^{+}} e^{-\lambda_{j}^{+} t}+\frac{m_{j}^{+} m_{j}^{-}+b m_{j}^{-}}{b \lambda_{j}^{-}} e^{-\lambda_{j}^{-} t}+\frac{m_{j}^{+} m_{j}^{-}+b m_{j}^{+}}{b \lambda_{j}^{+}}-\frac{m_{j}^{+} m_{j}^{-}+b m_{j}^{-}}{b \lambda_{j}^{-}}}
\end{aligned}
$$

so that

$$
W_{j}(t)=e^{t A_{j}} I_{j}+e^{t A_{j}} W_{j}(0)
$$

Therefore

$$
\binom{D^{(0)}(r, t)}{V^{(0)}(r, t)}=\sum_{j=0}^{\infty} W_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right)+\frac{r^{2}}{2 r_{n}}\binom{1}{1}
$$

which completes the proof of the theorem.
We remark that the next-order terms in the power-series solution (3.8)-(3.9) can be determined in a similar way, however this procedure becomes increasingly more tedious.
6.2. Proof of Corollary 1. Here we present the proof of Corollary 1 which deals with the more general case of time-dependent fluxes through the surface of the nucleus.

The main steps are similar to those for the proof of Theorem 3.2. The nonhomogeneity of (3.13) is taken care of by decomposing

$$
\begin{aligned}
D^{(0)}(r, t) & =u(r, t)+\frac{r^{2}}{2 r_{n}} g(t) \\
V^{(0)}(r, t) & =v(r, t)+\frac{r^{2}}{2 r_{n}} h(t)
\end{aligned}
$$

The auxiliary functions $u$ and $v$ satisfy

$$
\begin{gather*}
\frac{\partial u}{\partial t}=\frac{\partial^{2} u}{\partial r^{2}}-a u+b v+s(r, t) \\
\frac{\partial v}{\partial t}=\alpha \frac{\partial^{2} v}{\partial r^{2}}+a u-b v+z(r, t) \\
u(r, 0)=-\frac{r^{2}}{2 r_{n}} g(0), \quad v(r, 0)=-\frac{r^{2}}{2 r_{n}} h(0) \\
\frac{\partial u}{\partial r}(0, t)=\frac{\partial v}{\partial r}(0, t)=0, \quad \frac{\partial u}{\partial r}\left(r_{n}, t\right)=\frac{\partial v}{\partial r}\left(r_{n}, t\right)=0 \tag{6.11}
\end{gather*}
$$

where

$$
\begin{aligned}
& s(r, t)=\left(\frac{1}{r_{n}}-\frac{a r^{2}}{2 r_{n}}\right) g(t)-\frac{r^{2}}{2 r_{n}} g^{\prime}(t)+\frac{b r^{2}}{2 r_{n}} h(t), \\
& z(r, t)=\left(\frac{\alpha}{r_{n}}-\frac{b r^{2}}{2 r_{n}}\right) h(t)-\frac{r^{2}}{2 r_{n}} h^{\prime}(t)+\frac{a r^{2}}{2 r_{n}} g(t) .
\end{aligned}
$$

Given the homogeneous boundary conditions (6.11), we assume the possibility of writing $u$ and $v$ as Fourier cosine series

$$
\begin{aligned}
& u(r, t)=\sum_{j=0}^{\infty} u_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right), \\
& v(r, t)=\sum_{j=0}^{\infty} v_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right),
\end{aligned}
$$

and, similarly,

$$
\begin{aligned}
s(r, t) & =\sum_{j=0}^{\infty} s_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right) \\
z(r, t) & =\sum_{j=0}^{\infty} z_{j}(t) \cos \left(\frac{j \pi r}{r_{n}}\right)
\end{aligned}
$$

where

$$
\begin{aligned}
s_{j}(t) & =\frac{2}{r_{n}} \int_{0}^{r_{n}} s(r, t) \cos \left(\frac{j \pi r}{r_{n}}\right) d r \\
z_{j}(t) & =\frac{2}{r_{n}} \int_{0}^{r_{n}} z(r, t) \cos \left(\frac{j \pi r}{r_{n}}\right) d r, \quad j \geq 0
\end{aligned}
$$

while the $u_{j}$ and $v_{j}$ 's obey

$$
\begin{align*}
u_{j}^{\prime} & =-\left(a+\sigma_{j}\right) u_{j}+b v_{j}+s_{j}  \tag{6.12}\\
v_{j}^{\prime} & =a u_{j}-\left(b+\alpha \sigma_{j}\right) v_{j}+z_{j} \tag{6.13}
\end{align*}
$$

given

$$
\sigma_{j}=\frac{j^{2} \pi^{2}}{r_{n}^{2}}
$$

Let

$$
W_{j}(t)=\binom{u_{j}(t)}{v_{j}(t)}, \quad A_{j}=\left(\begin{array}{cc}
-a-\sigma_{j} & b \\
a & -b-\alpha \sigma_{j}
\end{array}\right), \quad Y_{j}(t)=\binom{s_{j}(t)}{z_{j}(t)}
$$

for $j \geq 0$. Equations (6.12)-(6.13) then read

$$
W_{j}^{\prime}=A_{j} W_{j}+Y_{j}
$$

which can be solved by the integrating factor technique, yielding

$$
W_{j}(t)=e^{t A_{j}} \int_{0}^{t} e^{-\tau A_{j}} Y_{j}(\tau) d \tau+e^{t A_{j}} W_{j}(0)
$$



Figure 1. Dynamics of vitamin $\mathrm{D}_{3}$ on a preosteoblast (courtesy of Dr. Anja Nohe).
where the exponentials $e^{ \pm t A_{j}}$ are defined by (6.10) and $W_{j}(0)=\left(u_{j}(0), v_{j}(0)\right)^{\top}$ is determined from the initial conditions

$$
\begin{aligned}
u(r, 0) & =-\frac{r^{2}}{2 r_{n}} g(0)=\sum_{j=0}^{\infty} u_{j}(0) \cos \left(\frac{j \pi r}{r_{n}}\right), \\
v(r, 0) & =-\frac{r^{2}}{2 r_{n}} h(0)=\sum_{j=0}^{\infty} v_{j}(0) \cos \left(\frac{j \pi r}{r_{n}}\right) .
\end{aligned}
$$

Hence

$$
\begin{aligned}
u_{j}(0) & =\frac{2}{r_{n}} \int_{0}^{r_{n}}\left(-\frac{r^{2}}{2 r_{n}} g(0)\right) \cos \left(\frac{j \pi r}{r_{n}}\right) d r \\
v_{j}(0) & =\frac{2}{r_{n}} \int_{0}^{r_{n}}\left(-\frac{r^{2}}{2 r_{n}} h(0)\right) \cos \left(\frac{j \pi r}{r_{n}}\right) d r .
\end{aligned}
$$

This concludes the proof of the corollary.

## REFERENCES

[1] A. Atri, J. Amundson, D. Clapham and J. Sneyd, A single-pool model for intracellular calcium oscillations and waves in the Xenopus laevis Oocyte, Biophys. J., 65 (1993), 1727-1739.
[2] F. Bronner, Cytoplasmic transport of calcium and other inorganic ions, Comp. Biochem. Physiol., 115B (1996), 313-317.
[3] J. Buchanan, R. P. Gilbert and M. J. Ou, The kinetics of vitamin $D_{3}$ in the osteoblastic cell, Submitted to J. Theor. Biol., (2012).
[4] E. M. Costa and D. Feldman, Measurement of 1,25-Dihydroxyvitamin D3 receptor turnover by dense amino acid labeling: Changes during receptor up-regulation by vitamin $D$ metabolites, Endocrinology, 120 (1987), 1173-1178.
[5] M. C. Farach-Carson and P. J. Davis, Steroid hormone interactions with target cells: Cross talk between membrane and nuclear pathways, J. Pharm. Exp. Therap., 307 (2003), 839-845.
[6] R. Gilbert, A. Panasenko and A. Vasilic, Acoustic Propagation in a Random Saturated Medium: The Monophasic Case, Math. Mehods Appl. Sciences, 33 (2010), 2206-2214.
[7] K. Hackl and S. Ilic, Application of the multiscale FEM to the modeling of cancellous bone, Biomechan. Model. Mechanobiol., 9 (2010), 87-102.


Figure 2. Sketch of the cellular model with all the components involved.


Figure 3. Effective $\mathrm{D}_{3}$ concentration at the external membrane surface $r=r_{m}^{e}$ as given by the boundary conditions (2.5)-(2.7).
[8] V. V. Jikov, S. M. Kozlov and O.A. Oleinik, "Homogenization of Differential Operators and Integral Functionals," Springer, Berlin 1994.
[9] J. Keener and J. Sneyd, "Mathematical Physiology," Springer, Berlin, 1998.
[10] S. V. Komarova, R. J. Smith, S. J. Dixon, S. M. Sims and L. M. Wahl, Mathematical model predicts a critical role for osteoclast autocrine regulation in the control of bone remodeling, Bone, 33 (2003), 206-215.
[11] D. L. Lacey, E. Timms, h.-L. Tan, M. J. Kelley, C. R. Dunstan, T. Burgess, R. Elliott, A. Columbero, G. Elliott, S. Scully, H. Hsu, J. Sullivan, N. Hawkins, E. Davy, C. Capparelli, A. Eli, Y. X. Qian, S. Kaufman, I. Sarosi, V. Shalhoub, G. Senaldi, J. Guo, J. Delaney and W. J. Boyle, Osteoprotegerin ligand is a cytokine that regulates osteoclast differentiation and activation, Cell, 93 (1998), 165-176.
[12] D. A. Lauffenburger and J. Linderman, "Receptors: Models for Binding, Trafficking and Signaling," Oxford University Press, New York, 1996.


See next page for caption.
[13] V. Lemaire, F. L. Tobin, L. D. Greller, C. R. Cho and L. J. and Suva, Modeling the interactions between osteoblast and osteoclast activities in bone remodeling, J. Theor. Biol., 229 (2004), 293-309.
[14] I. Nemere, 24,25-Dihydroxyvitamin D3 suppresses the rapid actions of 1,25 Dihydroxyvitamin D3 and parathyroid hormone on calcium transport in chick intestine, Bone Miner. Res., 14 (1999), 1543-1549.
[15] I. Nemere, N. Garbi, G. J. Hammerling and R. C. Khanal, Intestinal cell calcium uptake and the targeted knockout of the 1,25D3-MARRS (Membrane-associated, rapid response steroidbinding) receptor/PDIA3/Erp57, J. Biol. Chem., 285 (2010), 31859-31866.


Figure 4. Snapshots of $D$ (left) and $V$ (right) over the entire cell $\left(0 \leq r \leq r_{m}^{e}=1\right)$ at $t=5,10,15,20,25$ and 30 . The vertical dotted lines at $r=r_{n}=0.4$ and $r=r_{m}^{i}=0.8$ separate the three regions of the cell (nucleus, cytoplasm and membrane).
[16] I. Nemere, R. J. Pietras and P. F. Blackmore, Membrane receptors for membrane hormones: Signal transduction and physiological significance, J. Cell Biochem., 88 (2003), 438-445.
[17] A. W. Norman, "Rapid Biological Responses Mediated by $1 \alpha, 25$-Dihydroxyvitamin $\mathrm{D}_{3}$," In Vitamin D (D. Feldman and F. H. Glorieux and J. W. Pike), Academic Press, New York, 1997.


Figure 5. Comparison between the numerical (solid line) and asymptotic (dashed line) solutions for $D$ (left) and $V$ (right) in the nucleus $\left(0 \leq r \leq r_{n}=1\right)$ at $t=1,10$ and 30 .
[18] I. Novak, P. Kraikivski and B. Slepchenko, Diffusion in cytoplasm: Effects of excluded volume due to internal membranes and cytoskeletal structures, Biophys. J., 97 (2009), 758-767.
[19] I. L. Novak, F. Gao, P. Kraikivski and B. Slepchenko, B. M. Diffusion amid random overlapping obstacles: Similarities, invariants, aproximations, J. Chem. Phys., 134 (2011), 154104.
[20] J. W. Pike, "The Vitamin D Receptor and Its Gene," In Vitamin D (D. Feldman, F. H. Glorieux and J. W. Pike), Academic Press, New York, 1997, 105-125.


Figure 6. Numerical values of $E(t)$ as a function of time (circles). For comparison, the theoretical curve $(1+\alpha) t$, with $\alpha=0.25$, is also plotted in solid line.
[21] W. S. Simonet, D. L. Lacey, C. R. Dunstan, M. Kelley, M.-S. Chang, R. Lothy, H. Q. Nguyen, S. Wooden, L. Bennett, T. Boone, G. Shimamoto, M. DeRose, R. Elliott, A. Columbero, H.L. Tan, G. Trail, J. Sullivan, E. Davy, N. Bucay, L. Renshaw-Gregg, T. M. Hughes, D. Hill, W. Pattison, P. Campbell, S. Sander, G. Van, J. Tarpley, P. Derby, R. Lee and W. J. Boyle, Osteoprotegerin: A novel secreted protein involved in the regulation of bone density, Cell, $\mathbf{8 9}$ (1997), 309-319.
[22] P. Slepchenko, I. Semenova, I. Zaliopin. and V. Rodianaov, Switching of membrane organelles between cytoskeletal transport systems is determined by regulation of the microtubule-based transport, J. Cell Biol., 179 (2007), 635-641.
[23] G. K. Witfield, P. W. Jurutka, C. A. Hausler, J.-C. Hsieh, T. K. Barthel, E. T. Jacobs, C. E. Dominguez, M. L. Thatcher and M. R. Hausler, "Nuclear Vitamin D Receptor: Control of Gene Transcription, and Novel Bioactions," In Vitamin D (D. Feldman, F. H. Glorieux and J. W. Pike), Academic Press, New York, 1997, 219-261.

Received May 06, 2012; Accepted November 16, 2012.

```
E-mail address: gilbert@math.udel.edu
E-mail address: guyenne@math.udel.edu
E-mail address: liuy@math.udel.edu
```



Figure 7. Relative $L_{2}$ error between the asymptotic and numerical solutions in the nucleus at $t=0.01$ (top) and $t=10$ (bottom), as a function of the number of terms in the Fourier series. The solid line (resp. dashed line) corresponds to $D$ (resp. $V$ ). Values for 1 up to 160 terms are plotted.


[^0]:    2010 Mathematics Subject Classification. Primary: 58F15, 58F17; Secondary: 53C35.
    Key words and phrases. RANKL, receptors, osteoblasts, reaction-diffusion, numerical simulations.

[^1]:    ${ }^{1}$ The membrane associated rapid response to steroids specific for vitamin D is also known as MARRS [14].

[^2]:    ${ }^{2}$ This has been spoken of as an intracellular bucket brigade where the hormone moves from one binding protein to the next [2].

