Modeling of reaction-diffusion transport into a core-shell geometry

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Abstract

Fickian diffusion into a core-shell geometry is modeled. The interior core mimics pancreatic Langerhan islets and the exterior shell acts as inert protection. The consumption of oxygen diffusing into the cells is approximated using Michaelis-Menten kinetics. The problem is transformed to dimensionless units and solved numerically. Two regimes are identified, one that is diffusion limited and the other consumption limited. A regression is fit that describes the concentration at the center of the cells as a function of the relevant physical parameters. It is determined that, in a cell culture environment, the cells will remain viable as long as the islet has a radius of around 142 μ m or less and the encapsulating shell has a radius of less than approximately 283 μ m. When the islet is on the order of 100 μ m it is possible for the cells to remain viable in environments with as little as $4.6 \times 10^{-2} \text{ mol/m}^{-3} \text{ O}_2$. These results indicate such an encapsulation scheme may be used to prepare artificial pancreas to treat diabetes.

1. Introduction

Type 1 diabetes is an autoimmune disease that destroys a patient's pancreatic, insulin-producing cells. Treatment requires blood sugar monitoring and the injection of insulin to keep the patient's blood sugar at safe levels. In the worst situations, continual monitoring is necessary via an external insulin pump that is worn under the patient's clothing. These extreme situations warrant the development of artificial pancreas technologies to assist stabilizing blood sugar levels.

Transplanting pancreatic islets has been investigated as a possible approach, but rejection is a significant problem [1, 2, 3, 4]. Immunosuppressive drugs reduce the possibility of rejection, although they are cumbersome to manage and potentially dangerous. A better approach is to develop a means to encapsulate pancreatic islets allowing oxygen (16 Da) and glucose (180 Da) to enter and insulin (5.7 kDa) to egress while preventing the body's immune response, *i.e.*, tumor necrosis factor alpha (25.6 kDa) from accessing the cells. Such an encapsulation scheme, shown in Fig. 1(a), is viable due to the relatively large size of TNA- α compared to the species being transported [5]. It has been proposed that alginate would be a good choice for the encapsulation layer [1]. In this manuscript a mathematical model is presented to examine the inward diffusion of molecules, in particular oxygen, through the encapsulation layer and into the cells.

In 1978 McElwain [6], building from the 1976 work of Lin [7], modeled the transportation of molecules into a cluster of cells. Fickian diffusion was assumed and the consumption of diffusing species was approximated using Michaelis-Menten kinetics. As a result of the non-linear consumption term, an analytical solution was not possible. A broad range of numerical methods have been used to address variations of this problem including the shooting method [7], the random choice method [8], singular Cauchy methods [9], and series decompositions [10]. In these, a membrane was sometimes considered present, but its impact on the diffusion rate was not [8]. Complex modeling has been achieved by finite element methods [11, 12, 13, 14, 15, 16, 17] to allow the study of complex geometries, fluid dynamics, and the presence of anoxic cores and cell death within

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Figure 1: Cells encapsulated in an alginate shell, (a) experimentally prepared, reproduced with permission from Ref. [1], and (b) the mathematical model studied here.

islets. These methods are powerful allowing even alternative geometries, such as hollow fibers [15, 16, 17], to be studied, but their development is costly, frequently requiring the use of multi-physics modeling software.

Here we extend the theoretical work of McElwain [6] and Lin [7], by modeling Fickian diffusion into a spherical core of cells that is encapsulated by a protective shell. The consumption of diffusing species is approximated by Michaelis-Menten kinetics. The results are presented in dimensionless form to allow broad applicability. Steady-state concentrations are calculated for a range of geometries. Two regimes of behavior are identified: one that is diffusion limited and another that is consumption limited. The concentration in the interior of the cell is determined as a function of the Michaelis-Menten parameters, diffusion coefficients, geometry, and the exterior concentration of diffusion species.

2. Model

Our model system, shown in Fig. 1(b), is based on experimentally synthesized systems such as those described in Ref. [2] and shown in Fig. 1(a). Spherical symmetry allows the radial coordinate, r, to be the sole geometric descriptor. The islet of cells has radius R_1 and the encapsulating shell has radius R_2 . The problem is broken into an interior region, solved over the domain $(0 < r < R_1)$, and an exterior region, solved over the domain $(0 < r < R_1)$, and an exterior region, solved over the domain $(k_1 < r < R_2)$; the variables associated with the interior and exterior regions are identified with the subscripts i and e respectively.

Assuming Fickian diffusion and Michaelis-Menten consumption, the steady-state concentration in the interior, $c_i(r)$, is determined from the equation

$$D_i \left(\frac{2}{r}\frac{dc_i}{dr} + \frac{d^2c_i}{dr^2}\right) - \frac{V_m c_i}{K_m + c_i} = 0,\tag{1}$$

where D is the diffusion coefficient and V_m and K_m are the Michaelis-Menten parameters corresponding to the maximum consumption rate and the Michaelis constant. The exterior steady-state concentration, $c_e(r)$, is determined from the equation

$$D_e\left(\frac{2}{r}\frac{dc_e}{dr} + \frac{d^2c_e}{dr^2}\right) = 0.$$
(2)

These two second-order ordinary differential equations have four boundary conditions. Due to spherical symmetry, there is no flux across the center of the sphere, therefore

$$\left. \frac{\partial c_i}{\partial r} \right|_{r=0} = 0. \tag{3}$$

The exterior is in contact with an infinite, well-stirred reservoir with constant concentration C_2 , yielding

$$c_e\left(R_2\right) = C_2.\tag{4}$$

The concentration distribution must be continuous at the interface, therefore

$$c_i\left(R_1\right) = c_e\left(R_1\right).\tag{5}$$

In steady state the flux must be equal at the interface. Using Fick's first law,

$$D_i \frac{dc_i}{dr}\Big|_{r=R_1} = D_e \frac{dc_e}{dr}\Big|_{r=R_1}.$$
(6)

This system of equations is transformed to dimensionless form to prepare a general solution. The dimensionless radial position is

$$\rho = \frac{r}{R_2}.\tag{7}$$

The dimensionless interior and exterior concentrations are

$$\chi_i = \frac{c_i}{C_2} \tag{8}$$

and

$$\chi_e = \frac{c_e}{C_2}.\tag{9}$$

The dimensionless Michaelis-Menten parameters are

$$\nu = \frac{V_m R_2^2}{D_i C_2} \tag{10}$$

and

$$\kappa = \frac{K_m}{C_2}.\tag{11}$$

The differential Eqs. 1 and 2 transform to

$$\frac{2}{\rho}\frac{d\chi_i}{d\rho} + \frac{d^2\chi_i}{d\rho^2} - \frac{\nu\chi_i}{\kappa + \chi_i} = 0 \quad \left(0 < \rho < \frac{R_1}{R_2}\right) \tag{12}$$

and

$$\frac{2}{\rho}\frac{d\chi_e}{d\rho} + \frac{d^2\chi_e}{d\rho^2} = 0 \quad \left(\frac{R_1}{R_2} < \rho < 1\right). \tag{13}$$

The boundary conditions 3, 4, 5, and 6 transform to

$$\left. \frac{d\chi_i}{d\rho} \right|_{\rho=0} = 0,\tag{14}$$

$$\chi_e\left(1\right) = 1,\tag{15}$$

$$\chi_i \left(\frac{R_1}{R_2}\right) = \chi_e \left(\frac{R_1}{R_2}\right),\tag{16}$$

and

$$D_i \frac{d\chi_i}{d\rho}\Big|_{\rho = \frac{R_1}{R_2}} = D_e \frac{d\chi_e}{d\rho}\Big|_{\rho = \frac{R_1}{R_2}}.$$
(17)

The first term in Eqs. 12 and 13 can be eliminated by the substitution

1

$$u_i(\rho) = \rho \chi_i(\rho) \tag{18}$$

and

$$\iota_e\left(\rho\right) = \rho\chi_e\left(\rho\right),\tag{19}$$

which yields

$$\frac{d^2 u_i}{d\rho^2} - \frac{\nu u_i}{\kappa + u_i \rho^{-1}} = 0$$
(20)

and

$$\frac{d^2 u_e}{d\rho^2} = 0. \tag{21}$$

The boundary conditions transform to

$$u_i(0) = 0,$$
 (22)

$$u_e\left(1\right) = 1,\tag{23}$$

$$u_i\left(\frac{R_1}{R_2}\right) = u_e\left(\frac{R_1}{R_2}\right),\tag{24}$$

and

$$D_i \frac{du_i}{d\rho} \bigg|_{\rho = \frac{R_1}{R_2}} = D_e \frac{du_e}{d\rho} \bigg|_{\rho = \frac{R_1}{R_2}}.$$
(25)

3. Results

A numerical solution to differential equations 20 and 21, with boundary conditions 22, 23, 24, and 25, is possible by combining the shooting method with the classical fourth-order Runge-Kutta algorithm [18]. The numerical results are verified using the matlab solver *boundary value problem* 4c (*bvp4c*). A detailed discussion of the numerical solution is included as supplementary information. The dimensionless solutions are presented in Fig. 2 for a wide range of diffusion coefficient ratios, D_i/D_e , dimensionless Michaelis-Menten parameters, ν and κ , and radius ratio, R_1/R_2 .

In terms of application, the most important value is the concentration in the center of the islet, χ_i ($\rho = 0$). This concentration is calculated for over 2.7×10^6 different combinations of R_1/R_2 , D_i/D_e , ν , and κ spanning parameter space. A Bayesian-ridge regression is fit to estimate the concentration across parameter space. In addition to the initial logarithmic transformations of ν and κ , a third-degree polynomial is also applied for the best accuracy of the regression. The resulting \mathbb{R}^2 value is 0.975 and the mean squared error is 0.0344. Details of the regression are given in the supplementary information to this manuscript. The resulting interior concentration is plotted in Fig. 3(a) and (b).

4. Discussion

Fig. 2(g) shows the highest consumption rate and 2(c) the lowest. When the consumption rate is low, the concentration profiles are flat inside $r < R_1$. The interior concentration, $\chi_i(0)$, is determined by the rate of diffusion through the shell. The diffusion coefficient ratios, D_i/D_e , separate the solutions some, with the smaller ratios resulting in a higher $\chi_i(0)$. Systems that have thinner encapsulating shells or a higher D_e have a greater $\chi_i(0)$. This represents a regime of behavior where diffusion limits the interior solution. In contrast the curves separate more when the consumption rate is high. In this regime, geometric effects, R_1/R_2 , become more important than differences in diffusion coefficients, D_i/D_e . The systems with smaller islets have higher $\chi_i(0)$ due to there being overall less consumption. This represents a regime of behavior where the consumption limits the interior solution.

It is noteworthy that there is a crossover between these two regimes. For example, in Fig. 2(a), where $\nu = 1.0$, $\kappa = 0.1$, and $D_i/D_e = 1.30$, the interior solution is approximately equal with $\chi_i(0) = 0.80$ for all three geometries, $R_1/R_2 = 0.25, 0.50$, and 0.75. Increasing the consumption rate shifts the solution toward the consumption limited regime, Fig. 2(d), while decreasing the consumption rate shifts the solution toward the diffusion limited regime, Fig. 2(b). A similar crossover is observed in Fig. 3. In Fig. 3(a), where the consumption parameters constant, the interior concentration is greatest for small values of D_i/D_e and large values of R_1/R_2 , *i.e.*, rapid external diffusion and thin external shells. Changing the geometry, $\chi_i(0)$ begins to drop when R_1/R_2 falls below approximately 0.6. In Fig. 3(b), where D_i/D_e and R_1/R_2 are constant, a significant drop in $\chi_i(0)$ is observed when ν becomes larger than approximately 1.75. Similar demarcations are observed throughout the dataset described by the regression presented in the supplementary information.

The ultimate question being addressed is, Over what parameter ranges do the cells remain viable? The answer to this can be approximated assuming the onset of hypoxia to be 2.5 mmHg [19, 20]. The solubility of O₂ in plasma is 1.39×10^{-3} mM/mmHg, but in most tissues the solubility is between 1/3 and 1/5 this value [21]. Therefore, the solubility of O₂ in the islet is approximated $1.39 \times 10^{-3}/4 = 3.475 \times 10^{-4}$ mM/mmHg; as a result, the onset of hypoxia is around 8.69×10^{-4} mol/m³. In cell culture experiments the oxygen concentration is approximately 150 mmHg, or 0.21 mol/m³. Using this as the surface concentration, C_2 in Eq. 8, the critical dimensionless concentration is $\chi^* = 4.1 \times 10^{-3}$.

This value of χ^* looks fairly small, but to understand the physical limits the other dimensionless parameters need to be transformed. It is clear from Figs. 2 and 3 that ν , defined in Eq. 10, is highly important for controlling the viability of cells. Fig. 3 demonstrates that when $R_1/R_2 = 0.5$ and $D_i/D_e = 1.5$ keeping $\chi_i(0) > \chi^*$ requires $\nu < 10.0$. Substituting $V_m = 3.4 \times 10^{-2} \text{ mol/m}^3 \text{s}$ [22], $D_i = 1.3 \times 10^{-9} \text{ m}^2/\text{s}$ [23], and $C_2 = 0.21 \text{ mol/m}^3$, it is found that keeping $\nu < 10.0$ requires $R_2 < 283 \,\mu\text{m}$, which makes $R_1 = 142 \,\mu\text{m}$. This is a reasonable size considering that the average Langerhans in humans are between 100 and 150 μm [24].

It is likely that in vivo the environmental concentration of O₂ will be less than 0.21 mol/m³; what is the lower limit of C₂? Assuming that we take the smallest possible islets, around 100 μ m, and have a reasonably thin encapsulation layer, approximately $R_1/R_2 = 0.75$, then $R_2 = 133 \ \mu$ m. Maintaining the limit $\nu < 10.0$, the resulting C₂ is $4.6 \times 10^{-2} \ \text{mol/m}^3$, which is approximately 25% the oxygen concentration of cell culture experiments. Fig. 4 shows the relationship between C₂ and R₂ that defines the hypoxia threshold. Values to the right of the line results in cell hypoxia.

The calculations here assume uniform porosity in the alginate, cellular tissue, and environment, but this is known to not be the case. It is possible to modify the model using partition coefficients. For the exterior region, the partition coefficient, due to the limited porosity in the cross-linked alginate, is

$$k_e = \frac{c_e^{sat}}{C_2},\tag{26}$$

where c_e^{sat} is the saturated concentration of the alginate. Similarly, the partition coefficient for the interior region, due to the limited porosity of tissue, is

$$k_i = \frac{c_i^{sat}}{C_2},\tag{27}$$

where c_i^{sat} is the saturated concentration of the tissue.

Typically $k_i < k_e < 1$, but it is difficult to accurately measure the partition coefficients and associated saturated concentrations for alginate and tissue. The value of the coefficient will vary, depending on the alginate processing, cell density, and hydration. As accurate data becomes available, the effect of porosity differences can be included in the model as shown here.

The differential equations 20 and 21 are not affected by the partition coefficients, but the boundary conditions are. The boundary condition in the center of the sphere, expressed in Eqs. 3, 14, and 22, is

not effected, but the remaining three boundary conditions are. The boundary condition at $r = R_2$, Eq. 4, incorporates the partition coefficient k_e and transforms to

$$c_e\left(R_2\right) = k_e C_2.\tag{28}$$

In dimensionless form this becomes

$$\chi_e\left(1\right) = k_e,\tag{29}$$

and further transforms to

$$u_e\left(1\right) = k_e.\tag{30}$$

The boundary condition ensuring continuity at the core-shell interface, Eq. 5, incorporates the partition coefficients and transforms to

$$k_e c_i (R_1) = k_i c_e (R_1).$$
(31)

In dimensionless form this becomes

$$k_e \chi_i \left(\frac{R_1}{R_2}\right) = k_i \chi_e \left(\frac{R_1}{R_2}\right),\tag{32}$$

and further transforms to

$$k_e u_i \left(\frac{R_1}{R_2}\right) = k_i u_e \left(\frac{R_1}{R_2}\right). \tag{33}$$

The boundary condition ensuring the continuity of flux across the core-shell interface, Eq. 6, incorporates the partition coefficients and transforms to

$$k_e D_i \frac{dc_i}{dr}\Big|_{r=R_1} = k_i D_e \frac{dc_e}{dr}\Big|_{r=R_1}.$$
(34)

In dimensionless form this becomes

$$k_e D_i \frac{d\chi_i}{d\rho} \bigg|_{\rho = \frac{R_1}{R_2}} = k_i D_e \frac{d\chi_e}{d\rho} \bigg|_{\rho = \frac{R_1}{R_2}},\tag{35}$$

and further transforms to

$$k_e D_i \frac{du_i}{d\rho} \bigg|_{\rho = \frac{R_1}{R_2}} = k_i D_e \frac{du_e}{d\rho} \bigg|_{\rho = \frac{R_1}{R_2}}.$$
(36)

5. Conclusions

A dimensionless model of Fickian diffusion into a spherical core-shell geometry, where the interior mimics cellular tissue and the exterior inert protective encapsulation, is created. The concentration of diffusing species is calculated for a wide range of dimensionless parameters. Using these results, along with physical constants from literature, it is concluded that it is possible to devise an encapsulation-based artificial pancreas technology.

Using physically reasonable values for the Michaelis-Menten parameters, diffusion constants, and environmental oxygen, it is determined that the cells will remain viable only if the islet is less than 142 μ m, which is the size of actual pancreatic islets in humans. Further, if the islet is held at its minimum size, 100 μ m, it is possible to keep the cells alive if the environmental O₂ concentration is approximately 25% that of cell cultures, around $4.6 \times 10^{-2} \text{ mol/m}^3$. Throughout these calculations there is uncertainty; for example the diffusion coefficient of O₂ in alginate depends strongly on the processing. These variations in physical constants will impact the quantitative accuracy of the calculated results, but the overall conclusions will not be impacted.

The geometry used in the model presented here is clearly an oversimplification. Irregularities in islet and alginate shape will lead to regions with shorter and longer diffusing paths. The consumption rate of oxygen is not truly represented by Michaelis-Menten kinetics because it coupled to the concentration of glucose, which diffuses slower than O_2 . There will likely be interface effects, both at the exterior and where the alginate and cells meet. It is possible to develop highly adjustable finite element simulations to include these complexities, and the results presented here provide compelling motivation for the development of such calculations.

The mathematical model presented here is potentially a starting point for addressing other engineering problems. For example, it might find use in studying spherical catalysts in which the exterior has undergone permanent damage. These results might also serve as the starting point for a model of tumor growth inside tissues, for example, in soft tissue sarcomas.

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Figure 2: The radial concentration profile as a function of Michaelis-Menten parameters and radius ratio. In each of the plots, three radius ratios, R_1/R_2 , are examined with the values 0.25, 0.50, and 0.75, with the darkest line representing 0.25 and the lightest line 0.75. Three diffusion coefficient ratios, D_i/D_e , are examined with the values 1.05, 1.30, and 1.55. The plots in the left, center, and right columns have Michaelis constant, κ , values 0.1, 0.5, and 1.0 respectively; the plots in the top, center, and bottom rows have ν values 1.0, 5.0, and 10.0 respectively. The bottom left plot has the greatest consumption rate and the upper right has the lowest.



Figure 3: The concentration in the center of the islet, $\chi_i(0)$, (a) calculated as a function of D_i/D_e and R_1/R_2 while holding $\nu = 0.5$ and $\kappa = 0.1$ constant, and (b) calculated as a function of $\log_{10}(\kappa)$ and $\log_{10}(\nu)$ while holding $R_1/R_2 = 0.5$ and $D_i/D_e = 1.5$ constant.



Figure 4: The relationship between C_2 and R_2 that defines the hypoxia threshold. The region left of the line represents values where the cells are viable, and the right values where hypoxia occurs.