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Reduced Differential Transform Method to Study the Mathematical Model of Tumor Invasion and Metastasis

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Abstract- In this paper we present Non-linear mathematical model of tumor Invasion and Metastasis. Model focuses on three key variables involved in invasion process, namely, tumor cells, host tissue (Extracellular matrix) and matrix degradative enzymes associated with the tumor cells. The model gives system of non-linear partial differential equations. The Reduced Differential Transform Method is applied to obtain the approximate solution of this model. The study shows the ability of the method for solving nonlinear partial differential equations. It is observed that the method does not required linearization and weak nonlinearity assumptions and is very simple in computation.

Keywords- Tumor cells, Extracellular matrix, matrix degradative enzymes, reduced differential Transform Method

1. INTRODUCTION

Cancer invasion is a cell and tissue-driven process that physical, includes cellular. and molecular determinants to adapt and react throughout its progression (Friedl P. et al., 2011). Most of the clinical tumor patients die from tumor invasion and metastasis. Tumor development is a very complex multistep process involving many intracellular and extracellular phenomena which are strongly nonlinear and time varying (d'Onofrio, A., 2006 & 2008; Bellomo N. et al., 2006). An important role in the process of cancer invasion is performed by matrix degradative enzymes (MDEs) such as metalloproteases (MMPs). They are produced by tumor cells and digest the ECM, which enables the migration of cancer cells through the tissue (Liotta, L. A., 1983; Stevenson et al., 1993).

Within the last three decades a number of mathematical models for tumor growth have been developed as part of the quest to understand tumor growth dynamics. Gatenby and Gawlinski present one of the first models of tumor invasion in the papers (Gatenby, R. A. et al., 1995 & 1996) considers the competition between healthy host cells and modified (tumour) cells and proposes and analyses several models formulated in terms of ordinary differential equations. Gatenby and Gawlinski (Gatenby, R. A. et al., 1996) present a reaction-diffusion model for the

investigation of the role of the alteration of the microenvironmental acidity induced by cancer cells

for their invasion into the organism. In this paper, we study the continuum models of avascular tumor growth investigated by (Chaplain et al.1996, 2000 & 2006). Chaplain used numerical solution (finite difference method) to solve the above problem. A variety of methods, exact, approximate, and purely numerical are available for the solution of this type of problems such as Adomian decomposition method, Homotopy perturbation method and Differential Transform method.

The reduced differential transform method (RDTM) was first proposed by the Turkish mathematician Keskin and Oturanc in the year 2009 to solve linear and non-linear ordinary as well as partial differential equations (Keskin, Y. et al., 2011). This method has received much attention since then and applied to solve a wide variety of problems. This reduced differential transform method is introduced mainly to overcome the demerits of complex calculation of the usual differential transform method. Then a number of authors like (Keskin, Y. et al., 2009 & 2010), (Cenesiz, et.al. 2010), (Taha B. A., 2011), (Taha and Wahab, 2012) solved many equations using this method. This paper is presented to solve tumor invasion and metastasis model by RDTM

2. PROBLEM FORMULATION

In our model, we focus on three key variables implicated in the Tumor invasion process, namely, tumor cells, host tissue (extracellular matrix) and matrix- degradative enzymes associated with the

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tumor cell. By definition, haptotaxis is the directed migratory response of cells to gradients of fixed or bound chemicals. Also, haptotaxis is a major component of directed movement in tumour cell invasion. Indeed, there has been much recent effort to characterise such directed movement (Klominek, J. et al., 1993; Lawrence J.A. et al., 1996). We therefore refer to this directed movement of tumor cells in our model as haptotaxis. To incorporate this directed movement of tumor cells in our model, we take the haptotactic flux to be

$$J_{hapto} = \chi n \Delta f \tag{1}$$

Where χ is the (positive constant) haptotactic coefficient.

And to describe the random motility of the tumor cells we assume a flux of the form

$$J_{random} = -D(f, m)\nabla n \tag{2}$$

Where D(f,m) may be a constant or a function of either the MDE or ECM concentration.

In our partial differential equation model, we do not consider any proliferation of tumor cells. We focus entirely on the cell-matrix interactions and how these interactions affect tumor cell migration. The conservation equation for the tumor cell density n (from Eq. (1) and (2)) is therefore given by

$$\frac{\partial n}{\partial t} + \nabla \cdot (J_{random} + J_{hapto}) = 0 \tag{3}$$

Hence, the partial differential equation governing tumor cell motion (in the absence of cell proliferation) is.

$$\frac{\partial n}{\partial t} = \nabla \cdot (D(f, m) \nabla n) - \chi \nabla \cdot (n \nabla f) \tag{4}$$

In our model we choose $D(f,m) = D_n$, a constant, the tumor cell random motility coefficient.

The ECM contains many macro-molecules, including fibronectin, laminin and collagen, which can be degraded by MDEs (Chambers, A. F. et al., 1997). We assume that the MDEs degrade ECM upon contact and hence the degradation process is modeled by

$$\frac{\partial f}{\partial t} = -\delta mf \tag{5}$$

Where δ is a positive constant.

Active MDEs are produced (or activated) by the tumor cells, diffuse throughout the tissue and undergo some form of decay (either passive or active). Therefore, the governing equation of MDE concentration is given by:

$$\frac{\partial m}{\partial t} = D_m \nabla^2 m + g(n, m) - h(n, m, f) \tag{6}$$

Where D_m is a positive constant, the MDE diffusion coefficient, $g = \mu n$ (MDE production by the tumor cells) and $h = \lambda m$ (natural decay), as for simplicity we assume that there is a linear relationship between the density of tumor cells and the level of active MDEs in the surrounding tissues. Hence the complete system describing the interactions of the tumor cells (denoted by n), extra cellular matrix (ECM, denoted by f), and matrix degrading enzymes (MDE, denoted by m) is given by (Gatenby, R. A. et al, 1996)

$$\frac{\partial n}{\partial t} = D_n \nabla^2 n - \chi \nabla \cdot (n \nabla f) \tag{7}$$

$$\frac{\partial f}{\partial t} = -\delta mf \tag{8}$$

$$\frac{\partial m}{\partial t} = D_m \nabla^2 m + \mu n - \lambda m \tag{9}$$

Where D_n is the tumor cell random motility coefficient, D_m is the MDE diffusion coefficient, χ is the haptotactic coefficient, and λ, μ, δ are the positive constants. Non-dimensionalise of Eq. (7), (8), (9) by setting

$$\overline{n} = \frac{n}{n_0}, \overline{f} = \frac{f}{f_0}, \overline{m} = \frac{m}{m_0}, \overline{x} = \frac{x}{L}, \overline{t} = \frac{t}{\tau}$$
 (10)

Where n_0 is the tumor cell density, f_0 is the ECM density, m_0 is the MDE concentration, L is the length

scale, and
$$au$$
 is the time ($au = \frac{L^2}{D}$, where D is a

reference chemical diffusion coefficient). By dropping the tildes for notational convenience, we obtain the scaled system of equations:

$$\frac{\partial n}{\partial t} = d_n \nabla^2 n - \gamma \nabla \cdot (n \nabla f)$$
 (11)

$$(5) \qquad \frac{\partial f}{\partial t} = -nmf \tag{12}$$

$$\frac{\partial m}{\partial t} = d_m \nabla^2 m + \omega n - \beta m \tag{13}$$

Where
$$d_n = D_n/D$$
, $\gamma = \chi f_0/D$, $\eta = \tau m_0 \delta$, $d_m = D_m/D$, $\omega = \tau \mu n_0/m_0$ and $\beta = \tau \lambda$

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We assume that initially there is a nodule of cells already present and the tumor is centered on x = 0 with n having density distribution.

$$n(x,0) = \exp\left(-\frac{x^2}{\epsilon}\right) \tag{14}$$

Also, we suppose that the tumor has already degraded some of its surrounding tissue and so initially profile of ECM to be.

$$f(x,0) = 1 - 0.5 \exp\left(-\frac{x^2}{\epsilon}\right)$$
 (15)

Finally, we assume that the initial MDE concentration profile is proportional to the n(x, 0) and then we assume,

$$m(x,0) = 0.5 \exp\left(-\frac{x^2}{\epsilon}\right) \tag{16}$$

Where \in is a positive constant. We assume that there is no-flux of tumor cells or MDE across the boundary of the domain, namely x=0 and x=1. These boundary conditions are represented by the following equations.

$$-D_n \frac{\partial n}{\partial x} + n\chi \frac{\partial f}{\partial x} = 0$$
 And $\frac{\partial m}{\partial x} = 0$, at $x = 0,1$

3. REDUCED DIFFERENTIAL TRANSFORM METHOD (RDTM)

The basic definition of the reduced differential Transform method is as follows:

Definition-1: If the function u(x,t) is analytic and differentiable continuously with respect to time t and space x in the domain of interest, then

$$U(k) = \frac{1}{k!} \left| \frac{\partial^k}{\partial x^k} u(x, t) \right| \text{ at } t = 0$$
 (17)

Where the t-dimensional spectrum function U(k) is the transformed function of u(x,t). Here the lower case function u(x,t) represents the original function while the upper case function U(k) stands for the transformed function.

Definition 2: The inverse reduced differential transform of U(k) is defined as follows:

$$u(x,t) = \sum_{k=0}^{\infty} U(k)t^{k}$$
(18)

Thus combining Eq. (17) and (18), we can express the solution as follows:

$$u(x,t) = \sum_{k=0}^{\infty} \left(\frac{1}{k!} \left[\frac{\partial^k}{\partial x^k} u(x,t) \right] \text{ at } t = 0 \right) t^k$$
(19)

The basic concept of reduced differential transform method mainly comes from the power series expansion. Few fundamental mathematical operations performed by this reduced differential method are listed below:

Table 1. Some Transformed functions

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Original Function	Transformed function
u(x,t)	$U(k) = \frac{1}{k!} \left[\frac{\partial^k}{\partial x^k} u(x, t) \right] \text{at}$
	t = 0
$w(x,t) = u(x,t) \pm v(x,t)$	$W(k) = U(k) \pm V(k)$
w(x,t) = cu(x,t)	W(k) = cU(k), where c is
	constant
$w(x,t) = x^m t^n$	$W(k) = \chi^m \delta(k - n)$
	where $\delta(\mathbf{k} - \mathbf{n}) = \begin{cases} 1, k = n \\ 0, k \neq n \end{cases}$
$w(x,t) = \chi^m t^n u(x,t)$	$W(k) = \chi^m U(k - n)$
w(x,t) = u(x,t)v(x,t)	$W(k) = \sum_{r=0}^{k} U(r)V(k-r) = \sum_{r=0}^{k} V(r)U(k-r)$
$w(x,t) = \frac{\partial^r}{\partial t^r} u(x,t)$	$W(k) = \frac{(k+r)!}{k!}U(k+r)$
$w(x,t) = \frac{\partial}{\partial x}u(x,t)$	$W(k) = \frac{\partial}{\partial x} U(k)$
$w(x,t) = \frac{\partial^2}{\partial x^2} u(x,t)$	$W(k) = \frac{\partial^2}{\partial x^2} U(k)$

4. SOLUTION BY RDTM:

Applying RDTM to Eq. (11), (12) & (13) we have following equations:

$$(k+1)N(k+1) = d_n \frac{\partial^2}{\partial x^2} N(k) -$$

$$\gamma \left[\sum_{r=0}^{k} \frac{\partial}{\partial x} N(k-r) \frac{\partial}{\partial x} F(r) + \sum_{r=0}^{k} \frac{N(k-r)}{\partial x^{2}} F(r) \right]$$

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(20)

$$(k+1)F(k+1) = -\eta \sum_{r=0}^{k} M(k-r)F(r)$$
(21)

$$(k+1)M(k+1) = d_m \frac{\partial^2}{\partial x^2} M(k) + \omega N(k)$$

$$-\beta M(k)$$
(22)

Also applying RDTM to the Initial condition Eq. (14), (15) & (16), we have,

$$N(0) = \exp\left(-\frac{x^2}{\epsilon}\right), \quad F(0) = 1 - 0.5 \exp\left(-\frac{x^2}{\epsilon}\right)$$

$$M(0) = 0.5 \exp\left(-\frac{x^2}{\epsilon}\right)$$
(23)

Now taking k = 0 in Eq. (20), (21), (22) and using Eq. (23), we have,

$$N(1) = -\frac{e^{\frac{-x^2}{\epsilon}}}{\epsilon} \left[2d_n \left(1 - \frac{2x^2}{\epsilon} \right) + \gamma e^{\frac{-x^2}{\epsilon}} \left(1 - \frac{4x^2}{\epsilon} \right) \right]$$

$$F(1) = -\frac{\eta e^{\frac{-x^2}{\epsilon}}}{2} \left(1 - \frac{e^{\frac{-x^2}{\epsilon}}}{2} \right)$$

$$M(1) = -e^{\frac{-x^2}{\epsilon}} \left[\frac{d_m}{\epsilon} \left(1 - \frac{2x^2}{\epsilon} \right) - \omega + \frac{\beta}{2} \right]$$

$$N(2) = -\frac{\frac{2d_n}{\epsilon^2}}{2} \begin{cases} d_n \left(3 - \frac{12x^2}{\epsilon} - \frac{4x^4}{\epsilon^2} \right) + \\ \gamma e^{\frac{-x^2}{\epsilon}} \left(3 - \frac{24x^2}{\epsilon} + \frac{16x^4}{\epsilon^2} \right) \end{cases}$$

$$N(2) = -\frac{e^{\frac{-x^2}{\epsilon}}}{2} \begin{cases} \frac{-16d_n}{\epsilon} \left(x^2 - \frac{x^4}{\epsilon} \right) + \\ \frac{8\gamma e^{\frac{-x^2}{\epsilon}}}{\epsilon} \left(\frac{3x^4}{\epsilon} - 2x^2 \right) \\ -2\eta x^2 \left(e^{\frac{-x^2}{\epsilon}} - 1 \right) \\ + \left(1 - \frac{2x^2}{\epsilon} \right) \left(2d_n + \gamma e^{\frac{-x^2}{\epsilon}} \right) \end{cases}$$

$$-\frac{\gamma \eta e^{\frac{-x^2}{\epsilon}}}{\epsilon} \left\{ \frac{d_m}{\epsilon} \left(-1 + \frac{2x^2}{\epsilon} \right) + e^{\frac{-x^2}{\epsilon}} \left(1 - \frac{4x^2}{\epsilon} \right) \right\}$$

Again for k = 1 in recurrence relation, we have,

$$F(2) = \frac{\eta}{2} e^{\frac{-x^2}{\epsilon}} \left(1 - \frac{e^{\frac{-x^2}{\epsilon}}}{2} \right) \left[\frac{d_m}{\epsilon} \left(1 - \frac{2x^2}{\epsilon} \right) - \omega + \frac{\beta}{2} + \frac{\eta e^{\frac{-x^2}{\epsilon}}}{4} \right]$$

$$\int_{d_m} \left\{ \frac{2d_m}{\epsilon^2} \left(-3 + \frac{12x^2}{\epsilon} - \frac{4x^4}{\epsilon^2} \right) + \frac{1}{\epsilon} \left(1 - \frac{2x^2}{\epsilon} \right) (2\omega - \beta) \right\}$$

$$M(2) = -\frac{e^{\frac{-x^2}{\epsilon}}}{2} + \omega \left\{ \frac{2d_n}{\epsilon} \left(1 - \frac{2x^2}{\epsilon} \right) + \frac{\gamma e^{\frac{-x^2}{\epsilon}}}{\epsilon} \left(1 - \frac{4x^2}{\epsilon} \right) \right\}$$

$$-\beta \left\{ \frac{d_m}{\epsilon} \left(1 - \frac{2x^2}{\epsilon} \right) - \omega + \frac{\beta}{2} \right\}$$

And so on.

Finally, the inverse reduced differential transform of N(k), F(k), M(k) are obtained from following relations

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$$n(x,t) = \sum_{k=0}^{\infty} N(k)t^{k} , \quad f(x,t) = \sum_{k=0}^{\infty} F(k)t^{k} , \quad m(x,t) = \sum_{k=0}^{\infty} M(k)t^{k}$$

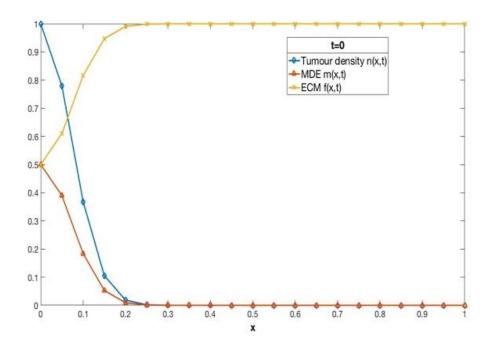


Figure 1 One-dimensional RDTM solution of system of Eq. (11)-(13) with constant tumor cell diffusion showing the cell diffusion showing the cell density, MDE concentration and ECM density for t = 0,

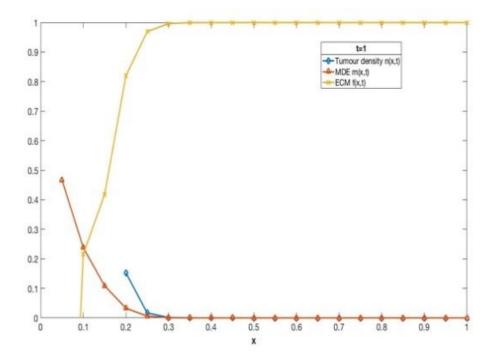


Figure 2 One-dimensional RDTM solution of system of Eq. (11)-(13) with constant tumor cell diffusion showing the cell density, MDE concentration and ECM density for t = 1,

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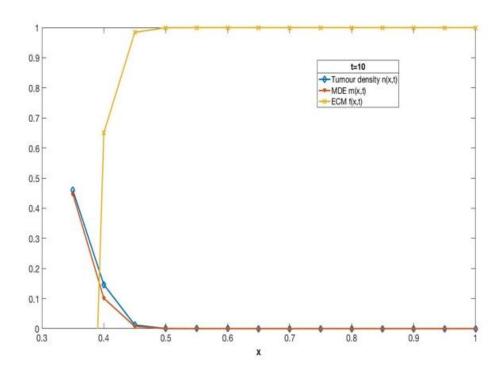


Figure 3 One-dimensional RDTM solution of system of Eq. (11)-(13) with constant tumor cell diffusion showing the cell density, MDE concentration and ECM density for t = 10,

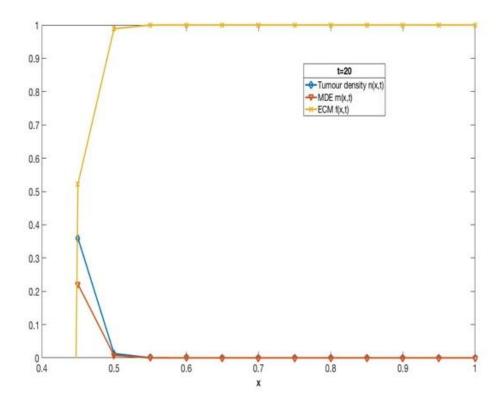


Figure 4 One-dimensional RDTM solution of system of Eq. (11)-(13) with constant tumor cell diffusion showing the cell density, MDE concentration and ECM density for t = 20,

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5. RESULT AND CONCLUSIONS

Figure 1-4 show four snapshots in time of the tumor cell density, ECM density, and MDE concentration. Initially, by t = 1 cancer cells have migrated a small distance into the domain. By t = 10 cancer cells have migrated almost half way through the domain. And as time evolves by t = 20 cancer cells continue to migrate to regions where high ECM densities are situated. Also, the ECM profile shows clearly the degradation by the MDEs. As the MDEs degrade the ECM, the tumor cells invade via combination of diffusion and haptotaxis. The tumor density distribution shows a small cluster of cells built up at the leading edge of the tumor due to haptotactic migration. As time evolves, this cluster of cells migrates further from the tumor main body and continues to invade the ECM at slower rate.

In this paper, Reduced Differential Transform method is applied for the approximate solution of mathematical model of tumor invasion and metastasises. Results obtained from RDTM are very near to the results obtained from other Numerical methods like Adomian Decomposition Method and Homotopy Perturbation Method (Mahiddin, N. et al., 2014). As RDTM is very simple in computation compare to the other method, thus it is very effective and convenient. We suggest applying this method to solve other Biological problems.

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