

Formulating and Analysis of Mathematical Modeling in Intracellular Signaling Pathways

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Abstract

Because of the inherent stochastic, the flagging elements in a clonal populace of cells display cell-to-cell fluctuation at the single-cell level, which is particular from the populace normal elements. Oftentimes, stream cytometry is broadly used to secure the single-cell level estimations by blocking cytokine discharge with reagents, for example, Golgiplug™. In any case, Golgiplug™ can change the flagging elements, making estimations be deluding. Consequently, we built up a mathematical model to construe the normal single-cell elements dependent on the stream cytometry estimations within the sight of Golgiplug™ with saccharine (LPS)-instigated NFκB motioning for instance. Initial, a mathematical model was created dependent on the earlier learning. At that point, normal single-cell elements of two key atoms (TNFα and IκBα) in the NFκB flagging pathway were estimated through stream cytometry within the sight of Golgiplug™ to approve the model and expand its expectation exactness. In particular, a parameter determination and estimation plan chosen key model parameters and assessed their qualities. Inadmissible results from the parameter estimation guided consequent analyses and fitting model enhancements, and the refined model was aligned again through the parameter estimation. The surmised model had the option to make forecasts that were reliable with the exploratory estimations, which will be utilized to develop a semi-stochastic model later on.

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1. INTRODUCTION

To incorporate of different flagging pathways, their authoritative translation factors and downstream effectors qualities is required for cells to react to different sign they experience in their miniaturized scale condition. Hence, seeing how data is detected and handled by cells and the flagging pathways that are locked in by various upgrades can help explain cell practices and reactions. Regularly, cell signal elements and the reaction to boosts have been examined utilizing a mix of mathematical demonstrating and trial examination. A dominant part of these investigations has demonstrated cell motioning at the populace level and utilized populace arrived at the midpoint of estimations, for example, Western smears to deduce the elements of various proteins in the flagging pathway, just as the conceivable system structure of flagging pathways.

Notwithstanding, with ongoing advances in the capacity to quantify quality and protein articulation at the single-cell level (checked on in), it has gotten conceivable to break down flagging elements at the single-cell level. As opposed to the perceptions from populace normal investigations, the single-

cell studies have exhibited that individual cells in a clonal populace may react contrastingly to a similar improvement, and the populace level estimations could veil the transient elements of individual cells.

This fluctuation in the reactions of individual cells represents a test to their execution in science and prescription.

Along these lines, it is critical to comprehend the stochasticity and heterogeneity in the single-cell reactions that may be missed in populace arrived at the midpoint of estimations. Advances in exploratory instruments for single-cell examination have prompted a huge increment in single-cell ponders. In spite of these headways, it is as yet hard to contemplate the single-cell flagging elements because of complex connections at numerous levels between various proteins that are engaged with signal transduction. Computational displaying has been proposed as reciprocal ways to deal with conquer a portion of these constraints and increase bits of knowledge that can't be acquired exclusively through analyses.

Parameter Estimation

Since we added the Golgiplug module to the model created by Caldwell et al. 2012 the incorporated powerful model (the model displayed in and Equation (1)) was quantitatively adjusted by evaluating its parameters utilizing exploratory estimations because of various LPS focuses within the sight of Golgiplug. The model parameter estimates were assessed by limiting the contrast between the exploratory estimations and the model expectations of the protein focus. In this work, we utilized stream cytometry to gauge two key particles in the LPS-instigated NFκB flagging pathway: TNFα and IκBα. Since stream cytometry doesn't give direct estimations of protein focus, the mean fluorescence force (MFI), which is a proportion of the quantity of duplicates of the objective atom per cell, was utilized to derive the protein fixation by expecting a straight connection among MFI and protein fixation. The test information and model expectation were thought about dependent on crease changes of MFI, which are characterized as pursues:

$$y_{I\kappa B\alpha}(t) = \frac{(x_{I\kappa B\alpha}(t) + x_{I\kappa B\alpha N}(t) + x_{NF\kappa B-I\kappa B\alpha}(t) + x_{NF\kappa B-I\kappa B\alpha N}(t))}{(x_{I\kappa B\alpha,0} + x_{I\kappa B\alpha,0} + x_{NF\kappa B-I\kappa B\alpha,0} + x_{NF\kappa B-I\kappa B\alpha,0})} \approx \frac{I_{I\kappa B\alpha}(t) - I_{I\kappa B\alpha,C}}{I_{I\kappa B\alpha,0} - I_{I\kappa B\alpha,C}}$$

$$y_{TNF\alpha}(t) = \frac{x_{TNF\alpha}(t)}{x_{TNF\alpha,0}} \approx \frac{I_{TNF\alpha}(t) - I_{TNF\alpha,C}}{I_{TNF\alpha,0} - I_{TNF\alpha,C}}$$

One of the biggest challenges in estimating parameters of signaling pathways with a large number of parameters is the parameter identifiability issue. That is, the exact values of some model parameters cannot be uniquely determined from experimental measurement even if a large amount of experimental measurements are available. As the proposed model has a large number of parameters, not all the model parameters can be estimated.

$$S_i = \begin{bmatrix} \frac{\partial y_i(t_1)}{\partial \theta_1} & \dots & \frac{\partial y_i(t_1)}{\partial \theta_{n_p}} \\ \vdots & \ddots & \vdots \\ \frac{\partial y_i(t_{N_i})}{\partial \theta_1} & \dots & \frac{\partial y_i(t_{N_i})}{\partial \theta_{n_p}} \end{bmatrix}, \quad \forall i = \{I\kappa B\alpha, TNF\alpha\}$$

where n_p is the number of parameters in θ in Equation (2), and $\partial y_i(t_l) / \partial \theta_j$ quantifies the effect of a parameter θ_j on an output y_i at $t = t_l$, $\forall l = 1, \dots, N_t$, where N_t is the number of measurement instants. $\partial y_i(t_l) / \partial \theta_j$ can be computed by the following equation:

$$\frac{\partial y_i(t_l)}{\partial \theta_j} = \frac{\partial g_i(t_l)}{\partial x^T} \frac{\partial x}{\partial \theta_j} + \frac{\partial g_i(t_l)}{\partial \theta_j}$$

Additionally, the term $\partial x / \partial \theta_j$ in Equation (5) can be computed by integrating the following

$$\frac{d}{dt} \frac{\partial x(t_l)}{\partial \theta_j} = \frac{\partial f(t_l)}{\partial x^T} \frac{\partial x}{\partial \theta_j} + \frac{\partial f(t_l)}{\partial \theta_j}$$

equation along with Equation (2)

Intracellular signaling

Essential cell forms (development, division, motility and so on.) are driven by intracellular and intercellular correspondence. The last is performed through direct contact between cells or solvent elements like hormones or cytokines. Ecological signs are gotten by on-film portions of the receptors, which initiate a course of biochemical responses in cell cytoplasm and

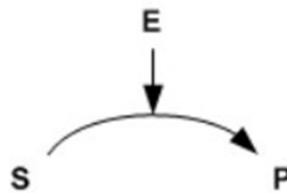


Figure 1. Standard graphical notation for a catalyzed (by enzyme) reaction

Dynamics of the biochemical pathways

The most popular mathematical basis for modelling biochemical reactions network dynamics are Continuous-Time Markov Chain (CTMC), that describes a system in a simplified

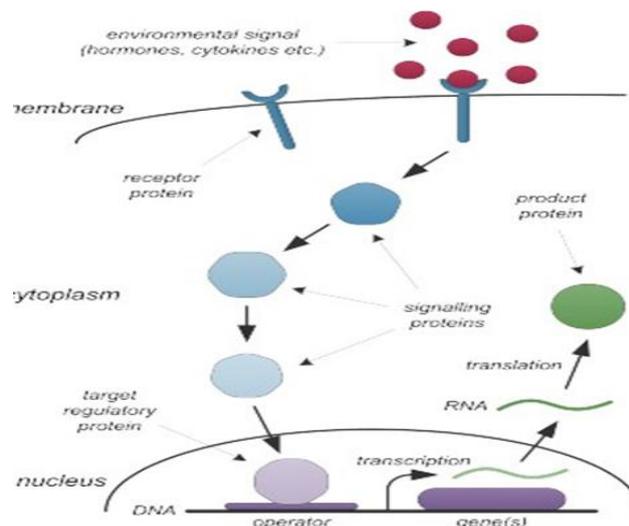


Figure 2: Simple generic scheme of a cellular signaling pathway altering target gene expression.

Mathematical modeling of signaling pathways

For complex organic frameworks, for example, malignant growth, a methodology that has been utilized to comprehend the procedures of multiplication, apoptosis or separation in them is the foundation of mathematical models to portray flagging procedure and the communication between the flagging atoms. In the wake of picking the organic pathway to be portrayed, the biochemical responses and their energy are depicted with the assistance of conventional differential conditions. (Tributes). As detailed by Orton, while building up the mathematical model it is often required to make a few presumptions to disentangle complex procedures. After the suspicions being made, it is important to depict the motor parameters, the rate constants, and the underlying focus. The meaning of energy pursues either the mass-activity or the approach. The advantage of utilizing mathematical models is that it encourages researchers to increase new experiences into cases which are hard to be tended to by trial draws near. Pathways are probably not going to be direct. They rather have circles -feed forward and input or cross-talks between them where particles of a pathway are associated with different atoms from an alternate pathway. From past examinations, it is outstanding that sign transduction is one of the fundamental trifling assignments which are identified with numerous perplexing maladies, for example, malignant growth. Hence, mathematical models can unwind conceivable convoluted practices in an organic framework.

2. REVIEW OF LITERATURE

Hence, with the difference in the cell science and progressing cell pathways coming about because of the diverse ailment condition, glycosylation can be adjusted essentially. Such illness related changed glycosylation can happen it is possible that either of the two different ways: I) the locales can be hyper or hypo glycosylated as well as new destinations can be glycosylated, ii) the glycan moiety of a similar glycosylation site can be modified [Pan Setal. 2011; Brooks SA, 2019].

Consequently, focusing on these redox-tweaking biochemical pathways of malignancy cells are plausible, potential and promising restorative methodology that may empower helpful selectivity and to beat sedate opposition. Because of this explanation, such a large number of redox framework focusing on medication are rising with the potential any expectation of malignancy the board [Trachootham Det al. 2009; Pervaiz S et al. 2004; Rohlena J et al. 2013; Fulda Setal. 2010].

Late trial work investigated the cell reaction to iperoxo-instigated M22 receptor incitement in Chinese hamster ovary (CHO) cells (Kruse et al., 2013; Schrage et al., 2013, 2015). The cell reaction was estimated by unique mass redistribution (DMR), a system to measure the intracellular mass development through optical thickness (Schröder et al., 2011). Since the DMR reaction can be thought to be reliant on the M22 receptor-subordinate flagging our point was to demonstrate and ponder the comparing response framework. The pathway comprises of proteins just as the optional errand person cyclic adenosine monophosphate (cAMP). The individual biochemical responses are chiefly outstanding (Pierce et al., 2012; Linderman, 2019; Sunahara and Taussig, 2012; Taylor et al., 2012), however as far as we could possibly know no exertion has been taken so far to infer a mathematical model, particularly for CHO cells, which are significant in pharmaceutical research and for the mechanical creation of recombinant protein therapeutics (De Jesus and Wurm, 2011; Walsh, 2015).

In this work, we built up a mass activity based mathematical portrayal of the M22 receptor-subordinate flagging system. Our created model comprises of 79 responses, out and out including 64 significant proteins and auxiliary flag-bearer particles portrayed in writing. In our joint flagging and optional flag-bearer model, all authoritative and (de-) phosphorylation occasions are unequivocally considered so as to empower resulting stoichiometric grid and transition dispersion examination (Wiback et al., 2004). In spite of the fact that this sort of examination is normally utilized for metabolic systems, our unequivocal displaying of official and phosphorylation occasions empower the adaption of these strategies to a blended flagging and optional emissary framework. The value of applying stoichiometric grid examination procedures to flagging pathways has for example been shown by Behre and Schuster (2009), who adjusted basic motion mode (EFM) investigation to this circumstance. We here show, how the realized motion examining procedure (Smith, 1996) can be stretched out to fuse halfway accessible test data (here: cAMP creation, phosphodiesterase 4 (PDE4) initiation). We tried our joined displaying and information driven examining strategy by anticipating key flagging instruments known from writing, yet not expressly encoded into the model. Our proposed requirement transition examining (CFS) strategy takes into consideration subjective forecasts of downstream incitement consequences for actin and tubulin levels, which here fill in as markers for the mass redistribution impact. These subjective forecasts are in concurrence with the trial perceptions, which proposes CFS as a procedure for model checking. This is additionally underlined by the likelihood to join CFS and EFM investigation yielding a measurable positioning of EFM as per their normal organic significance.

GPCR-instigated flagging is outstanding in like manner writing (Pierce et al., 2002; Linderman, 2009; Taylor et al., 2012; Sunahara and Taussig, 2012). Explicitly the connection to the cyclic AMP (cAMP) an optional ambassador particle)-initiated flagging is in the focal point of flow pharmaceutical research (Milligan and Kostenis, 2016; Hu et al., 2010). Figure 1 portrays a schematic portrayal of the entire arrangement of significant atoms and their interchange, which are considered in our model. Specifically, the procedure of receptor-initiated G protein (GP) enactment is all around examined, where the ligand-bound receptor changes its physical structure and the idle related GP collaborates with the receptor and separates into its subunits (Pierce et al., 2002). In this manner the alpha-I/alpha-s and beta/gamma subunits are initiated and can communicate autonomously with different proteins like adenylyl cyclase (AC) (Sunahara and Taussig, 2002; Milligan and Kostenis, 2016). The GP subunit alpha-o has no huge impact on AC however it has an effect on the DMR (Milligan and Kostenis, 2016). Air conditioning is one of the most significant proteins inside the GP-interceded pathway and liable for the optional delegate creation. The enormous number of AC and GP subtypes causes a profoundly perplexing sub-connect with many cross-responses (Milligan and Kostenis, 2016; Sunahara and Taussig, 2012). Likewise the receptor enactment cycle itself isn't insignificant. This initial phase in the flagging course is exceptionally fascinating for pharmaceutical research and prompted well-created models for receptor actuation and restraint (Woodroffe et al., 2019; Chen, 2013; Strange, 2019; Bornheimer et al., 2014).

3. OBJECTIVE OF THE STUDY

- [1] Identification of flagging and metabolic pathways change in EGFRvIII transformed condition
- [2] Visualization of changed flagging and metabolic pathways between association in a coordinated frame

worknaturalview in EGFRvIIItransformed condition

[3] Identification of potential ways of between association between changed flagging pathways with metabolic pathways under EGFRvIII condition

3. MATERIAL AND METHODS

Materials

Cell Culture RAW264.7 cells were acquired from ATCC (Manassas, VA, USA). Dulbecco's Modified Eagle Medium (DMEM) and penicillin/streptomycin were gotten from Invitrogen (Carlsbad, CA, USA). Ox-like serum and fetal ox-like serum (FBS) were acquired from Atlanta Biologicals (Flowery Branch, GA, USA). Ultrapure LPS got from *S. minnesota* was gotten from Invivogen (San Diego, CA, USA). RAW264.7 macrophages were refined in DMEM enhanced with 10% FBS, penicillin (200U/mL) and streptomycin (200 µg/mL) at 37 °C in a 5% CO₂ condition.

Flow Cytometry Analysis

The declaration of TNF α and I κ B α under various exploratory conditions was resolved utilizing stream cytometry. RAW264.7 cells were seeded into round-bottomed 96-well plate and animated with various groupings of LPS for the demonstrated time. Golgiplug (BD Biosciences, San Jose, CA, USA) was included alongside LPS for TNF α discovery trials to square discharge of TNF α . Cells were then recolored with Alexa Flour 700 fluorescence-labeled TNF α neutralizer (BD Biosciences) and PE-conjugated I κ B α counter acting agent (Cell Signaling Technology, Danvers, MA, USA) utilizing the producer's proposed convention. Recolored cells were broke down utilizing a BD Fortessa stream cytometer (BD Biosciences) at the Texas A&M Health Science Center College of Medicine Cell Analysis Facility. Ten thousands occasions for every example were procured, and the information were dissected utilizing FlowJo software (Tree Star, OR, USA).

of the TNF α -incited NF κ B flagging was incorporated into the adjusted model to down direct the LPS-initiated motioning through deubiquitinating of TRAF6.

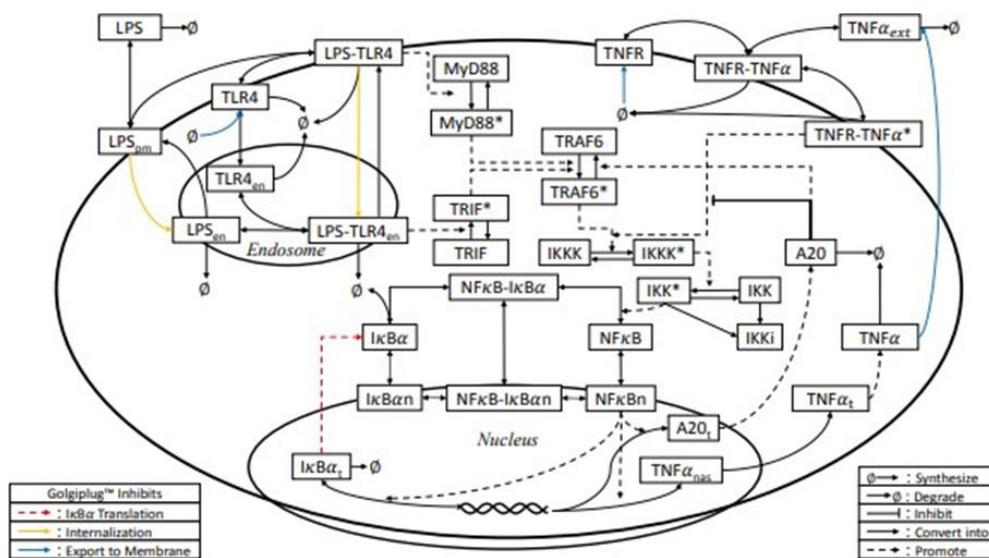


Figure 3. Schematic diagram for the LPS-NF κ B-TNF α signaling pathway. Due to space limitation, TRIF-dependent regulation of TNF α production, I κ B β and I κ B ϵ -dependent NF κ B deactivation and IF2 α -induced translation inhibition are not illustrated.

4. DATA ANALYSIS

Profiles of once more combined intracellular TNF α under the incitement of LPS within the sight of GolgiplugTM exhibited that the TNF α generation expanded around one hour after the incitement (Figure 2). At around a similar time, the I κ B α fixation arrived at its base, which is reliable with test perceptions in the writing [46–48]. Along these lines, the I κ B α focus expanded because of the enlistment of I κ B transcript (I κ Bt) by atomic interpretation of NF κ B, while the TNF α generation rate backed off past 4 h of LPS incitement (Figure 2). It ought to be noticed that no analyses were led past 6 h after LPS was added to the cell culture dependent on the producer's rule on Golgiplug use. This is in all likelihood dependent on the way that Golgiplug may incite the apoptosis of cells presented to it for quite a while [49,50]. Subsequently, the aligned model is increasingly reasonable to portray the early NF κ B flagging pathway (≤ 6 h) upon the LPS incitement.

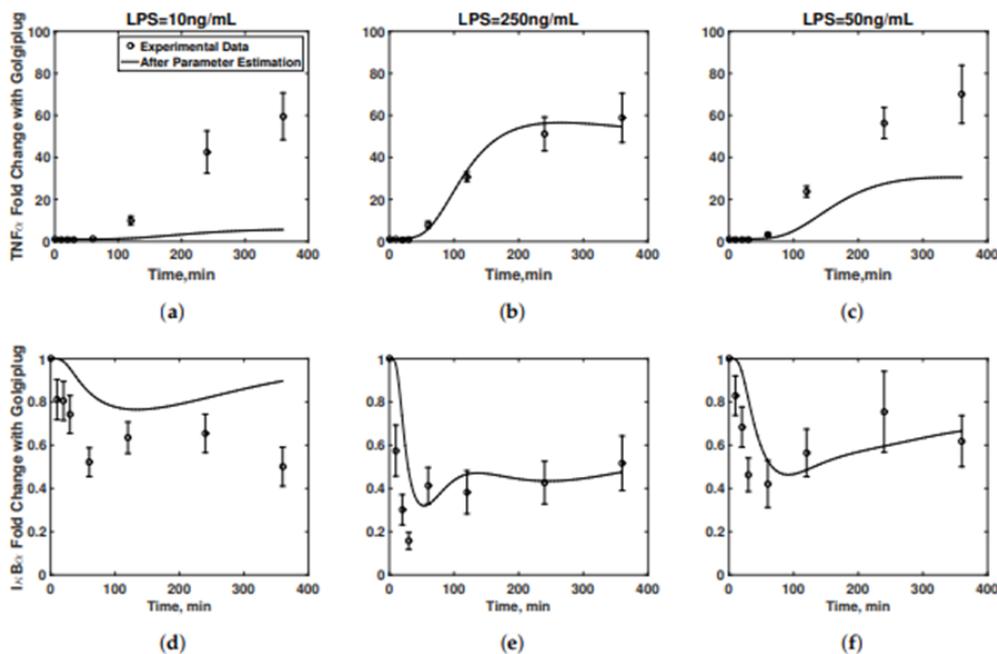


Figure 4. Parameter estimation before considering the GolgiplugTM-induced ER stress. (a–c) Measured (empty circle) and simulated (solid line) fold changes of intracellular TNF α concentrations over time were plotted under different LPS concentrations in the presence of Golgiplug

5. CONCLUSIONS

We fundamentally removed the normal single-cell elements of the LPS-instigated NF κ B flagging pathway through the mix of affectability investigation and a parameter choice plan with stream cytometry information of key protein intermediates. In light of the estimations and the model structure, key model parameters were distinguished and evaluated to boost the expectation precision of the adjusted model while abstaining from over fitting. The befuddle between the model forecasts and exploratory perceptions significantly after the parameter estimation uncovered the presence of a formerly unconsidered, yet significant, component identified with Golgiplug which was thusly approved by examinations and prompted the update of the proposed model. At that point, the resultant model was approved, and there produced profiles from the refreshed model were in great concurrence with exploratory datasets under three unique LPS fixations. This model can be utilized as the ostensible model to develop a deterministic model that has parameters with conveyances and can be utilized to

ponder the stochasticity in flagging.

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